

# Markov Decision Processes with Rewards

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## Abstract

We present a formalization of Markov Decision Processes with rewards. In particular we first build on Hölzl's formalization [1] of MDPs and extend them with rewards. We proceed with an analysis of the expected total discounted reward criterion for infinite horizon MDPs. The central result is the construction of the iteration rule for the Bellman operator. We prove the optimality equations for this operator and show the existence of an optimal stationary deterministic solution. The analysis can be used to obtain dynamic programming algorithms such as value iteration and policy iteration to solve Markov Decision Processes with formal guarantees. Our formalization is based upon chapters 5 and 6 in Puterman's book [2].

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## 1 Bounded Functions

**theory** *Bounded-Functions*

**imports**

*HOL.Topological-Spaces*

*HOL-Analysis.Uniform-Limit*

*HOL-Probability.Probability*

**begin**

### 1.1 Definition

**definition**  $bfun = \{f. \text{bounded } (range\ f)\}$

**typedef** (**overloaded**) ( $'a, 'b$ )  $bfun$  ( $\langle(- \Rightarrow_b -)\rangle$  [22] 21) =

$bfun::('a \Rightarrow 'b :: \text{metric-space}) \text{ set}$

**morphisms**  $apply-bfun$   $Bfun$

$\langle proof \rangle$

**declare** [ $coercion$   $apply-bfun :: ('a \Rightarrow_b ('b :: \text{metric-space})) \Rightarrow 'a \Rightarrow 'b$ ]]

**setup-lifting**  $type-definition-bfun$

**lemma**  $bounded-apply-bfun[intro, simp]: bounded ((apply-bfun\ x) \text{ ' } X)$   
 $\langle proof \rangle$

**lemma**  $apply-bfun-bdd-above[simp, intro]:$

**fixes**  $f :: 'c \Rightarrow_b \text{real}$

**shows**  $bdd-above (f \text{ ' } X)$

$\langle proof \rangle$

**lemma**  $bfun-eqI[intro]: (\bigwedge x. apply-bfun\ f\ x = apply-bfun\ g\ x) \Longrightarrow f = g$

$\langle proof \rangle$

**lemma**  $bfun-eqD[dest]: f = g \Longrightarrow (\bigwedge x. apply-bfun\ f\ x = apply-bfun\ g\ x)$

$\langle proof \rangle$

**lemma**  $bfunE:$

**assumes**  $f \in \text{bfun}$   
**obtains**  $g$  where  $f = \text{apply-bfun } g$   
 $\langle \text{proof} \rangle$

**lemma** *const-bfun*:  $(\lambda x. b) \in \text{bfun}$   
 $\langle \text{proof} \rangle$

**lift-definition** *const-bfun*:: $'b \Rightarrow ('a \Rightarrow_b ('b :: \text{metric-space}))$  **is**  $\lambda(c::'b)$   
 $\cdot. c$   
 $\langle \text{proof} \rangle$

**lemma** *bounded-dist-le-SUP-dist*:  
 $\text{bounded } (\text{range } f) \Longrightarrow \text{bounded } (\text{range } g) \Longrightarrow \text{dist } (f x) (g x) \leq (\text{SUP } x. \text{dist } (f x) (g x))$   
 $\langle \text{proof} \rangle$

**instantiation** *bfun* ::  $(\text{type}, \text{metric-space}) \text{ metric-space}$   
**begin**

**lift-definition** *dist-bfun* ::  $('a \Rightarrow_b 'b) \Rightarrow ('a \Rightarrow_b 'b) \Rightarrow \text{real}$   
**is**  $\lambda f g. (\text{SUP } x. \text{dist } (f x) (g x))$   $\langle \text{proof} \rangle$

**definition** *uniformity-bfun* ::  $(('a \Rightarrow_b 'b) \times 'a \Rightarrow_b 'b) \text{ filter}$   
**where** *uniformity-bfun* =  $(\text{INF } e \in \{0 < ..\}. \text{principal } \{(x, y). \text{dist } x y < e\})$

**definition** *open-bfun* ::  $('a \Rightarrow_b 'b) \text{ set} \Rightarrow \text{bool}$   
**where** *open-bfun*  $S = (\forall x \in S. \forall_F (x', y) \text{ in } \text{uniformity}. x' = x \longrightarrow y \in S)$

**lemma** *dist-bounded*:  
**fixes**  $f g :: 'a \Rightarrow_b 'b$   
**shows**  $\text{dist } (f x) (g x) \leq \text{dist } f g$   
 $\langle \text{proof} \rangle$

**lemma** *dist-bound*:  
**fixes**  $f g :: 'a \Rightarrow_b ('b :: \text{metric-space})$   
**assumes**  $\bigwedge x. \text{dist } (f x) (g x) \leq b$   
**shows**  $\text{dist } f g \leq b$   
 $\langle \text{proof} \rangle$

**lemma** *dist-fun-lt-imp-dist-val-lt*:  
**fixes**  $f g :: 'a \Rightarrow_b 'b$   
**assumes**  $\text{dist } f g < e$   
**shows**  $\text{dist } (f x) (g x) < e$   
 $\langle \text{proof} \rangle$

**instance**  
 $\langle \text{proof} \rangle$

**end**

**lift-definition**  $PiC::'a \text{ set} \Rightarrow ('a \Rightarrow ('b :: \text{metric-space}) \text{ set}) \Rightarrow ('a \Rightarrow_b 'b) \text{ set}$   
**is**  $\lambda I X. Pi I X \cap \text{bfun}$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{mem-PiC-iff}: x \in PiC I X \longleftrightarrow \text{apply-bfun } x \in Pi I X$   
 $\langle \text{proof} \rangle$

**lemmas**  $\text{mem-PiCD} = \text{mem-PiC-iff}[\text{THEN iffD1}]$   
**and**  $\text{mem-PiCI} = \text{mem-PiC-iff}[\text{THEN iffD2}]$

**lemma**  $\text{tendsto-bfun-uniform-limit}$ :  
**fixes**  $f::'i \Rightarrow 'a \Rightarrow_b ('b :: \text{metric-space})$   
**assumes**  $(f \longrightarrow l) F$   
**shows**  $\text{uniform-limit UNIV } f \ l \ F$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{uniform-limit-tendsto-bfun}$ :  
**fixes**  $f::'i \Rightarrow 'a \Rightarrow_b ('b :: \text{metric-space})$   
**and**  $l::'a \Rightarrow_b 'b$   
**assumes**  $\text{uniform-limit UNIV } f \ l \ F$   
**shows**  $(f \longrightarrow l) F$   
 $\langle \text{proof} \rangle$

## 1.2 Supremum Norm

**instantiation**  $\text{bfun} :: (\text{type}, \text{real-normed-vector}) \text{ real-vector}$   
**begin**

**lemma**  $\text{uminus-cont}: f \in \text{bfun} \Longrightarrow (\lambda x. - f x) \in \text{bfun}$  **for**  $f::'a \Rightarrow 'b$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{plus-cont}: f \in \text{bfun} \Longrightarrow g \in \text{bfun} \Longrightarrow (\lambda x. f x + g x) \in \text{bfun}$   
**for**  $f g::'a \Rightarrow 'b$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{minus-cont}: f \in \text{bfun} \Longrightarrow g \in \text{bfun} \Longrightarrow (\lambda x. f x - g x) \in \text{bfun}$   
**for**  $f g::'a \Rightarrow 'b$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{scaleR-cont}: f \in \text{bfun} \Longrightarrow (\lambda x. a *_R f x) \in \text{bfun}$  **for**  $f::'a \Rightarrow 'b$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{bfun-normI}[\text{intro}]: (\bigwedge x. \text{norm } (f x) \leq b) \Longrightarrow f \in \text{bfun}$

*<proof>*

**lift-definition** *uminus-bfun*::('a  $\Rightarrow_b$  'b)  $\Rightarrow$  ('a  $\Rightarrow_b$  'b) **is**  $\lambda f x. - f x$   
*<proof>*

**lift-definition** *plus-bfun*::('a  $\Rightarrow_b$  'b)  $\Rightarrow$  ('a  $\Rightarrow_b$  'b)  $\Rightarrow$  'a  $\Rightarrow_b$  'b **is**  $\lambda f g x. f x + g x$   
*<proof>*

**lift-definition** *minus-bfun*::('a  $\Rightarrow_b$  'b)  $\Rightarrow$  ('a  $\Rightarrow_b$  'b)  $\Rightarrow$  'a  $\Rightarrow_b$  'b **is**  $\lambda f g x. f x - g x$   
*<proof>*

**lift-definition** *zero-bfun*::'a  $\Rightarrow_b$  'b **is**  $\lambda-. 0$   
*<proof>*

**lemma** *const-bfun-0-eq-0[simp]*: *const-bfun 0 = 0*  
*<proof>*

**lift-definition** *scaleR-bfun*::*real*  $\Rightarrow$  ('a  $\Rightarrow_b$  'b)  $\Rightarrow$  'a  $\Rightarrow_b$  'b **is**  $\lambda r g x. r *_R g x$   
*<proof>*

**lemmas** [*simp*] =  
*const-bfun.rep-eq*  
*uminus-bfun.rep-eq*  
*plus-bfun.rep-eq*  
*minus-bfun.rep-eq*  
*zero-bfun.rep-eq*  
*scaleR-bfun.rep-eq*

**instance**  
*<proof>*  
**end**

**lemma** *scaleR-cont'*:  $f \in \text{bfun} \Longrightarrow (\lambda x. a * f x) \in \text{bfun}$  **for**  $f :: 'a \Rightarrow \text{real}$   
*<proof>*

**lemma** *bfun-norm-le-SUP-norm*:  
 $f \in \text{bfun} \Longrightarrow \text{norm} (f x) \leq (\text{SUP } x. \text{norm} (f x))$   
*<proof>*

**instantiation** *bfun* :: (*type*, *real-normed-vector*) *real-normed-vector*  
**begin**

**definition** *norm-bfun* :: ('a, 'b) *bfun*  $\Rightarrow$  *real*  
**where** *norm-bfun f = dist f 0*

**definition**  $sgn (f::('a,'b) bfun) = f /_R norm f$

**instance**

$\langle proof \rangle$

**end**

**lemma**  $norm-bfun-def'$ :  $norm f = (\bigsqcup x. norm ((f :: 'a \Rightarrow_b 'b :: real-normed-vector) x))$

$\langle proof \rangle$

**lemma**  $norm-le-norm-bfun$ :  $norm (apply-bfun f x) \leq norm f$

$\langle proof \rangle$

**lemma**  $abs-le-norm-bfun$ :  $abs (apply-bfun f x) \leq norm f$

$\langle proof \rangle$

**lemma**  $le-norm-bfun$ :  $apply-bfun f x \leq norm f$

$\langle proof \rangle$

### 1.3 Complete Space

**lemma**  $tendsto-add$ :  $P \longrightarrow (L :: 'a :: real-normed-vector) \Longrightarrow (\lambda n.$

$P n + c) \longrightarrow L + c$

$\langle proof \rangle$

**lemma**  $lim-add$ :  $convergent P \Longrightarrow lim (\lambda n. P n + (c :: 'a :: real-normed-vector))$

$= lim P + c$

$\langle proof \rangle$

**lemma**  $complete-bfun$ :

**assumes**  $cauchy-f$ :  $Cauchy (f :: nat \Rightarrow ('a, 'b :: \{complete-space, real-normed-vector\}) bfun)$

**shows**  $convergent f$

$\langle proof \rangle$

**lemma**  $norm-bound$ :

**fixes**  $f :: ('a, 'b::real-normed-vector) bfun$

**assumes**  $\bigwedge x. norm (apply-bfun f x) \leq b$

**shows**  $norm f \leq b$

$\langle proof \rangle$

**lemma**  $bfun-bounded-norm-range$ :  $bounded (range (\lambda s. norm (apply-bfun v s)))$

$\langle proof \rangle$

**instance**  $bfun :: (type, banach) banach$

$\langle proof \rangle$

**lemma** *bfun-prob-space-integrable*:  
**assumes** *prob-space*  $S$   $v \in \text{borel-measurable } S$   
**assumes**  $(v :: 'a \Rightarrow 'b :: \{\text{second-countable-topology, banach}\}) \in \text{bfun}$   
**shows** *integrable*  $S$   $v$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-integral-bound*:  
**assumes**  $(v :: 'a \Rightarrow 'c :: \{\text{euclidean-space}\}) \in \text{bfun}$   
**shows**  $(\lambda S. \int x. v \ x \ \partial(S :: 'a \text{ pmf})) \in \text{bfun}$   
 $\langle \text{proof} \rangle$

**lemma** *scale-bfun[intro]*:  $f \in \text{bfun} \Longrightarrow (\lambda x. (k :: \text{real}) * f \ x) \in \text{bfun}$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-spec[intro]*:  $f \in \text{bfun} \Longrightarrow (\lambda x. f \ (g \ x)) \in \text{bfun}$   
 $\langle \text{proof} \rangle$

**lemma** *apply-bfun-bfun[simp]*: *apply-bfun*  $f \in \text{bfun}$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-integral-bound'[intro]*:  $(v :: 'a \Rightarrow 'c :: \{\text{euclidean-space}\}) \in \text{bfun} \Longrightarrow$   
 $(\lambda S. \int x. v \ x \ \partial((F \ S) :: 'a \ \text{pmf})) \in \text{bfun}$   
 $\langle \text{proof} \rangle$

**lift-definition** *bfun-comp* ::  $('a \Rightarrow 'b) \Rightarrow ('b \Rightarrow_b 'c :: \text{metric-space}) \Rightarrow$   
 $('a \Rightarrow_b 'c)$  **is**  
 $\lambda g \ \text{bf} \ x. \ \text{bf} \ (g \ x)$   
 $\langle \text{proof} \rangle$

## 1.4 Order Instance

**class** *ordered-real-normed-vector* = *real-normed-vector* + *ordered-real-vector*

**instance** *real* :: *ordered-real-normed-vector*  
 $\langle \text{proof} \rangle$

**instantiation** *bfun* ::  $(-, \text{ordered-real-normed-vector}) \text{ ordered-real-normed-vector}$   
**begin**

**definition** *less-eq-bfun*  $f \ g \equiv \forall x. \ \text{apply-bfun} \ f \ x \leq \ \text{apply-bfun} \ g \ x$

**definition** *less-bfun*  $f \ g \equiv \forall x. \ \text{apply-bfun} \ f \ x \leq \ \text{apply-bfun} \ g \ x \wedge (\exists y. \ f \ y < g \ y)$

**instance**  
 $\langle \text{proof} \rangle$   
**end**



**lemma** *less-eq-bfunI*[*intro*]:  $(\bigwedge x. \text{apply-bfun } f \ x \leq \text{apply-bfun } g \ x) \implies f \leq g$   
 ⟨*proof*⟩

**lemma** *less-eq-bfunD*[*dest*]:  $f \leq g \implies (\bigwedge x. \text{apply-bfun } f \ x \leq \text{apply-bfun } g \ x)$   
 ⟨*proof*⟩

## 1.5 Miscellaneous

**instantiation** *bfun* :: (*type*, *one*) *one* **begin**

**lift-definition** *one-bfun* :: '*s*  $\Rightarrow_b$  '*d*::{*metric-space*, *one*} **is**  $\lambda x. 1$   
 ⟨*proof*⟩

**instance**  
 ⟨*proof*⟩  
**end**

**declare** *one-bfun.rep-eq* [*simp*]

**lemma** *apply-bfun-one* [*simp*]:  $\text{apply-bfun } (1 :: - \Rightarrow_b \text{real}) \ x = 1$   
 ⟨*proof*⟩

**lemma** *norm-bfun-one*[*simp*]:  $\text{norm } (1 :: 'a \Rightarrow_b \text{real}) = 1$   
 ⟨*proof*⟩

**lemma** *range-bfunI*[*intro*]:  $\text{bounded } (\text{range } f) \implies f \in \text{bfun}$   
 ⟨*proof*⟩

**lemma** *finite-bfun*[*simp*]:  $(\lambda(i:::\text{finite}). f \ i) \in \text{bfun}$   
 ⟨*proof*⟩

**lemma** *bounded-apply-bfun'*:  
**assumes** *bounded* ((*F* :: '*c*  $\Rightarrow$  '*d*  $\Rightarrow_b$  '*b*::*real-normed-vector*) '*S*)  
**shows** *bounded* (( $\lambda b. (F \ b) \ x$ ) '*S*)  
 ⟨*proof*⟩

**lemma** *bfun-tendsto-apply-bfun*:  
**assumes** *h*: (*F* :: (*nat*  $\Rightarrow$  '*a*  $\Rightarrow_b$  *real*))  $\longrightarrow$  (*y* :: '*a*  $\Rightarrow_b$  *real*)  
**shows** ( $\lambda n. F \ n \ x$ )  $\longrightarrow$  *y* *x*  
 ⟨*proof*⟩

## 1.6 Bounded Functions and Vectors

**lemma** *vec-bfun*[*simp*, *intro*]:  $(\$) \ x \in \text{bfun}$   
 ⟨*proof*⟩

**lemma** *norm-bfun-le-norm-vec*:  $\text{norm } (\text{bfun.Bfun } ((\$) (x :: \text{real}^c :: \text{finite}))) \leq \text{norm } x$   
 ⟨proof⟩

**lemma** *bounded-linear-bfun-nth*:  $\text{bounded-linear } f \implies \text{bounded-linear } (\lambda v. \text{bfun.Bfun } ((\$) (f v)))$   
 ⟨proof⟩

**lemma** *norm-vec-le-norm-bfun*:  
 $\text{norm } (\text{vec-lambda } (\text{apply-bfun } (x :: 'd::\text{finite} \Rightarrow_b \text{real}))) \leq \text{norm } x * \text{card } (\text{UNIV} :: 'd \text{ set})$   
 ⟨proof⟩

end

## 2 Bounded Linear Functions

**theory** *Blinfun-Util*  
**imports**  
   *HOL-Analysis.Bounded-Linear-Function*  
   *Bounded-Functions*  
**begin**

### 2.1 Composition

**lemma** *blinfun-compose-id[simp]*:  
 $\text{id-blinfun } o_L f = f$   
 $f o_L \text{id-blinfun} = f$   
 ⟨proof⟩

**lemma** *blinfun-compose-assoc*:  $F o_L G o_L H = F o_L (G o_L H)$   
 ⟨proof⟩

**lemma** *blinfun-compose-diff-right*:  $f o_L (g - h) = (f o_L g) - (f o_L h)$   
 ⟨proof⟩

### 2.2 Power

**overloading**  
 $\text{blinfunpow} \equiv \text{compow} :: \text{nat} \Rightarrow ('a::\text{real-normed-vector} \Rightarrow_L 'a) \Rightarrow ('a \Rightarrow_L 'a)$   
**begin**

**primrec** *blinfunpow* ::  $\text{nat} \Rightarrow ('a::\text{real-normed-vector} \Rightarrow_L 'a) \Rightarrow ('a \Rightarrow_L 'a)$   
**where**  
    $\text{blinfunpow } 0 f = \text{id-blinfun}$   
    $|\ \text{blinfunpow } (\text{Suc } n) f = f o_L \text{blinfunpow } n f$

end

**lemma** *bounded-pow-blinfun*[intro]:

**assumes** *bounded* (*range* ( $F :: \text{nat} \Rightarrow 'a :: \text{real-normed-vector} \Rightarrow_L 'a$ ))  
**shows** *bounded* (*range* ( $\lambda t. (F t) \widehat{\sim} (\text{Suc } n)$ ))  
*<proof>*

**lemma** *blincomp-scaleR-right*: ( $a *_R (F :: 'a :: \text{real-normed-vector} \Rightarrow_L 'a)$ )  $\widehat{\sim} t = a \widehat{\sim} t *_R F \widehat{\sim} t$

*<proof>*

**lemma** *summable-inv-Q*:

**fixes**  $Q :: 'a :: \text{banach} \Rightarrow_L 'a$   
**assumes** *onorm-le*:  $\text{norm } (id\text{-blinfun} - Q) < 1$   
**shows** *summable* ( $\lambda n. (id\text{-blinfun} - Q) \widehat{\sim} n$ )  
*<proof>*

**lemma** *blinfunpow-assoc*: ( $F :: 'a :: \text{real-normed-vector} \Rightarrow_L 'a$ )  $\widehat{\sim} (\text{Suc } n) = (F \widehat{\sim} n) o_L F$   
*<proof>*

**lemma** *norm-blinfunpow-le*:  $\text{norm } ((f :: 'b :: \text{real-normed-vector} \Rightarrow_L 'b) \widehat{\sim} n) \leq \text{norm } f \widehat{\sim} n$   
*<proof>*

**lemma** *blinfunpow-nonneg*:

**assumes**  $\bigwedge v. 0 \leq v \implies 0 \leq \text{blinfun-apply } (f :: ('b :: \{\text{ord}, \text{real-normed-vector}\} \Rightarrow_L 'b)) v$   
**shows**  $0 \leq v \implies 0 \leq (f \widehat{\sim} n) v$   
*<proof>*

**lemma** *blinfunpow-mono*:

**assumes**  $\bigwedge u v. u \leq v \implies (f :: 'b :: \{\text{ord}, \text{real-normed-vector}\} \Rightarrow_L 'b) u \leq f v$   
**shows**  $u \leq v \implies (f \widehat{\sim} n) u \leq (f \widehat{\sim} n) v$   
*<proof>*

**lemma** *banach-blinfun*:

**fixes**  $C :: 'b :: \{\text{real-normed-vector}, \text{complete-space}\} \Rightarrow_L 'b$   
**assumes** *norm*  $C < 1$   
**shows**  $\exists! v. C v = v \bigwedge v. (\lambda n. (C \widehat{\sim} n) v) \longrightarrow (\text{THE } v. C v = v)$   
*<proof>*

## 2.3 Geometric Sum

**lemma** *inv-one-sub-Q*:

**fixes**  $Q :: 'a :: \text{banach} \Rightarrow_L 'a$   
**assumes** *onorm-le*:  $\text{norm } (id\text{-blinfun} - Q) < 1$   
**shows**  $(Q o_L (\sum i. (id\text{-blinfun} - Q) \widehat{\sim} i)) = id\text{-blinfun}$   
**and**  $(\sum i. (id\text{-blinfun} - Q) \widehat{\sim} i) o_L Q = id\text{-blinfun}$

$\langle \text{proof} \rangle$

**lemma** *inv-norm-le*:

**fixes**  $Q :: 'a :: \text{banach} \Rightarrow_L 'a$

**assumes**  $\text{norm } Q < 1$

**shows**  $(\text{id-blinfun} - Q) \circ_L (\sum i. Q \hat{\sim} i) = \text{id-blinfun}$

$(\sum i. Q \hat{\sim} i) \circ_L (\text{id-blinfun} - Q) = \text{id-blinfun}$

$\langle \text{proof} \rangle$

**lemma** *inv-norm-le'*:

**fixes**  $Q :: 'a :: \text{banach} \Rightarrow_L 'a$

**assumes**  $\text{norm } Q < 1$

**shows**  $(\text{id-blinfun} - Q) ((\sum i. Q \hat{\sim} i) x) = x$

$(\sum i. Q \hat{\sim} i) ((\text{id-blinfun} - Q) x) = x$

$\langle \text{proof} \rangle$

## 2.4 Inverses

**definition**  $\text{is-inverse}_L X Y \iff X \circ_L Y = \text{id-blinfun} \wedge Y \circ_L X = \text{id-blinfun}$

**abbreviation**  $\text{invertible}_L X \equiv \exists X'. \text{is-inverse}_L X X'$

**lemma** *is-inverse<sub>L</sub>-I[intro]*:

**assumes**  $X \circ_L Y = \text{id-blinfun}$   $Y \circ_L X = \text{id-blinfun}$

**shows**  $\text{is-inverse}_L X Y$

$\langle \text{proof} \rangle$

**lemma** *is-inverse<sub>L</sub>-D[dest]*:

**assumes**  $\text{is-inverse}_L X Y$

**shows**  $X \circ_L Y = \text{id-blinfun}$   $Y \circ_L X = \text{id-blinfun}$

$\langle \text{proof} \rangle$

**lemma** *invertible<sub>L</sub>-D[dest]*:

**assumes**  $\text{invertible}_L f$

**obtains**  $g$  **where**  $f \circ_L g = \text{id-blinfun}$   $g \circ_L f = \text{id-blinfun}$

$\langle \text{proof} \rangle$

**lemma** *invertible<sub>L</sub>-I[intro]*:

**assumes**  $f \circ_L g = \text{id-blinfun}$   $g \circ_L f = \text{id-blinfun}$

**shows**  $\text{invertible}_L f$

$\langle \text{proof} \rangle$

**lemma** *is-inverse<sub>L</sub>-comm*:  $\text{is-inverse}_L X Y \iff \text{is-inverse}_L Y X$

$\langle \text{proof} \rangle$

**lemma** *is-inverse<sub>L</sub>-unique*:  $\text{is-inverse}_L f g \implies \text{is-inverse}_L f h \implies g = h$

$\langle \text{proof} \rangle$

**lemma** *is-inverse<sub>L</sub>-ex1*:  $is\_inverse_L f g \implies \exists! h. is\_inverse_L f h$   
 ⟨proof⟩

**lemma** *is-inverse<sub>L</sub>-ex1'*:  $\exists x. is\_inverse_L f x \implies \exists! x. is\_inverse_L f x$   
 ⟨proof⟩

**definition**  $inv_L f = (THE g. is\_inverse_L f g)$

**lemma** *inv<sub>L</sub>-eq*:  
**assumes**  $is\_inverse_L f g$   
**shows**  $inv_L f = g$   
 ⟨proof⟩

**lemma** *inv<sub>L</sub>-I*:  
**assumes**  $f o_L g = id\_blinfun g o_L f = id\_blinfun$   
**shows**  $g = inv_L f$   
 ⟨proof⟩

**lemma** *inv-app1[simp]*:  $invertible_L X \implies (inv_L X) o_L X = id\_blinfun$   
 ⟨proof⟩

**lemma** *inv-app2[simp]*:  $invertible_L X \implies X o_L (inv_L X) = id\_blinfun$   
 ⟨proof⟩

**lemma** *inv-app1'[simp]*:  $invertible_L X \implies inv_L X (X v) = v$   
 ⟨proof⟩

**lemma** *inv-app2'[simp]*:  $invertible_L X \implies X (inv_L X v) = v$   
 ⟨proof⟩

**lemma** *inv<sub>L</sub>-inv<sub>L</sub>[simp]*:  $invertible_L X \implies inv_L (inv_L X) = X$   
 ⟨proof⟩

**lemma** *inv<sub>L</sub>-cancel-iff*:  
**assumes**  $invertible_L f$   
**shows**  $f x = y \iff x = inv_L f y$   
 ⟨proof⟩

**lemma** *invertible<sub>L</sub>-inf-sum*:  
**assumes**  $norm (X :: 'b :: banach \Rightarrow_L 'b) < 1$   
**shows**  $invertible_L (id\_blinfun - X)$   
 ⟨proof⟩

**lemma** *inv<sub>L</sub>-inf-sum*:  
**fixes**  $X :: 'b :: banach \Rightarrow_L -$   
**assumes**  $norm X < 1$   
**shows**  $inv_L (id\_blinfun - X) = (\sum i. X \hat{=} i)$   
 ⟨proof⟩

**lemma** *is-inverse<sub>L</sub>-compose*:

**assumes** *invertible<sub>L</sub> f invertible<sub>L</sub> g*

**shows** *is-inverse<sub>L</sub> (f o<sub>L</sub> g) (inv<sub>L</sub> g o<sub>L</sub> inv<sub>L</sub> f)*

*<proof>*

**lemma** *invertible<sub>L</sub>-compose: invertible<sub>L</sub> f  $\implies$  invertible<sub>L</sub> g  $\implies$  invertible<sub>L</sub> (f o<sub>L</sub> g)*

*<proof>*

**lemma** *inv<sub>L</sub>-compose*:

**assumes** *invertible<sub>L</sub> f invertible<sub>L</sub> g*

**shows** *inv<sub>L</sub> (f o<sub>L</sub> g) = (inv<sub>L</sub> g) o<sub>L</sub> (inv<sub>L</sub> f)*

*<proof>*

**lemma** *inv<sub>L</sub>-id-blinfun[simp]: inv<sub>L</sub> id-blinfun = id-blinfun*

*<proof>*

## 2.5 Norm

**lemma** *bounded-range-subset*:

*bounded (range f :: real set)  $\implies$  bounded (f ‘ X)*

*<proof>*

**lemma** *bounded-const: bounded ((λ-. x) ‘ X)*

*<proof>*

**lift-definition** *bfun-pos :: ('d  $\Rightarrow_b$  real)  $\Rightarrow$  ('d  $\Rightarrow_b$  real) is λf i. if f i < 0 then -f i else f i*

*<proof>*

**lemma** *bfun-pos-zero[simp]: bfun-pos f = 0  $\longleftrightarrow$  f = 0*

*<proof>*

**lift-definition** *bfun-nonneg :: ('d  $\Rightarrow_b$  real)  $\Rightarrow$  ('d  $\Rightarrow_b$  real) is λf i. if f i ≤ 0 then 0 else f i*

*<proof>*

**lemma** *bfun-nonneg-split: bfun-nonneg x - bfun-nonneg (- x) = x*

*<proof>*

**lemma** *blinfun-split: blinfun-apply f x = f (bfun-nonneg x) - f (bfun-nonneg (- x))*

*<proof>*

**lemma** *bfun-nonneg-pos: bfun-nonneg x + bfun-nonneg (-x) = bfun-pos x*

*<proof>*

**lemma** *bfun-nonneg*:  $0 \leq \text{bfun-nonneg } f$   
 ⟨proof⟩

**lemma** *bfun-pos-eq-nonneg*:  $\text{bfun-pos } n = \text{bfun-nonneg } n + \text{bfun-nonneg } (-n)$   
 ⟨proof⟩

**lemma** *blinfun-mono-norm-pos*:  
**fixes**  $f :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('d \Rightarrow_b \text{real})$   
**assumes**  $\bigwedge v :: 'c \Rightarrow_b \text{real}. v \geq 0 \implies f v \geq 0$   
**shows**  $\text{norm } (f n) \leq \text{norm } (f (\text{bfun-pos } n))$   
 ⟨proof⟩

**lemma** *norm-bfun-pos[simp]*:  $\text{norm } (\text{bfun-pos } f) = \text{norm } f$   
 ⟨proof⟩

**lemma** *norm-blinfun-mono-eq-nonneg*:  
**fixes**  $f :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('d \Rightarrow_b \text{real})$   
**assumes**  $\bigwedge v :: 'c \Rightarrow_b \text{real}. v \geq 0 \implies f v \geq 0$   
**shows**  $\text{norm } f = (\bigsqcup v \in \{v. v \geq 0\}. \text{norm } (f v) / \text{norm } v)$   
 ⟨proof⟩

**lemma** *norm-blinfun-normalized-le*:  $\text{norm } (\text{blinfun-apply } f v) / \text{norm } v \leq \text{norm } f$   
 ⟨proof⟩

**lemma** *norm-blinfun-mono-eq-nonneg'*:  
**fixes**  $f :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('d \Rightarrow_b \text{real})$   
**assumes**  $\bigwedge v :: 'c \Rightarrow_b \text{real}. 0 \leq v \implies 0 \leq f v$   
**shows**  $\text{norm } f = (\bigsqcup x \in \{x. \text{norm } x = 1 \wedge x \geq 0\}. \text{norm } (f x))$   
 ⟨proof⟩

**lemma** *norm-blinfun-mono-le-norm-one*:  
**fixes**  $f :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('d \Rightarrow_b \text{real})$   
**assumes**  $\bigwedge v :: 'c \Rightarrow_b \text{real}. v \geq 0 \implies f v \geq 0$   
**assumes**  $\text{norm } x = 1 \implies 0 \leq x$   
**shows**  $\text{norm } (f x) \leq \text{norm } (f 1)$   
 ⟨proof⟩

**lemma** *norm-blinfun-mono-eq-one*:  
**fixes**  $f :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('d \Rightarrow_b \text{real})$   
**assumes**  $\bigwedge v :: 'c \Rightarrow_b \text{real}. v \geq 0 \implies f v \geq 0$   
**shows**  $\text{norm } f = \text{norm } (f 1)$   
 ⟨proof⟩

## 2.6 Miscellaneous

**lemma** *bounded-linear-apply-bfun*: *bounded-linear*  $(\lambda x. \text{apply-bfun } x i)$   
 ⟨proof⟩

**lemma** *lim-blinfun-apply: convergent*  $X \implies (\lambda n. \text{blinfun-apply } (X \ n) \ u) \longrightarrow \text{lim } X \ u$   
 ⟨proof⟩

**lemma** *bounded-apply-blinfun:*  
**assumes** *bounded*  $((F :: 'c \Rightarrow 'd::\text{real-normed-vector} \Rightarrow_L 'b::\text{real-normed-vector})$   
 ‘  $S$ )  
**shows** *bounded*  $((\lambda b. \text{blinfun-apply } (F \ b) \ x) \text{ ‘ } S)$   
 ⟨proof⟩

**lemma** *tendsto-blinfun-apply:*  $(\lambda n. X \ n) \longrightarrow L \implies (\lambda n. \text{blinfun-apply } (X \ n) \ u) \longrightarrow L \ u$   
 ⟨proof⟩

**definition** *nonneg-blinfun*  $(Q :: -::\{\text{ordered-real-normed-vector}\} \Rightarrow_L -::\{\text{ordered-ab-group-add, ordered-real-normed-vector}\}) \equiv (\forall v \geq 0. \text{blinfun-apply } Q \ v \geq 0)$

**definition** *blinfun-le*  $Q \ R = \text{nonneg-blinfun } (R - Q)$

**lemma** *nonneg-blinfun-nonneg[dest]:* *nonneg-blinfun*  $Q \implies 0 \leq v \implies 0 \leq Q \ v$   
 ⟨proof⟩

**lemma** *nonneg-blinfun-mono[dest]:* *nonneg-blinfun*  $Q \implies u \leq v \implies Q \ u \leq Q \ v$   
 ⟨proof⟩

**lemma** *nonneg-id-blinfun:* *nonneg-blinfun* *id-blinfun*  
 ⟨proof⟩

**lemma** *blinfun-nonneg-eq:*  
**assumes**  $\forall v \geq 0. \text{blinfun-apply } (f::('c \Rightarrow_b \text{real}) \Rightarrow_L ('c \Rightarrow_b \text{real})) \ v = \text{blinfun-apply } g \ v$   
**shows**  $f = g$   
 ⟨proof⟩

**lemma** *bfun-zero-le-one:*  $0 \leq (1 :: 'c \Rightarrow_b \text{real})$   
 ⟨proof⟩

**lemma** *norm-nonneg-blinfun-one:*  
**assumes** *nonneg-blinfun*  $(X :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('c \Rightarrow_b \text{real}))$   
**shows**  $\text{norm } X = \text{norm } (\text{blinfun-apply } X \ 1)$



$\langle \text{proof} \rangle$

**lemma** *blinfun-apply-mono*:  $\text{nonneg-blinfun } X \implies 0 \leq v \implies \text{blinfun-le } X \ Y \implies X \ v \leq Y \ v$   
 $\langle \text{proof} \rangle$

**lemma** *nonneg-blinfun-scaleR[intro]*:  $\text{nonneg-blinfun } B \implies 0 \leq c \implies \text{nonneg-blinfun } (c *_{\mathbb{R}} B)$   
 $\langle \text{proof} \rangle$

**lemma** *nonneg-blinfun-compose[intro]*:  $\text{nonneg-blinfun } B \implies \text{nonneg-blinfun } C \implies \text{nonneg-blinfun } (C \circ_L B)$   
 $\langle \text{proof} \rangle$

**lemma** *matrix-le-norm-mono*:  
**assumes**  $\text{nonneg-blinfun } (C :: ('c \Rightarrow_b \text{real}) \Rightarrow_L ('c \Rightarrow_b \text{real}))$   
**and**  $\text{nonneg-blinfun } (D - C)$   
**shows**  $\text{norm } C \leq \text{norm } D$   
 $\langle \text{proof} \rangle$

**lemma** *bounded-subset*:  $Y \subseteq X \implies \text{bounded } (f \text{ ` } X) \implies \text{bounded } (f \text{ ` } Y)$   
 $\langle \text{proof} \rangle$

**lemma** *bounded-subset-range*:  $\text{bounded } (\text{range } f) \implies \text{bounded } (f \text{ ` } Y)$   
 $\langle \text{proof} \rangle$

**lift-definition** *bfun-if* ::  $('b \Rightarrow \text{bool}) \Rightarrow ('b \Rightarrow_b 'c :: \text{metric-space}) \Rightarrow ('b \Rightarrow_b 'c) \Rightarrow ('b \Rightarrow_b 'c)$  **is**  $\lambda b \ u \ v \ s. \text{if } b \ s \text{ then } u \ s \text{ else } v \ s$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-if-add*:  $\text{bfun-if } b \ (w + z) \ (u + v) = \text{bfun-if } b \ w \ u + \text{bfun-if } b \ z \ v$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-if-zero-add*:  $\text{bfun-if } b \ 0 \ (u + v) = \text{bfun-if } b \ 0 \ u + \text{bfun-if } b \ 0 \ v$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-if-zero-le*:  $0 \leq v \implies \text{bfun-if } b \ 0 \ v \leq v$   
 $\langle \text{proof} \rangle$

**lemma** *bfun-if-eq*:  $(\bigwedge i. P \ i \implies \text{apply-bfun } v \ i = \text{apply-bfun } u \ i) \implies (\bigwedge i. \neg P \ i \implies v \ i = \text{apply-bfun } w \ i) \implies \text{bfun-if } P \ u \ w = v$

*<proof>*

**lemma** *bfun-if-scaleR*:  $c *_{\mathbb{R}} \text{bfun-if } b \ v1 \ v2 = \text{bfun-if } b \ (c *_{\mathbb{R}} \ v1) \ (c *_{\mathbb{R}} \ v2)$   
*<proof>*

**lemma** *summable-blinfun-apply*:  
**assumes** *summable*  $(f :: \text{nat} \Rightarrow 'a :: \text{real-normed-vector} \Rightarrow_L 'a)$   
**shows** *summable*  $(\lambda n. f \ n \ v)$   
*<proof>*

**lemma** *blinfun-apply-suminf*:  
**assumes** *summable*  $(f :: \text{nat} \Rightarrow 'a :: \text{real-normed-vector} \Rightarrow_L 'a)$   
**shows**  $(\sum k. \text{blinfun-apply } (f \ k) \ v) = (\sum k. f \ k) \ v$   
*<proof>*  
**end**  
**theory** *MDP-reward-Util*  
**imports** *Blinfun-Util*  
**begin**

## 3 Auxiliary Lemmas

### 3.1 Summability

**lemma** *summable-powser-const*:  
**fixes**  $c :: \text{real}$   
**assumes**  $|c| < 1$   
**shows** *summable*  $(\lambda n. c^{\wedge} n * x)$   
*<proof>*

### 3.2 Infinite sums

**lemma** *suminf-split-head'*:  
*summable*  $(f :: \text{nat} \Rightarrow 'x :: \text{real-normed-vector}) \implies \text{suminf } f = f \ 0$   
 $+ (\sum n. f \ (\text{Suc } n))$   
*<proof>*

**lemma** *sum-disc-lim*:  
**assumes**  $|c :: \text{real}| < 1$   
**shows**  $(\sum x. c^{\wedge} x * B) = B / (1 - c)$   
*<proof>*

### 3.3 Bounded Functions

**lemma** *suminf-apply-bfun*:  
**fixes**  $f :: \text{nat} \Rightarrow 'c \Rightarrow_b \text{real}$

**assumes** *summable f*  
**shows**  $(\sum i. f i) x = (\sum i. f i x)$   
 $\langle proof \rangle$

**lemma** *sum-apply-bfun*:  
**fixes**  $f :: nat \Rightarrow 'c \Rightarrow_b real$   
**shows**  $(\sum i < n. f i) x = (\sum i < n. apply-bfun f i) x$   
 $\langle proof \rangle$

### 3.4 Push-Forward of a Bounded Function

**lemma** *integrable-bfun-prob-space [simp]*:  
*integrable (measure-pmf P) ( $\lambda t. apply-bfun f (F t) :: real$ )*  
 $\langle proof \rangle$

**lift-definition** *push-exp* ::  $('b \Rightarrow 'c pmf) \Rightarrow ('c \Rightarrow_b real) \Rightarrow ('b \Rightarrow_b real)$  **is**  
 $\lambda c f s. measure-pmf.expectation (c s) f$   
 $\langle proof \rangle$

**declare** *push-exp.rep-eq [simp]*

**lemma** *norm-push-exp-le-norm*:  $norm (push-exp d x) \leq norm x$   
 $\langle proof \rangle$

**lemma** *push-exp-bounded-linear [simp]*: *bounded-linear (push-exp d)*  
 $\langle proof \rangle$

**lemma** *onorm-push-exp [simp]*: *onorm (push-exp d) = 1*  
 $\langle proof \rangle$

**lemma** *push-exp-return [simp]*: *push-exp return-pmf = id*  
 $\langle proof \rangle$

### 3.5 Boundedness

**lemma** *bounded-abs [intro]*:  
*bounded (X' :: real set)  $\implies$  bounded (abs ' X')*  
 $\langle proof \rangle$

**lemma** *bounded-abs-range [intro]*:  
*bounded (range f :: real set)  $\implies$  bounded (range ( $\lambda x. abs (f x)$ ))*  
 $\langle proof \rangle$

### 3.6 Probability Theory

**lemma** *integral-measure-pmf-bind*:  
**assumes**  $(\bigwedge x. |f :: 'b \Rightarrow real| x| \leq B)$   
**shows**  $(\int x. f x \partial((measure-pmf M) \gg (\lambda x. measure-pmf (N x))))$   
 $= (\int x. \int y. f y \partial N x \partial M)$

*<proof>*

**lemma** *lemma-4-3-1'*:

**assumes** *set-pmf*  $p \subseteq W$

**and** *bounded*  $((w :: 'c \Rightarrow \text{real}) \text{ ' } W)$

**and**  $W \neq \{\}$

**and** *measure-pmf.expectation*  $p \ w = (\bigsqcup p \in \{p. \text{set-pmf } p \subseteq W\}.$

*measure-pmf.expectation*  $p \ w)$

**shows**  $\exists x \in W. \text{measure-pmf.expectation } p \ w = w \ x$

*<proof>*

**lemma** *lemma-4-3-1*:

**assumes** *set-pmf*  $p \subseteq W$  *integrable*  $(\text{measure-pmf } p) \ w$  *bounded*  $((w :: 'c \Rightarrow \text{real}) \text{ ' } W)$

**shows** *measure-pmf.expectation*  $p \ w \leq \bigsqcup (w \text{ ' } W)$

*<proof>*

**lemma** *bounded-integrable*:

**assumes** *bounded*  $(\text{range } v) \ v \in \text{borel-measurable } (\text{measure-pmf } p)$

**shows** *integrable*  $(\text{measure-pmf } p) \ (v :: 'c \Rightarrow \text{real})$

*<proof>*

### 3.7 Argmax

**lemma** *finite-is-arg-max*: *finite*  $X \Longrightarrow X \neq \{\} \Longrightarrow \exists x. \text{is-arg-max } (f :: 'c \Rightarrow \text{real}) \ (\lambda x. x \in X) \ x$

*<proof>*

**lemma** *finite-arg-max-le*:

**assumes** *finite*  $(X :: 'c \text{ set}) \ X \neq \{\}$

**shows**  $s \in X \Longrightarrow (f :: 'c \Rightarrow \text{real}) \ s \leq f (\text{arg-max-on } (f :: 'c \Rightarrow \text{real}) \ X)$

*<proof>*

**lemma** *arg-max-on-in*:

**assumes** *finite*  $(X :: 'c \text{ set}) \ X \neq \{\}$

**shows**  $(\text{arg-max-on } (f :: 'c \Rightarrow \text{real}) \ X) \in X$

*<proof>*

**lemma** *finite-arg-max-eq-Max*:

**assumes** *finite*  $(X :: 'c \text{ set}) \ X \neq \{\}$

**shows**  $(f :: 'c \Rightarrow \text{real}) \ (\text{arg-max-on } f \ X) = \text{Max } (f \text{ ' } X)$

*<proof>*

**lemma** *arg-max-SUP*: *is-arg-max*  $(f :: 'b \Rightarrow \text{real}) \ (\lambda x. x \in X) \ m \Longrightarrow f \ m = (\bigsqcup (f \text{ ' } X))$

*<proof>*

**definition** *has-max*  $X \equiv \exists x \in X. \forall x' \in X. x' \leq x$

**definition** *has-arg-max*  $f X \equiv \exists x. \text{is-arg-max } f (\lambda x. x \in X) x$

**lemma** *has-max*  $((f :: 'b \Rightarrow \text{real}) ' X) \longleftrightarrow \text{has-arg-max } f X$   
*<proof>*

**lemma** *has-arg-max-is-arg-max*: *has-arg-max*  $f X \Longrightarrow \text{is-arg-max } f$   
 $(\lambda x. x \in X) (\text{arg-max } f (\lambda x. x \in X))$   
*<proof>*

**lemma** *has-arg-max-arg-max*: *has-arg-max*  $f X \Longrightarrow (\text{arg-max } f (\lambda x. x \in X)) \in X$   
*<proof>*

**lemma** *app-arg-max-ge*: *has-arg-max*  $(f :: 'b \Rightarrow \text{real}) X \Longrightarrow x \in X$   
 $\Longrightarrow f x \leq f (\text{arg-max-on } f X)$   
*<proof>*

**lemma** *app-arg-max-eq-SUP*: *has-arg-max*  $(f :: 'b \Rightarrow \text{real}) X \Longrightarrow f$   
 $(\text{arg-max-on } f X) = \bigsqcup (f ' X)$   
*<proof>*

**lemma** *SUP-is-arg-max*:  
**assumes**  $x \in X$  *bdd-above*  $(f ' X) (f :: 'c \Rightarrow \text{real}) x = \bigsqcup (f ' X)$   
**shows** *is-arg-max*  $f (\lambda x. x \in X) x$   
*<proof>*

**lemma** *is-arg-max-linorderI[intro]*: **fixes**  $f :: 'c \Rightarrow 'b :: \text{linorder}$   
**assumes**  $P x \wedge y. (P y \Longrightarrow f x \geq f y)$   
**shows** *is-arg-max*  $f P x$   
*<proof>*

**lemma** *is-arg-max-linorderD[dest]*: **fixes**  $f :: 'c \Rightarrow 'b :: \text{linorder}$   
**assumes** *is-arg-max*  $f P x$   
**shows**  $P x (P y \Longrightarrow f x \geq f y)$   
*<proof>*

**lemma** *is-arg-max-cong*:  
**assumes**  $\bigwedge x. P x \Longrightarrow f x = g x$   
**shows** *is-arg-max*  $f P x \longleftrightarrow \text{is-arg-max } g P x$   
*<proof>*

**lemma** *is-arg-max-cong'*:  
**assumes**  $\bigwedge x. P x \Longrightarrow f x = g x$   
**shows** *is-arg-max*  $f P = \text{is-arg-max } g P$   
*<proof>*

**lemma** *is-arg-max-congI*:

**assumes**  $is\text{-arg-max } f P x \wedge x. P x \implies f x = g x$   
**shows**  $is\text{-arg-max } g P x$   
 $\langle proof \rangle$

### 3.8 Contraction Mappings

**definition**  $is\text{-contraction } C \equiv \exists l. 0 \leq l \wedge l < 1 \wedge (\forall v u. dist (C v) (C u) \leq l * dist v u)$

**lemma**  $banach'$ :

**fixes**  $C :: 'b :: complete\text{-space} \Rightarrow 'b$   
**assumes**  $is\text{-contraction } C$   
**shows**  $\exists! v. C v = v \wedge v. (\lambda n. (C \hat{\sim} n) v) \longrightarrow (THE v. C v = v)$   
 $\langle proof \rangle$

**lemma**  $contraction\text{-dist}$ :

**fixes**  $C :: 'b :: complete\text{-space} \Rightarrow 'b$   
**assumes**  $\wedge v u. dist (C v) (C u) \leq c * dist v u$   
**assumes**  $0 \leq c < 1$   
**shows**  $(1 - c) * dist v (THE v. C v = v) \leq dist v (C v)$   
 $\langle proof \rangle$

### 3.9 Limits

**lemma**  $tendsto\text{-bfun-sandwich}$ :

**assumes**  
 $(f :: nat \Rightarrow 'b \Rightarrow_b real) \longrightarrow x (g :: nat \Rightarrow 'b \Rightarrow_b real) \longrightarrow x$   
 $eventually (\lambda n. f n \leq h n) sequentially eventually (\lambda n. h n \leq g n)$   
 $sequentially$   
**shows**  $(h :: nat \Rightarrow 'b \Rightarrow_b real) \longrightarrow x$   
 $\langle proof \rangle$

### 3.10 Supremum

**lemma**  $SUP\text{-add-le}$ :

**assumes**  $X \neq \{\}$   $bounded (B ' X) bounded (A' ' X)$   
**shows**  $(\bigsqcup c \in X. (B :: 'a \Rightarrow real) c + A' c) \leq (\bigsqcup b \in X. B b) + (\bigsqcup a \in X. A' a)$   
 $\langle proof \rangle$

**lemma**  $le\text{-SUP-diff}'$ :

**assumes**  $ne: X \neq \{\}$   
**and**  $bdd: bounded (B ' X) bounded (A' ' X)$   
**and**  $sup\text{-le}: (\bigsqcup a \in X. (A' :: 'a \Rightarrow real) a) \leq (\bigsqcup b \in X. B b)$   
**shows**  $(\bigsqcup b \in X. B b) - (\bigsqcup a \in X. (A' :: 'a \Rightarrow real) a) \leq (\bigsqcup c \in X. B c - A' c)$   
 $\langle proof \rangle$

**lemma**  $le\text{-SUP-diff}$ :

**fixes**  $A' :: 'a \Rightarrow real$

**assumes**  $X \neq \{\}$  *bounded*  $(B \text{ ' } X)$  *bounded*  $(A' \text{ ' } X)$   $(\bigsqcup a \in X. A' a) \leq (\bigsqcup b \in X. B b)$   
**shows**  $0 \leq (\bigsqcup c \in X. B c - A' c)$   
 $\langle \text{proof} \rangle$

**lemma** *bounded-SUP-mul[simp]*:  
 $X \neq \{\} \implies 0 \leq l \implies \text{bounded } (f \text{ ' } X) \implies (\bigsqcup x \in X. (l :: \text{real}) * f x) = (l * (\bigsqcup x \in X. f x))$   
 $\langle \text{proof} \rangle$

**lemma** *abs-cSUP-le[intro]*:  
 $X \neq \{\} \implies \text{bounded } (F \text{ ' } X) \implies |\bigsqcup x \in X. (F x) :: \text{real}| \leq (\bigsqcup x \in X. |F x|)$   
 $\langle \text{proof} \rangle$

## 4 Least argmax

**definition** *least-arg-max*  $f P = (\text{LEAST } x. \text{is-arg-max } f P x)$

**lemma** *least-arg-max-prop*:  $\exists x :: 'a :: \text{wellorder}. P x \implies \text{finite } \{x. P x\} \implies P (\text{least-arg-max } (f :: - \Rightarrow \text{real}) P)$   
 $\langle \text{proof} \rangle$

**lemma** *is-arg-max-apply-eq*: *is-arg-max*  $(f :: - \Rightarrow - :: \text{linorder}) P x \implies \text{is-arg-max } f P y \implies f x = f y$   
 $\langle \text{proof} \rangle$

**lemma** *least-arg-max-apply*:  
**assumes** *is-arg-max*  $(f :: - \Rightarrow - :: \text{linorder}) P (x :: \text{wellorder})$   
**shows**  $f (\text{least-arg-max } f P) = f x$   
 $\langle \text{proof} \rangle$

**lemma** *apply-arg-max-eq-max*: *finite*  $\{x. P x\} \implies \text{is-arg-max } (f :: - \Rightarrow - :: \text{linorder}) P x \implies f x = \text{Max } (f \text{ ' } \{x. P x\})$   
 $\langle \text{proof} \rangle$

**lemma** *apply-arg-max-eq-max'*: *finite*  $X \implies \text{is-arg-max } (f :: - \Rightarrow - :: \text{linorder}) (\lambda x. x \in X) x \implies (\text{MAX } x \in X. f x) = f x$   
 $\langle \text{proof} \rangle$

**lemma** *least-arg-max-is-arg-max*:  $P \neq \{\} \implies \text{finite } P \implies \text{is-arg-max } f (\lambda x :: \text{wellorder}. x \in P) (\text{least-arg-max } (f :: - \Rightarrow \text{real}) (\lambda x. x \in P))$   
 $\langle \text{proof} \rangle$

**lemma** *is-arg-max-const*: *is-arg-max*  $(f :: - \Rightarrow - :: \text{linorder}) (\lambda y. y = c) x \longleftrightarrow x = c$   
 $\langle \text{proof} \rangle$

**lemma** *least-arg-max-cong'*:

```

assumes  $\bigwedge x. \text{is-arg-max } f P x = \text{is-arg-max } g P x$ 
shows  $\text{least-arg-max } f P = \text{least-arg-max } g P$ 
<proof>

```

end

## 5 Discrete-Time Markov Decision Processes with Arbitrary State Spaces

In this file we construct discrete-time Markov decision processes, e.g. with arbitrary state spaces. Proofs and definitions are adapted from `Markov_Models.Discrete_Time_Markov_Process`.

```

theory MDP-cont
imports HOL-Probability.Probability
begin

```

```

lemma Ionescu-Tulcea-C-eq:
assumes  $\bigwedge i h. h \in \text{space } (PiM \{0..<i\} N) \implies P i h = P' i h$ 
assumes  $h: \text{Ionescu-Tulcea } P N \text{ Ionescu-Tulcea } P' N$ 
shows  $\text{Ionescu-Tulcea.C } P N 0 n (\lambda x. \text{undefined}) = \text{Ionescu-Tulcea.C } P' N 0 n (\lambda x. \text{undefined})$ 
<proof>

```

```

lemma Ionescu-Tulcea-CI-eq:
assumes  $\bigwedge i h. h \in \text{space } (PiM \{0..<i\} N) \implies P i h = P' i h$ 
assumes  $h: \text{Ionescu-Tulcea } P N \text{ Ionescu-Tulcea } P' N$ 
shows  $\text{Ionescu-Tulcea.CI } P N = \text{Ionescu-Tulcea.CI } P' N$ 
<proof>

```

```

lemma measure-eqI-PiM-sequence:
fixes  $M :: \text{nat} \Rightarrow 'a \text{ measure}$ 
assumes  $*[simp]: \text{sets } P = PiM \text{ UNIV } M \text{ sets } Q = PiM \text{ UNIV } M$ 
assumes  $\text{eq}: \bigwedge A n. (\bigwedge i. A i \in \text{sets } (M i)) \implies$ 
 $P (\text{prod-emb UNIV } M \{..n\} (Pi_E \{..n\} A)) = Q (\text{prod-emb UNIV } M \{..n\} (Pi_E \{..n\} A))$ 
assumes  $A: \text{finite-measure } P$ 
shows  $P = Q$ 
<proof>

```

```

lemma distr-cong-simp:
 $M = K \implies \text{sets } N = \text{sets } L \implies (\bigwedge x. x \in \text{space } M = \text{simp} \implies f x = g x) \implies \text{distr } M N f = \text{distr } K L g$ 
<proof>

```

### 5.1 Definition and Basic Properties

```

locale discrete-MDP =

```



**fixes**  $Ms :: 's \text{ measure}$   
**and**  $Ma :: 'a \text{ measure}$   
**and**  $A :: 's \Rightarrow 'a \text{ set}$   
**and**  $K :: 's \times 'a \Rightarrow 's \text{ measure}$

**assumes**  $A\text{-}s: \bigwedge s. A \ s \in \text{sets } Ma$

**assumes**  $A\text{-}ne: \bigwedge s. A \ s \neq \{\}$

**assumes**  $ex\text{-}pol: \exists \delta \in Ms \rightarrow_M Ma. \forall s. \delta \ s \in A \ s$

**assumes**  $K[\text{measurable}]: K \in Ms \otimes_M Ma \rightarrow_M \text{prob-algebra } Ms$   
**begin**

**lemma**  $space\text{-}prodI[\text{intro}]: x \in \text{space } A' \Longrightarrow y \in \text{space } B \Longrightarrow (x,y) \in \text{space } (A' \otimes_M B)$   
 $\langle \text{proof} \rangle$

**abbreviation**  $M \equiv Ms \otimes_M Ma$   
**abbreviation**  $Ma\text{-}A \ s \equiv \text{restrict-space } Ma \ (A \ s)$

**lemma**  $space\text{-}ma[\text{intro}]: s \in \text{space } Ms \Longrightarrow a \in \text{space } Ma \Longrightarrow (s,a) \in \text{space } M$   
 $\langle \text{proof} \rangle$

**lemma**  $space\text{-}x0[\text{simp}]: x0 \in \text{space } (\text{prob-algebra } Ms) \Longrightarrow \text{space } x0 = \text{space } Ms$   
 $\langle \text{proof} \rangle$

**lemma**  $A\text{-}subs\text{-}Ma: A \ s \subseteq \text{space } Ma$   
 $\langle \text{proof} \rangle$

**lemma**  $space\text{-}Ma\text{-}A\text{-}subset: s \in \text{space } Ms \Longrightarrow \text{space } (Ma\text{-}A \ s) \subseteq A \ s$   
 $\langle \text{proof} \rangle$

**lemma**  $K\text{-}restrict [\text{measurable}]: K \in (Ms \otimes_M Ma\text{-}A \ s) \rightarrow_M \text{prob-algebra } Ms$   
 $\langle \text{proof} \rangle$

**lemma**  $measurable\text{-}K\text{-}act[\text{measurable}, \text{intro}]: s \in \text{space } Ms \Longrightarrow (\lambda a. K \ (s, a)) \in Ma \rightarrow_M \text{prob-algebra } Ms$   
 $\langle \text{proof} \rangle$

**lemma**  $measurable\text{-}K\text{-}st[\text{measurable}, \text{intro}]: a \in \text{space } Ma \Longrightarrow (\lambda s. K \ (s, a)) \in Ms \rightarrow_M \text{prob-algebra } Ms$   
 $\langle \text{proof} \rangle$

**lemma**  $space\text{-}K[\text{simp}]: sa \in \text{space } M \Longrightarrow \text{space } (K \ sa) = \text{space } Ms$   
 $\langle \text{proof} \rangle$

**lemma** *space-K2[simp]*:  $s \in \text{space } Ms \implies a \in \text{space } Ma \implies \text{space } (K (s, a)) = \text{space } Ms$   
 ⟨proof⟩

**lemma** *space-K-E*:  $s' \in \text{space } (K (s, a)) \implies s \in \text{space } Ms \implies a \in \text{space } Ma \implies s' \in \text{space } Ms$   
 ⟨proof⟩

**lemma** *sets-K*:  $sa \in \text{space } M \implies \text{sets } (K sa) = \text{sets } Ms$   
 ⟨proof⟩

**lemma** *sets-K'[measurable-cong]*:  $s \in \text{space } Ms \implies a \in \text{space } Ma \implies \text{sets } (K (s, a)) = \text{sets } Ms$   
 ⟨proof⟩

**lemma** *prob-space-K[intro]*:  $sa \in \text{space } M \implies \text{prob-space } (K sa)$   
 ⟨proof⟩

**lemma** *prob-space-K2[intro]*:  $s \in \text{space } Ms \implies a \in \text{space } Ma \implies \text{prob-space } (K (s, a))$   
 ⟨proof⟩

**lemma** *K-in-space [intro]*:  $m \in \text{space } M \implies K m \in \text{space } (\text{prob-algebra } Ms)$   
 ⟨proof⟩

## 5.2 Policies

**type-synonym**  $('c, 'd) \text{ pol} = \text{nat} \Rightarrow ((\text{nat} \Rightarrow 'c \times 'd) \times 'c) \Rightarrow 'd$   
*measure*

**abbreviation**  $H \ i \equiv Pi_M \ \{0..<i\} \ (\lambda-. M)$

**abbreviation**  $Hs \ i \equiv H \ i \ \otimes_M \ Ms$

**lemma** *space-H1*:  $j < (i :: \text{nat}) \implies \omega \in \text{space } (H \ i) \implies \omega \ j \in \text{space } M$   
 ⟨proof⟩

**lemma** *space-case-nat[intro]*:  
**assumes**  $\omega \in \text{space } (H \ i) \ s \in \text{space } Ms$   
**shows**  $\text{case-nat } s \ (\text{fst} \circ \omega) \ i \in \text{space } Ms$   
 ⟨proof⟩

**lemma** *undefined-in-H0*:  $(\lambda-. \text{undefined}) \in \text{space } (H \ (0 :: \text{nat}))$   
 ⟨proof⟩

**lemma** *sets-K-Suc*[*measurable-cong*]:  $h \in \text{space } (H \text{ (Suc } n)) \implies \text{sets } (K \text{ (h } n)) = \text{sets } Ms$   
 ⟨*proof*⟩

A decision rule is a function from states to distributions over enabled actions.

**definition** *is-dec0*  $d \equiv d \in Ms \rightarrow_M \text{prob-algebra } Ma$

**definition** *is-dec* ( $t :: \text{nat}$ )  $d \equiv (d \in Hs \ t \rightarrow_M \text{prob-algebra } Ma)$

**lemma** *is-dec0*  $d \implies \text{is-dec } t \ (\lambda(-,s). \ d \ s)$   
 ⟨*proof*⟩

A policy is a function from histories to valid decision rules.

**definition** *is-policy* ::  $('s, 'a) \text{pol} \Rightarrow \text{bool}$  **where**  
*is-policy*  $p \equiv \forall i. \ \text{is-dec } i \ (p \ i)$

**abbreviation** *p0* ::  $('s, 'a) \text{pol} \Rightarrow 's \Rightarrow 'a \text{ measure}$  **where**  
*p0*  $p \ s \equiv p \ (0 :: \text{nat}) \ (\lambda-. \ \text{undefined}, \ s)$

**context**

**fixes**  $p$  **assumes**  $p[\text{simp}]$ : *is-policy*  $p$

**begin**

**lemma** *is-policyD*[*measurable*]:  $p \ i \in Hs \ i \rightarrow_M \text{prob-algebra } Ma$   
 ⟨*proof*⟩

**lemma** *space-policy*[*simp*]:  $hs \in \text{space } (Hs \ i) \implies \text{space } (p \ i \ hs) = \text{space } Ma$   
 ⟨*proof*⟩

**lemma** *space-policy'*[*simp*]:  $h \in \text{space } (H \ i) \implies s \in \text{space } Ms \implies \text{space } (p \ i \ (h,s)) = \text{space } Ma$   
 ⟨*proof*⟩

**lemma** *space-policyI*[*intro*]:  
**assumes**  $s \in \text{space } Ms \ h \in \text{space } (H \ i) \ a \in \text{space } Ma$   
**shows**  $a \in \text{space } (p \ i \ (h,s))$   
 ⟨*proof*⟩

**lemma** *sets-policy*[*simp*]:  $hs \in \text{space } (Hs \ i) \implies \text{sets } (p \ i \ hs) = \text{sets } Ma$   
 ⟨*proof*⟩

**lemma** *sets-policy'*[*measurable-cong, simp*]:  
 $h \in \text{space } (H \ i) \implies s \in \text{space } Ms \implies \text{sets } (p \ i \ (h,s)) = \text{sets } Ma$   
 ⟨*proof*⟩

**lemma** *sets-policy'*[*measurable-cong, simp*]:  
 $h \in \text{space } ((Pi_M \{ \} (\lambda \cdot M))) \implies s \in \text{space } Ms \implies \text{sets } (p \ 0 \ (h,s))$   
 $= \text{sets } Ma$   
 ⟨*proof*⟩

**lemma** *policy-prob-space*:  $hs \in \text{space } (Hs \ i) \implies \text{prob-space } (p \ i \ hs)$   
 ⟨*proof*⟩

**lemma** *policy-prob-space'*:  $h \in \text{space } (H \ i) \implies s \in \text{space } Ms \implies$   
 $\text{prob-space } (p \ i \ (h,s))$   
 ⟨*proof*⟩

**lemma** *prob-space-p0*:  $x \in \text{space } Ms \implies \text{prob-space } (p0 \ p \ x)$   
 ⟨*proof*⟩

**lemma** *p0-sets*[*measurable-cong*]:  $x \in \text{space } Ms \implies \text{sets } (p \ 0 \ (\lambda \cdot$   
 $\text{undefined},x)) = \text{sets } Ma$   
 ⟨*proof*⟩

**lemma** *space-p0*[*simp*]:  $s \in \text{space } Ms \implies \text{space } (p0 \ p \ s) = \text{space } Ma$   
 ⟨*proof*⟩

**lemma** *return-policy-prob-algebra* [*measurable*]:  
 $h \in \text{space } (H \ n) \implies x \in \text{space } Ms \implies (\lambda a. \text{return } M \ (x, a)) \in p \ n$   
 $(h, x) \rightarrow_M \text{prob-algebra } M$   
 ⟨*proof*⟩  
**end**

### 5.3 Successor Policy

To shift the policy by one step, we provide a single state-action pair as history

**definition** *Suc-policy*  $p \ sa = (\lambda i \ (h, s). p \ (Suc \ i) \ (\lambda i'. \text{case-nat } sa \ h$   
 $i', s))$

**lemma** *p-as-Suc-policy*:  $p \ (Suc \ i) \ (h, s) = \text{Suc-policy } p \ ((h \ 0)) \ i \ (\lambda i.$   
 $h \ (Suc \ i), s)$   
 ⟨*proof*⟩

**lemma** *is-policy-Suc-policy*[*intro*]:  
**assumes**  $s: sa \in \text{space } M$  **and**  $p: \text{is-policy } p$   
**shows**  $\text{is-policy } (Suc\text{-policy } p \ sa)$   
 ⟨*proof*⟩

**lemma** *Suc-policy-measurable-step*[*measurable*]:  
**assumes**  $\text{is-policy } p$   
**shows**  $(\lambda x. \text{Suc-policy } p \ (fst \ (fst \ x)) \ n \ (snd \ (fst \ x), \ snd \ x)) \in$   
 $(M \ \otimes_M \ Pi_M \ \{0..<n\} \ (\lambda \cdot M)) \ \otimes_M \ Ms \rightarrow_M \text{prob-algebra } Ma$   
 ⟨*proof*⟩

## 5.4 Single-Step Distribution

$K'$  takes a policy, a distribution over  $s$ , the epoch, and a history, produces a distribution over the next state-action pair.

**definition**  $K' :: ('s, 'a) \text{ pol} \Rightarrow 's \text{ measure} \Rightarrow \text{nat} \Rightarrow (\text{nat} \Rightarrow ('s \times 'a)) \Rightarrow ('s \times 'a) \text{ measure}$

**where**

```

 $K' p s0 n \omega = \text{do} \{$ 
 $s \leftarrow \text{case-nat } s0 (K \circ \omega) n;$ 
 $a \leftarrow p n (\omega, s);$ 
 $\text{return } M (s, a)$ 

```

}

**lemma** *prob-space- $K'$* :

**assumes**  $p$ : *is-policy*  $p$  **and**  $x$ :  $x0 \in \text{space (prob-algebra } Ms)$  **and**  $h$ :  $h \in \text{space (H } n)$

**shows** *prob-space*  $(K' p x0 n h)$

*<proof>*

**lemma** *measurable- $K'$ [measurable]*:

**assumes**  $p$ : *is-policy*  $p$  **and**  $x$ :  $x \in \text{space (prob-algebra } Ms)$

**shows**  $K' p x i \in H i \rightarrow_M \text{prob-algebra } M$

*<proof>*

## 5.5 Initial State-Action Distribution

$K0$  produces the initial state-action distribution from a state distribution and a policy.

**definition**  $K0 p s0 = K' p s0 0 (\lambda-. \text{undefined})$

**lemma**  *$K0$ -def'*:

```

 $K0 p s0 = \text{do} \{$ 
 $s \leftarrow s0;$ 
 $a \leftarrow p0 p s;$ 
 $\text{return } M (s, a)\}$ 

```

*<proof>*

**lemma**  *$K0$ -prob[measurable]*: *is-policy*  $p \implies K0 p \in \text{prob-algebra } Ms$

$\rightarrow_M \text{prob-algebra } M$

*<proof>*

**lemma** *prob-space- $K0$* : *is-policy*  $p \implies x0 \in \text{space (prob-algebra } Ms)$

$\implies \text{prob-space (} K0 p x0)$

*<proof>*

**lemma** *space- $K0$ [simp]*: *is-policy*  $p \implies s \in \text{space (prob-algebra } Ms)$

$\implies \text{space (} K0 p s) = \text{space } M$

*<proof>*

**lemma** *sets-K0[measurable-cong]*:  
**assumes** *is-policy*  $p$   $s \in \text{space}$  (*prob-algebra*  $M$   $s$ )  
**shows** *sets* ( $K0$   $p$   $s$ ) = *sets*  $M$   
 $\langle \text{proof} \rangle$

**lemma** *K0-return-eq-p0*:  
**assumes** *is-policy*  $p$   $s \in \text{space}$   $M$   $s$   
**shows**  $K0$   $p$  (*return*  $M$   $s$ ) =  $p0$   $p$   $s$   $\gg$  ( $\lambda a.$  *return*  $M$  ( $s, a$ ))  
 $\langle \text{proof} \rangle$

**lemma** *M-ne-policy[intro]*: *is-policy*  $p \implies s \in \text{space}$  (*prob-algebra*  $M$   $s$ )  
 $\implies \text{space}$   $M \neq \{\}$   
 $\langle \text{proof} \rangle$

## 5.6 Sequence Space of the MDP

We can instantiate *Ionescu-Tulcea* with  $K'$ .

**lemma** *IT-K'*: *is-policy*  $p \implies x \in \text{space}$  (*prob-algebra*  $M$   $s$ )  $\implies$  *Ionescu-Tulcea*  
 $(K' p x)$   $(\lambda-. M)$   
 $\langle \text{proof} \rangle$

**definition** *lim-sequence* ::  $(s, a)$  *pol*  $\Rightarrow$   $s$  *measure*  $\Rightarrow$  ( $\text{nat} \Rightarrow (s \times a)$ ) *measure*

**where**

*lim-sequence*  $p x = \text{projective-family.lim UNIV}$  (*Ionescu-Tulcea.CI*  
 $(K' p x)$   $(\lambda-. M)$ )  $(\lambda-. M)$

**lemma**

**assumes**  $x: x \in \text{space}$  (*prob-algebra*  $M$   $s$ ) **and**  $p: \text{is-policy}$   $p$   
**shows** *space-lim-sequence*: *space* (*lim-sequence*  $p x$ ) = *space*  $(\prod_{M$   
 $i \in \text{UNIV}. M)$

**and** *sets-lim-sequence[measurable-cong]*: *sets* (*lim-sequence*  $p x$ ) =  
*sets*  $(\prod_{M} i \in \text{UNIV}. M)$

**and** *emeasure-lim-sequence-emb*:  $\bigwedge J X.$  *finite*  $J \implies X \in \text{sets}$   $(\prod_{M$   
 $j \in J. M) \implies$

*emeasure* (*lim-sequence*  $p x$ ) (*prod-emb UNIV*  $(\lambda-. M)$   $J X$ ) =  
*emeasure* (*Ionescu-Tulcea.CI*  $(K' p x)$   $(\lambda-. M)$   $J$ )  $X$

**and** *emeasure-lim-sequence-emb-I0o*:  $\bigwedge n X.$   $X \in \text{sets}$   $(\prod_{M} i \in$   
 $\{0..<n\}. M) \implies$

*emeasure* (*lim-sequence*  $p x$ ) (*prod-emb UNIV*  $(\lambda-. M)$   $\{0..<n\}$   
 $X$ ) =

*emeasure* (*Ionescu-Tulcea.C*  $(K' p x)$   $(\lambda-. M)$   $0 n$   $(\lambda x.$  *undefined*))

$X$

$\langle \text{proof} \rangle$

**lemma** *lim-sequence-prob-space*:

**assumes** *is-policy*  $p$   $s \in \text{space}$  (*prob-algebra*  $M$   $s$ )

**shows** *prob-space* (*lim-sequence*  $p s$ )

⟨proof⟩

## 5.7 Measurability of the Sequence Space

**lemma** *lim-sequence[measurable]*:

**assumes** *p*: *is-policy p*

**shows** *lim-sequence p*  $\in$  *prob-algebra Ms*  $\rightarrow_M$  *prob-algebra* ( $\Pi_M$   
 $i \in UNIV. M$ )

⟨proof⟩

**lemma** *lim-sequence-aux[measurable]*:

**assumes** *p*: *is-policy p*

**assumes** *f*:  $\bigwedge x. x \in \text{space } M \implies \text{is-policy } (f x)$

**assumes** *f'*:  $\bigwedge n. (\lambda x. f (\text{fst } (\text{fst } x)) n (\text{snd } (\text{fst } x), \text{snd } x)) \in$

$(M \otimes_M Pi_M \{0..<n\} (\lambda-. M)) \otimes_M Ms \rightarrow_M \text{prob-algebra } Ma$

**assumes** *gm*:  $g \in M \rightarrow_M \text{prob-algebra } Ms$

**shows**  $(\lambda x. \text{lim-sequence } (f x) (g x)) \in M \rightarrow_M \text{prob-algebra } (Pi_M$   
 $UNIV (\lambda-. M))$

⟨proof⟩

**lemma** *lim-sequence-Suc-return[measurable]*:

**assumes** *p*: *is-policy p*

**assumes** *s*:  $s \in \text{space } Ms$

**shows**  $(\lambda x. \text{lim-sequence } (\text{Suc-policy } p (s, \text{snd } x)) (\text{return } Ms (\text{fst}$   
 $x))) \in$

$M \rightarrow_M \text{prob-algebra } (Pi_M UNIV (\lambda-. M))$

⟨proof⟩

**lemma** *lim-sequence-Suc-K[measurable]*:

**assumes** *is-policy p*

**shows**  $(\lambda x. \text{lim-sequence } (\text{Suc-policy } p x) (K x)) \in M \rightarrow_M \text{prob-algebra}$   
 $(Pi_M UNIV (\lambda-. M))$

⟨proof⟩

## 5.8 Iteration Rule

**lemma** *step-C*:

**assumes** *x*:  $x \in \text{space } (\text{prob-algebra } Ms)$  **and** *p*: *is-policy p*

**shows** *Ionescu-Tulcea.C* ( $K' p x$ )  $(\lambda-. M) 0 1 (\lambda-. \text{undefined}) \gg=$

*Ionescu-Tulcea.C* ( $K' p x$ )  $(\lambda-. M) 1 n =$

$K 0 p x \gg= (\lambda a. \text{Ionescu-Tulcea.C } (K' p x) (\lambda-. M) 1 n (\text{case-nat}$   
 $a (\lambda-. \text{undefined})))$

⟨proof⟩

**lemma** *lim-sequence-eq*:

**assumes** *x*:  $x \in \text{space } (\text{prob-algebra } Ms)$  **assumes** *p*: *is-policy p*

**shows** *lim-sequence p x* =

$K 0 p x \gg= (\lambda y. \text{distr } (\text{lim-sequence } (\text{Suc-policy } p y) (K y)) (\Pi_M$   
 $- \in UNIV. M) (\text{case-nat } y))$

$(\text{is } - = ?B p x)$

⟨proof⟩

## 5.9 Stream Space of the MDP

**definition**  $\text{lim-stream} :: ('s, 'a) \text{pol} \Rightarrow 's \text{ measure} \Rightarrow ('s \times 'a) \text{ stream measure}$

**where**

$\text{lim-stream } p \ x = \text{distr } (\text{lim-sequence } p \ x) \ (\text{stream-space } M) \ \text{to-stream}$

**lemma**  $\text{space-lim-stream}$ :  $\text{space } (\text{lim-stream } p \ x) = \text{streams } (\text{space } M)$   
⟨proof⟩

**lemma**  $\text{sets-lim-stream[measurable-cong]}$ :  $\text{sets } (\text{lim-stream } p \ x) = \text{sets } (\text{stream-space } M)$   
⟨proof⟩

**lemma**  $\text{lim-stream[measurable]}$ :

**assumes**  $\text{is-policy } p$

**shows**  $\text{lim-stream } p \in \text{prob-algebra } Ms \rightarrow_M \text{prob-algebra } (\text{stream-space } M)$

⟨proof⟩

**lemma**  $\text{lim-stream-Suc[measurable]}$ :

**assumes**  $p$ :  $\text{is-policy } p$

**shows**  $(\lambda a. \text{lim-stream } (\text{Suc-policy } p \ a) \ (K \ a)) \in M \rightarrow_M \text{prob-algebra } (\text{stream-space } M)$

⟨proof⟩

**lemma**  $\text{space-stream-space-M-ne}$ :  $x \in \text{space } M \implies \text{space } (\text{stream-space } M) \neq \{\}$

⟨proof⟩

**lemma**  $\text{prob-space-lim-stream[intro]}$ :

**assumes**  $\text{is-policy } p \ x \in \text{space } (\text{prob-algebra } Ms)$

**shows**  $\text{prob-space } (\text{lim-stream } p \ x)$

⟨proof⟩

**lemma**  $\text{prob-space-step}$ :

**assumes**  $\text{is-policy } p \ x \in \text{space } M$

**shows**  $\text{prob-space } (\text{lim-stream } (\text{Suc-policy } p \ x) \ (K \ x))$

⟨proof⟩

**lemma**  $\text{lim-stream-eq}$ :

**assumes**  $p$ :  $\text{is-policy } p$

**assumes**  $x$ :  $x \in \text{space } (\text{prob-algebra } Ms)$

**shows**  $\text{lim-stream } p \ x = \text{do } \{$

$y \leftarrow K0 \ p \ x;$

$\omega \leftarrow \text{lim-stream } (\text{Suc-policy } p \ y) \ (K \ y);$



```

    return (stream-space M) (y ##  $\omega$ )
  }
  <proof>

end
end

```

```

theory MDP-disc
  imports
    MDP-cont
    HOL-Library.Omega-Words-Fun
begin

```

## 6 Markov Decision Processes with Discrete State Spaces

```

lemma (in prob-space) integral-stream-space:
  fixes f :: 'a stream  $\Rightarrow$  ('b :: {banach, second-countable-topology, real-normed-vector})
  assumes int-f: integrable (stream-space M) f
  assumes [measurable]: f  $\in$  borel-measurable (stream-space M)
  shows ( $\int X. f X \partial$ stream-space M) = ( $\int x. (\int X. f (x ## X) \partial$ stream-space M)  $\partial$ M)
  <proof>

```

```

lemma prefix-cons:
  Omega-Words-Fun.prefix (Suc n) seq = seq 0 # Omega-Words-Fun.prefix
  n ( $\lambda n. seq (Suc n)$ )
  <proof>

```

```

lemma restrict-Suc: restrict y {0..<Suc i} (Suc n) = (restrict ( $\lambda n. y$ 
  (Suc n)) {0..<i}) n
  <proof>

```

```

lemma prefix-restrict: Omega-Words-Fun.prefix i (restrict y {0..<i})
  = Omega-Words-Fun.prefix i y
  <proof>

```

```

lemma prefix-measurable[measurable]:
  Omega-Words-Fun.prefix i  $\in$  PiM {0..<i}
  ( $\lambda. count-space (UNIV :: ('s :: countable  $\times$  'a :: countable) set))  $\rightarrow_M$ 
  count-space UNIV
  <proof>$ 
```

```

no-notation Omega-Words-Fun.build (infixr <##> 65)

```

```

locale discrete-MDP =
  fixes A :: 's :: countable  $\Rightarrow$  'a :: countable set — enabled actions

```

**and**  $K :: 's \times 'a \Rightarrow 's \text{ pmf}$  — MDP kernel, transition probabilities  
**assumes**  
 $A\text{-ne}: \bigwedge s. A\ s \neq \{\}$  — set of enabled actions is nonempty  
**begin**

## 6.1 Policies

Type synonym for decision rules.

**type-synonym**  $('c, 'd) \text{ dec} = 'c \Rightarrow 'd \text{ pmf}$

**definition**  $\text{is-dec} :: ('s, 'a) \text{ dec} \Rightarrow \text{bool}$  **where**  
 $\text{is-dec } d \equiv \forall s. d\ s \subseteq A\ s$

**lemma**  $\text{is-decI}$ [intro]:  
 $(\bigwedge s. \text{set-pmf } (d\ s) \subseteq A\ s) \Longrightarrow \text{is-dec } d$   
 $\langle \text{proof} \rangle$

**abbreviation**  $D_R \equiv \{d. \text{is-dec } d\}$

**definition**  $\text{is-dec-det} :: ('s \Rightarrow 'a) \Rightarrow \text{bool}$  **where**  
 $\text{is-dec-det } d \equiv \forall s. d\ s \in A\ s$

**abbreviation**  $D_D \equiv \{d. \text{is-dec-det } d\}$

**definition**  $\text{mk-dec-det } d\ s = \text{return-pmf } (d\ s)$

**lemma**  $\text{is-dec-mk-dec-det-iff}$  [simp]:  $\text{is-dec } (\text{mk-dec-det } d) \longleftrightarrow \text{is-dec-det } d$   
 $\langle \text{proof} \rangle$

**lemma**  $D\text{-det-to-MR}$ [intro]:  $\text{is-dec-det } d \Longrightarrow \text{is-dec } (\text{mk-dec-det } d)$   
 $\langle \text{proof} \rangle$

Due to the assumption  $A\ ?s \neq \{\}$ , a deterministic decision rule always exists. It immediately follows via  $\text{is-dec } (\text{mk-dec-det } ?d) = \text{is-dec-det } ?d$  that a randomized decision rule also exists.

**lemma**  $\text{SOME-is-dec-det}$ :  $\text{is-dec-det } (\lambda s. \text{SOME } a. a \in A\ s)$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{ex-dec-det}$  [simp]:  $\exists d. \text{is-dec-det } d$   
 $\langle \text{proof} \rangle$

**lemma**  $D\text{-det-ne}$  [simp]:  $D_D \neq \{\}$   
 $\langle \text{proof} \rangle$

**lemma**  $D_R\text{-ne}$  [simp]:  $D_R \neq \{\}$   
 $\langle \text{proof} \rangle$

**lemma** *ex-dec*[*intro*, *simp*]:  $\exists d. \text{is-dec } d$   
 $\langle \text{proof} \rangle$

Type synonym for policies.

**type-synonym**  $('c, 'd) \text{ pol} = ('c \times 'd) \text{ list} \Rightarrow ('c, 'd) \text{ dec}$

A policy assigns a decision rule to each observed past.

**definition** *is-policy* ::  $('s, 'a) \text{ pol} \Rightarrow \text{bool}$  **where**  
*is-policy*  $p \equiv \forall hs. \text{is-dec } (p \text{ } hs)$

**abbreviation**  $\Pi_{HR} \equiv \{p. \text{is-policy } p\}$

Deterministic policies

**definition** *is-deterministic*  $p \equiv \text{is-policy } p \wedge (\forall h \ s. \exists a. p \ h \ s = \text{return-pmf } a)$

**definition** *mk-det*  $p \ h \ s \equiv \text{return-pmf } (p \ h \ s)$

**abbreviation**  $\Pi_{HD} \equiv \{p. \forall h. p \ h \in D_D\}$

Markovian policies

**definition** *is-markovian*  $p \equiv \text{is-policy } p \wedge (\forall h \ h'. \text{length } h = \text{length } h' \longrightarrow p \ h = p \ h')$

**definition** *mk-markovian* ::  $(\text{nat} \Rightarrow ('s, 'a) \text{ dec}) \Rightarrow ('s, 'a) \text{ pol}$  **where**  
*mk-markovian*  $p \equiv (\lambda h. p \ (\text{length } h))$

**lemma** *is-markovian-mk-iff*[*simp*]:  $\text{is-markovian } (mk\text{-markovian } p) \longleftrightarrow (\forall n. \text{is-dec } (p \ n))$   
 $\langle \text{proof} \rangle$

**lemma** *is-markovian-mk*[*intro*]:  $\forall n. \text{is-dec } (p \ n) \Longrightarrow \text{is-markovian } (mk\text{-markovian } p)$   
 $\langle \text{proof} \rangle$

**lemma** *mk-markovian-nil* [*simp*]:  $mk\text{-markovian } p \ [] = p \ 0$   
 $\langle \text{proof} \rangle$

**definition** *mk-markovian-det*  $p \equiv (\lambda h \ s. \text{return-pmf } (p \ (\text{length } h) \ s))$

**abbreviation**  $\Pi_{MD} \equiv \{p. \forall n::\text{nat}. p \ n \in D_D\}$

**abbreviation**  $\Pi_{MR} \equiv \{p. \forall n. p \ n \in D_R\}$

**lemma**  $\Pi_{MR}\text{-imp-policies}$ [*intro*]:  $p \in \Pi_{MR} \Longrightarrow \text{mk-markovian } p \in \Pi_{HR}$   
 $\langle \text{proof} \rangle$

**lemma**  $\Pi_{MD}\text{-MR-iff}$ [*simp*]:  $(\lambda n. mk\text{-dec-det } (p \ n)) \in \Pi_{MR} \longleftrightarrow p \in \Pi_{MD}$

$\langle proof \rangle$

**lemma**  $\Pi_{MD-to-MR}[intro]: p \in \Pi_{MD} \implies (\lambda n. mk-dec-det (p n)) \in \Pi_{MR}$   
 $\langle proof \rangle$

**lemma**  $p-n-\pi-MD[intro]: p \in \Pi_{MD} \implies p n \in D_D$   
 $\langle proof \rangle$

**lemma**  $p-n-\pi-MR[intro]: p \in \Pi_{MR} \implies p n \in D_R$   
 $\langle proof \rangle$

**lemma**  $\Pi_{MD-ne}[simp]: \Pi_{MD} \neq \{\}$   
 $\langle proof \rangle$

**lemma**  $\Pi_{MR-ne}[simp]: \Pi_{MR} \neq \{\}$   
 $\langle proof \rangle$

**lemma**  $policies-ne[simp, intro]: \Pi_{HR} \neq \{\}$   
 $\langle proof \rangle$

Stationary policies

**definition**  $is-stationary p \equiv is-policy p \wedge (\forall h h'. p h = p h')$

**lemma**  $is-stationary-const-iff[simp]: is-stationary (\lambda-. d) = is-dec d$   
 $\langle proof \rangle$

**lemma**  $is-stationary-const[intro]: is-dec d \implies is-stationary (\lambda-. d)$   
 $\langle proof \rangle$

**abbreviation**  $mk-stationary p \equiv mk-markovian (\lambda-. p)$

**abbreviation**  $mk-stationary-det d \equiv mk-markovian (\lambda-. mk-dec-det d)$

### 6.1.1 Successor Policy

After taking the first step in the MDP, we will know which state and which action got selected during the initial epoch. To obtain a policy that acts as if the current epoch was the initial one, we prepend the observed state-action pair to the history. The result is again a policy, i.e. it satisfies *is-policy*.

**definition**  $\pi-Suc :: ('s, 'a) pol \Rightarrow 's \times 'a \Rightarrow ('s, 'a) pol$

**where**

$\pi-Suc p sa h = p (sa\#h)$

**lemma**  $is-policy-\pi-Suc [intro]: is-policy p \implies is-policy (\pi-Suc p sa)$   
 $\langle proof \rangle$

**lemma** *Suc-mk-markovian[simp]*:  $\pi$ -*Suc* (*mk-markovian*  $p$ )  $x = \text{mk-markovian}$   
 $(\lambda n. p (Suc\ n))$   
 $\langle \text{proof} \rangle$

## 6.2 Stream Space of the MDP

### 6.2.1 Initial State-Action Distribution

If we fix a decision rule  $d$  and an initial distribution of states  $S0$ , we obtain a distribution over state-action pairs in the following way: First, the initial state  $s$  is sampled from  $S0$ , then an action  $a$  is selected from  $d\ s$ .

**definition** *K0*  $d\ S0 = do\ \{$   
 $s \leftarrow S0;$   
 $a \leftarrow d\ s;$   
 $return\text{-}pmf\ (s,a)$   
 $\}$

**notation** *K0* ( $\langle K_0 \rangle$ )

**lemma** *K0-iff*:  $K0\ d\ S0 = S0 \gg= (\lambda s. map\text{-}pmf\ (\lambda a. (s,a))\ (d\ s))$   
 $\langle \text{proof} \rangle$

**lemma** *image-pair[simp]*:  $Pair\ x - \{p\} = (if\ x = fst\ p\ then\ \{snd\ p\}$   
 $else\ \{\})$   
 $\langle \text{proof} \rangle$

**lemma** *pmf-K0 [simp]*:  $pmf\ (K0\ d\ S0)\ (s,a) = pmf\ S0\ s * pmf\ (d\ s)$   
 $a$   
 $\langle \text{proof} \rangle$

**lemma** *set-pmf-K0*:  $set\text{-}pmf\ (K0\ p\ S0) = \{(s,a). s \in S0 \wedge a \in p\ s\}$   
 $\langle \text{proof} \rangle$

**lemma** *fst-K0[simp]*:  $map\text{-}pmf\ fst\ (K0\ p\ S0) = S0$   
 $\langle \text{proof} \rangle$

**abbreviation**  $S \equiv stream\text{-}space\ (count\text{-}space\ UNIV)$

We inherit the trace space from MDPs with continuous state-action spaces

**interpretation** *MDP-cont*:  $MDP\text{-}cont.\text{discrete}\text{-}MDP\ count\text{-}space\ UNIV$   
 $count\text{-}space\ UNIV\ A\ K$   
 $\langle \text{proof} \rangle$

**lemma** *count-space-M[simp]*:  $MDP\text{-}cont.M = count\text{-}space\ UNIV$   
 $\langle \text{proof} \rangle$

**lemma** *space-M[simp]*: *space MDP-cont.M = UNIV*  
 ⟨*proof*⟩

We reuse the stream space provided by *MDP-cont.lim-stream*

**definition**  $T :: ('s, 'a) \text{ pol} \Rightarrow 's \text{ pmf} \Rightarrow ('s \times 'a) \text{ stream measure}$   
**where**  $T p = \text{MDP-cont.lim-stream } (\lambda n (h, s). p (\text{Omega-Words-Fun.prefix } n h) s)$

**lemma** *sets-T[measurable-cong]*:  
 $\text{sets } (T p x) = \text{sets } S$   
 ⟨*proof*⟩

**lemma** *space-stream-space-ne[simp]*: *space S ≠ {}*  
 ⟨*proof*⟩

**lemma** *space-T[simp]*: *space (T p S0) = space S*  
 ⟨*proof*⟩

**lemma** *is-policy-MDP-cont[intro]*:  
**fixes**  $p :: ('s \times 'a) \text{ list} \Rightarrow 's \Rightarrow 'a \text{ pmf}$   
**shows** *MDP-cont.is-policy*  $(\lambda n (h, s). p (\text{Omega-Words-Fun.prefix } n h) s)$   
 ⟨*proof*⟩

**lemma** *prob-space-T[intro, simp]*: *prob-space (T p x)*  
 ⟨*proof*⟩

**lemma** *T-subprob[simp]*:  
 $T p S0 \in \text{space } (\text{subprob-algebra } S)$   
 ⟨*proof*⟩

**lemma** *T-subprob-space [simp]*: *subprob-space (T p S0)*  
 ⟨*proof*⟩

**lemma** *K0-MDP-cont-eq*:  
 $\text{MDP-cont.K0 } (\lambda x (h, s). \text{measure-pmf } (p (\text{Omega-Words-Fun.prefix } x h) s)) (\text{measure-pmf } S0) =$   
 $\text{K0 } (p []) S0$   
 ⟨*proof*⟩

## 6.2.2 Decomposition of the Stream Space

The distribution of traces/walks the MDP allows should intuitively satisfy the following rule:

1. select the initial state  $s$  from  $S0$
2. pass it to the decision rule  $p []$  to determine a distribution over actions

3. select the action  $a$

- finally pass the state-action pair  $(s, a)$  to the kernel  $K$  to get a new distribution over states  $s_0'$

Then the iteration repeats with the updated policy  $\pi\text{-Suc } p (s, a)$ .

The result carries over from  $\llbracket \text{MDP-cont.is-policy } ?p; ?x \in \text{space (prob-algebra (count-space UNIV))} \rrbracket \implies \text{MDP-cont.lim-stream } ?p \text{ } ?x = \text{MDP-cont.K0 } ?p \text{ } ?x \ggg (\lambda y. \text{MDP-cont.lim-stream } (\text{MDP-cont.Suc-policy } ?p \text{ } y) (\text{measure-pmf } (K \text{ } y)) \ggg (\lambda \omega. \text{return (stream-space MDP-cont.M) (y \#\#\ \omega)}))$ .

**lemma** *T-eq*:

**shows**  $T \text{ } p \text{ } S0 = \text{do } \{$   
 $\text{ } sa \leftarrow \text{measure-pmf } (K0 \text{ } (p \text{ []}) \text{ } S0);$   
 $\text{ } \omega \leftarrow T (\pi\text{-Suc } p \text{ } sa) (K \text{ } sa);$   
 $\text{ } \text{return } S (sa \text{ \#\#\ } \omega)$   
 $\}$   
 $\langle \text{proof} \rangle$

**lemma** *T-eq-distr*:

**shows**  $T \text{ } p \text{ } S0 = \text{measure-pmf } (K0 \text{ } (p \text{ []}) \text{ } S0) \ggg (\lambda sa. \text{distr } (T (\pi\text{-Suc } p \text{ } sa) (K \text{ } sa)) \text{ } S ((\#\#\ \text{ } sa)))$   
 $\langle \text{proof} \rangle$

The iteration rule lets us nicely decompose integrals (expected values) over functions on traces of the MDP.

**lemma** *integral-T*:

**fixes**  $f :: ('s \times 'a) \text{ stream} \Rightarrow \text{real}$   
**assumes**  $f\text{-bounded: } \bigwedge x. |f \text{ } x| \leq B$   
**assumes**  $f: f \in \text{borel-measurable } S$   
**shows**  $(\int t. f \text{ } t \text{ } \partial T \text{ } p \text{ } x) = \int sa. \int t'. f (sa \text{ \#\#\ } t') \text{ } \partial T (\pi\text{-Suc } p \text{ } sa) (K \text{ } sa) \text{ } \partial K0 \text{ } (p \text{ []}) \text{ } x$   
 $\langle \text{proof} \rangle$

**lemma** *nn-integral-T*:

**assumes**  $f: f \in \text{borel-measurable } S$   
**shows**  $(\int ^+ t. f \text{ } t \text{ } \partial T \text{ } p \text{ } x) = (\int ^+ sa. \int ^+ t'. f (sa \text{ \#\#\ } t') \text{ } \partial T (\pi\text{-Suc } p \text{ } sa) (K \text{ } sa) \text{ } \partial K0 \text{ } (p \text{ []}) \text{ } x)$   
 $\langle \text{proof} \rangle$

### 6.2.3 A Denotational View on the Stochastic Process

Many definitions on MDPs do not rely on the individual traces but only on the distribution of states and actions at each epoch.

We define this view on the trace space as the repeated iteration of  $K_0$  and  $K$ . It coincides with the definition of  $T$ .

**primrec**  $Pn :: ('s, 'a) \text{pol} \Rightarrow 's \text{ pmf} \Rightarrow \text{nat} \Rightarrow ('s \times 'a) \text{ pmf}$  **where**  
 $Pn \ p \ S0 \ 0 = K0 \ (p \ []) \ S0$   
 $| Pn \ p \ S0 \ (Suc \ n) = K0 \ (p \ []) \ S0 \ggg (\lambda sa. Pn \ (\pi\text{-Suc} \ p \ sa) \ (K \ sa) \ n)$

**declare**  $Pn.\text{simps}(2)[\text{simp del}]$

**lemma**  $Pn\text{-eq-}T$ :  $\text{measure-pmf} \ (Pn \ p \ S0 \ n) = \text{distr} \ (T \ p \ S0) \ (\text{count-space} \ UNIV) \ (\lambda t. t \ !! \ n)$   
 $\langle \text{proof} \rangle$

The definition of  $Pn$  also allows us to easily prove that only enabled actions can occur in the traces of the MDP.

**lemma**  $Pn\text{-in-}A$ :  $\text{is-policy} \ p \Longrightarrow (s, a) \in Pn \ p \ S0 \ n \Longrightarrow a \in A \ s$   
 $\langle \text{proof} \rangle$

**lemma**  $T\text{-in-}A$ :  
**assumes**  $\text{is-policy} \ p$   
**shows**  $AE \ t \ \text{in} \ T \ p \ S0. \ \text{snd} \ (t \ !! \ n) \in A \ (\text{fst} \ (t \ !! \ n))$   
 $\langle \text{proof} \rangle$

#### 6.2.4 State Process

Alongside  $Pn$ , we also define the state and action distributions as projections.

**definition**  $Xn \ p \ S0 \ n = \text{map-pmf} \ \text{fst} \ (Pn \ p \ S0 \ n)$

**lemma**  $X0$   $[\text{simp}]$ :  $Xn \ p \ S0 \ 0 = S0$   
 $\langle \text{proof} \rangle$

**lemma**  $Xn\text{-Suc}$ :  $Xn \ p \ S0 \ (Suc \ n) = Pn \ p \ S0 \ n \ggg K$   
 $\langle \text{proof} \rangle$

**lemma**  $Pn\text{-markovian-eq-}Xn\text{-bind}$ :  $Pn \ (\text{mk-markovian} \ p) \ S0 \ n = K0 \ (p \ n) \ (Xn \ (\text{mk-markovian} \ p) \ S0 \ n)$   
 $\langle \text{proof} \rangle$

**lemma**  $Xn\text{-Suc}'$ :  $Xn \ p \ S0 \ (Suc \ n) = K0 \ (p \ []) \ S0 \ggg (\lambda sa. Xn \ (\pi\text{-Suc} \ p \ sa) \ (K \ sa) \ n)$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{set-pmf-}X0$   $[\text{simp}]$ :  $\text{set-pmf} \ (Xn \ p \ S0 \ 0) = S0$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{set-pmf-}PSuc$ :  $\text{set-pmf} \ (Pn \ (\text{mk-markovian} \ p) \ S0 \ n) = \{(s, a). s \in \text{set-pmf} \ (Xn \ (\text{mk-markovian} \ p) \ S0 \ n) \wedge a \in p \ n \ s\}$   
 $\langle \text{proof} \rangle$



### 6.2.5 The Conditional Distribution of Actions

Actions are selected wrt. the whole history of state-action pairs encountered so far. The following definition defines the expected action selection when only the current state is given.

**definition**  $Y\text{-cond-}X\ p\ S0\ n\ x = \text{map-pmf}\ \text{snd}\ (\text{cond-pmf}\ (Pn\ p\ S0\ n)\ \{(s,a).\ s = x\})$

**lemma**  $\text{prob-}K0\text{-}X\ [\text{simp}]: \text{measure-pmf.prob}\ (K0\ p\ S0)\ \{(s, a).\ s = x\} = \text{pmf}\ S0\ x$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{prob-}Pn\text{-}X[\text{simp}]: \text{measure-pmf.prob}\ (Pn\ p\ S0\ n)\ \{(s, a).\ s = x\} = \text{pmf}\ (Xn\ p\ S0\ n)\ x$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{pmf-}Pn\text{-pair}$ :  
**assumes**  $sa \in \text{set-pmf}\ (Pn\ p\ S0\ n)$   
**shows**  $\text{pmf}\ (Pn\ p\ S0\ n)\ sa = \text{pmf}\ (Y\text{-cond-}X\ p\ S0\ n\ (\text{fst}\ sa))\ (\text{snd}\ sa) * \text{pmf}\ (Xn\ p\ S0\ n)\ (\text{fst}\ sa)$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{pmf-}Pn$ :  
**assumes**  $x \in \text{set-pmf}\ (Xn\ p\ S0\ n)$   
**shows**  $\text{pmf}\ (Pn\ p\ S0\ n)\ (x,a) = \text{pmf}\ (Y\text{-cond-}X\ p\ S0\ n\ x)\ a * \text{pmf}\ (Xn\ p\ S0\ n)\ x$   
 $\langle \text{proof} \rangle$

**lemma**  $\text{pmf-}Y\text{-cond-}X$ :  
**assumes**  $x \in \text{set-pmf}\ (Xn\ p\ S0\ n)$   
**shows**  $\text{pmf}\ (Y\text{-cond-}X\ p\ S0\ n\ x)\ a = \text{pmf}\ (Pn\ p\ S0\ n)\ (x,a) / \text{pmf}\ (Xn\ p\ S0\ n)\ x$   
 $\langle \text{proof} \rangle$

**lemma**  $Y\text{-cond-}X\ 0[\text{simp}]$ :  
**assumes**  $x \in \text{set-pmf}\ S0$   
**shows**  $Y\text{-cond-}X\ p\ S0\ 0\ x = p\ \square\ x$   
 $\langle \text{proof} \rangle$

**lemma**  $Y\text{-cond-}X\ \text{markovian}[\text{simp}]$ :  
**assumes**  $h: x \in Xn\ (\text{mk-markovian}\ p)\ S0\ n$   
**shows**  $Y\text{-cond-}X\ (\text{mk-markovian}\ p)\ S0\ n\ x = p\ n\ x$   
 $\langle \text{proof} \rangle$

**lemma**  $Pn\ \text{eq}\ Xn\ \text{Y-cond}$ :  $Pn\ p\ S0\ n = Xn\ p\ S0\ n \gg (\lambda x. \text{map-pmf}\ (\lambda a. (x, a))\ (Y\text{-cond-}X\ p\ S0\ n\ x))$   
 $\langle \text{proof} \rangle$

**lemma** *Pn-eq-Xn-Y-cond'*:

$Pn\ p\ S0\ n = Xn\ p\ S0\ n \gg= (\lambda s. Y\text{-cond-}X\ p\ S0\ n\ s \gg= (\lambda a. \text{return-pmf}\ (s,a)))$   
 ⟨proof⟩

**lemma** *Pn-markovian-Suc*:  $Pn\ (mk\text{-markovian}\ p)\ S0\ (Suc\ n) =$

$Pn\ (mk\text{-markovian}\ p)\ S0\ n \gg= (\lambda sa. K0\ (p\ (Suc\ n))\ (K\ sa))$   
 ⟨proof⟩

### 6.2.6 Action Process

The distribution of actions.

**definition**  $Yn\ p\ S0\ n = \text{map-pmf}\ \text{snd}\ (Pn\ p\ S0\ n)$

**lemma** *Y0*:  $Yn\ p\ S0\ 0 = S0 \gg= p\ []$

⟨proof⟩

For markovian policies, the decision rules at each epoch are independent of each other, hence we may express  $Yn$  solely in terms of  $Xn$  and the current decision rule.

**lemma** *Yn-markovian*:  $Yn\ (mk\text{-markovian}\ p)\ S0\ n = Xn\ (mk\text{-markovian}\ p)\ S0\ n \gg= p\ n$

⟨proof⟩

### 6.3 Restriction to Markovian Policies

**abbreviation**  $as\text{-markovian}\ p\ S0\ n\ x \equiv$

$\text{if } x \in (Xn\ p\ S0\ n) \text{ then } Y\text{-cond-}X\ p\ S0\ n\ x \text{ else } \text{return-pmf}\ (SOME\ a. a \in A\ x)$

For states which cannot occur we choose an arbitrary enabled action, as in this case we cannot make any statements about  $Y\text{-cond-}X$  (a distribution conditioned on an event with probability 0).

**lemma** *is- $\Pi_{MR}$ -as-markovian*:

**assumes**  $p$ : *is-policy*  $p$

**shows** *as-markovian*  $p\ S0 \in \Pi_{MR}$

⟨proof⟩

**lemma** *is-policy-as-markovian*:  $\text{is-policy}\ p \implies \text{is-policy}\ (mk\text{-markovian}\ (as\text{-markovian}\ p\ S0))$

⟨proof⟩

**theorem** *Pn-as-markovian-eq*:  $Pn\ (mk\text{-markovian}\ (as\text{-markovian}\ p\ S0))\ S0 = Pn\ p\ S0$

⟨proof⟩

## 6.4 MDPs without Initial Distribution

From now on, we assume a known, deterministic initial state. All results from the previous discussion carry over as we are now in the special case where the initial state is of the form *return-pmf s*.

**definition**  $\mathcal{T} p s \equiv T p (\text{return-pmf } s)$

**lemma**  *$\mathcal{T}$ -eq-return-distr*:  $\mathcal{T} p s =$   
 $\text{measure-pmf } (p \square s) \gg (\lambda a. \text{distr } (T (\pi\text{-Suc } p (s,a))) (K (s,a))) S$   
 $((\#\#) (s,a))$   
 $\langle \text{proof} \rangle$

**lemma**  *$\mathcal{T}$ -eq-return*:  
**shows**  $\mathcal{T} p s = \text{do } \{$   
 $y \leftarrow \text{measure-pmf } (p \square s);$   
 $\omega \leftarrow T (\pi\text{-Suc } p (s,y)) (K (s,y));$   
 $\text{return } S ((s,y) \#\# \omega)$   
 $\}$   
 $\langle \text{proof} \rangle$

**lemma**  *$\mathcal{T}$ -return*:  
**shows**  $T p S0 = \text{measure-pmf } S0 \gg \mathcal{T} p$   
 $\langle \text{proof} \rangle$

**lemma**  *$\mathcal{T}$ -return-eq*:  
 $\mathcal{T} p s = \text{do } \{$   
 $a \leftarrow \text{measure-pmf } (p \square s);$   
 $s' \leftarrow \text{measure-pmf } (K (s,a));$   
 $w \leftarrow T (\pi\text{-Suc } p (s,a)) (\text{return-pmf } s');$   
 $\text{return } S ((s,a) \#\# w)$   
 $\}$   
 $\langle \text{proof} \rangle$

**lemma**  *$\mathcal{T}$ -eq*:  
**shows**  $\mathcal{T} p s = \text{do } \{$   
 $a \leftarrow \text{measure-pmf } (p \square s);$   
 $s' \leftarrow \text{measure-pmf } (K (s,a));$   
 $w \leftarrow \mathcal{T} (\pi\text{-Suc } p (s,a)) s';$   
 $\text{return } S ((s,a) \#\# w)$   
 $\}$   
 $\langle \text{proof} \rangle$

**lemma**  *$\mathcal{T}$ -prob-space[intro]*: *prob-space*  $(\mathcal{T} p s)$   
 $\langle \text{proof} \rangle$

**lemma**  *$\mathcal{T}$ -sets[measurable-cong]*:  
 $\text{sets } (\mathcal{T} p s) = \text{sets } S$   
 $\langle \text{proof} \rangle$

**lemma** *measurable-ident-Suc'*[*measurable*]:

$(\lambda x. x) \in \mathcal{T} (\pi\text{-Suc } p \text{ } sa) \text{ } s' \rightarrow_M S$   
 $\langle \text{proof} \rangle$

**lemma** *nn-integral-T*:

**fixes**  $f :: ('s \times 'a) \text{ stream} \Rightarrow \text{real}$   
**assumes**  $f[\text{measurable}]$ :  $f \in \text{borel-measurable } S$   
**shows**  $(\int^{+t}. f \text{ } t \text{ } \partial \mathcal{T} \text{ } p \text{ } s)$   
 $= \int^{+a}. \int^{+s'}. \int^{+t'}. f ((s,a)\#\#t') \text{ } \partial \mathcal{T} (\pi\text{-Suc } p (s,a)) \text{ } s' \text{ } \partial K (s,a)$   
 $\partial p \text{ } \square \text{ } s$   
 $\langle \text{proof} \rangle$

**lemma** *integral-T*:

**fixes**  $f :: ('s \times 'a) \text{ stream} \Rightarrow \text{real}$   
**assumes**  $f\text{-bounded}$ :  $\bigwedge x. |f x| \leq B$   
**assumes**  $f[\text{measurable}]$ :  $f \in \text{borel-measurable } S$   
**shows**  $(\int t. f \text{ } t \text{ } \partial \mathcal{T} \text{ } p \text{ } s)$   
 $= \int a. \int s'. \int t'. f ((s,a)\#\#t') \text{ } \partial \mathcal{T} (\pi\text{-Suc } p (s,a)) \text{ } s' \text{ } \partial K (s,a) \text{ } \partial p$   
 $\square \text{ } s$   
 $\langle \text{proof} \rangle$

**lemma** *integrable-T-bounded*[*intro*]:

**fixes**  $f :: ('s \times 'a) \text{ stream} \Rightarrow 'd :: \{\text{second-countable-topology, banach}\}$   
**assumes**  $f[\text{measurable}]$ :  $f \in \text{borel-measurable } S$   
**assumes**  $b$ : *bounded* (*range*  $f$ )  
**shows** *integrable* ( $\mathcal{T} \text{ } p \text{ } s$ )  $f$   
 $\langle \text{proof} \rangle$

**definition**  $Pn' \text{ } p \text{ } s = Pn \text{ } p \text{ } (\text{return-pmf } s)$

**definition**  $Xn' \text{ } p \text{ } s = Xn \text{ } p \text{ } (\text{return-pmf } s)$

**definition**  $Yn' \text{ } p \text{ } s = Yn \text{ } p \text{ } (\text{return-pmf } s)$

**definition**  $K0' \text{ } d \text{ } s \equiv \text{map-pmf } (\lambda a. (s, a)) \text{ } (d \text{ } s)$

**definition**  $K\text{-st } d \text{ } s \equiv d \text{ } s \gg (\lambda a. K (s, a))$

**lemma** *pmf-K-st*:  $\text{pmf } (K\text{-st } d \text{ } s) \text{ } t = \int a. \text{pmf } (K(s, a)) \text{ } t \text{ } \partial d \text{ } s$   
 $\langle \text{proof} \rangle$

$K\text{-st}$  defines the distribution over the successor states for a given decision rule and state. It is mostly useful for markovian policies, as the information which action was selected is lost.

**lemma**  $P0'[\text{simp}]$ :  $Pn' \text{ } p \text{ } s \text{ } 0 = K0' (p \text{ } \square) \text{ } s$   
 $\langle \text{proof} \rangle$

**lemma**  $X0'[\text{simp}]$ :  $Xn' \text{ } p \text{ } s \text{ } 0 = \text{return-pmf } s$   
 $\langle \text{proof} \rangle$

**lemma**  $Pn\text{-return-pmf}$ :  $S0 \gg (\lambda s'. Pn \text{ } p \text{ } (\text{return-pmf } s') \text{ } n) = Pn \text{ } p$

$S0\ n$   
 $\langle proof \rangle$

**lemma**  $PSuc'$ :  $Pn'\ p\ s\ (Suc\ n) = K0'\ (p\ [])\ s \ggg (\lambda sa. K\ sa \ggg (\lambda s'. Pn'\ (\pi\text{-}Suc\ p\ sa)\ s'\ n))$   
 $\langle proof \rangle$

**lemma**  $PSuc'$ -markovian:  
 $Pn'\ (mk\text{-}markovian\ p)\ s\ (Suc\ n) = K\text{-}st\ (p\ 0)\ s \ggg (\lambda s'. Pn'\ (mk\text{-}markovian\ (p\ \circ\ Suc))\ s'\ n)$   
 $\langle proof \rangle$

**lemma**  $Xn'$ - $Suc$ :  $Xn'\ p\ s\ (Suc\ n) = Pn'\ p\ s\ n \ggg K$   
 $\langle proof \rangle$

**lemma**  $Xn'$ - $Pn'$ :  $Xn'\ p\ s\ n = map\text{-}pmf\ fst\ (Pn'\ p\ s\ n)$   
 $\langle proof \rangle$

**lemma**  $Suc$ - $Xn'$ :  $Xn'\ p\ s\ (Suc\ n) = p\ []\ s \ggg (\lambda a. K\ (s,a) \ggg (\lambda s'. Xn'\ (\pi\text{-}Suc\ p\ (s,a))\ s'\ n))$   
 $\langle proof \rangle$

**lemma**  $Suc$ - $Xn'$ -markovian:  
 $Xn'\ (mk\text{-}markovian\ p)\ s\ (Suc\ n) = K\text{-}st\ (p\ 0)\ s \ggg (\lambda s'. Xn'\ (mk\text{-}markovian\ (\lambda n. p\ (Suc\ n)))\ s'\ n)$   
 $\langle proof \rangle$

**lemma**  $Xn'$ -split:  $Xn'\ (mk\text{-}markovian\ p)\ s\ (n + m) = Xn'\ (mk\text{-}markovian\ p)\ s\ n \ggg (\lambda s. Xn'\ (mk\text{-}markovian\ (\lambda i. p\ (i + n)))\ s\ m)$   
 $\langle proof \rangle$

**lemma**  $Yn'$ -markovian:  $Yn'\ (mk\text{-}markovian\ p)\ s\ n = Xn'\ (mk\text{-}markovian\ p)\ s\ n \ggg p\ n$   
 $\langle proof \rangle$

**lemma**  $Pn'$ -markovian-eq- $Xn'$ -bind:  $Pn'\ (mk\text{-}markovian\ p)\ s\ n = Xn'\ (mk\text{-}markovian\ p)\ s\ n \ggg K0'\ (p\ n)$   
 $\langle proof \rangle$

**lemma**  $Pn'$ -eq- $\mathcal{T}$ :  $measure\text{-}pmf\ (Pn'\ p\ s\ n) = distr\ (\mathcal{T}\ p\ s)\ (count\text{-}space\ UNIV)\ (\lambda t. t\ !!\ n)$   
 $\langle proof \rangle$

**end**  
**end**

**theory**  $MDP\text{-}reward$

```

imports
  Bounded-Functions
  MDP-reward-Util
  Blinfun-Util
  MDP-disc
begin

```

## 7 Markov Decision Processes with Rewards

```

locale MDP-reward = discrete-MDP A K
for
  A and
  K :: 's :: countable × 'a :: countable ⇒ 's pmf +
fixes
  r :: ('s × 'a) ⇒ real and
  l :: real
assumes
  zero-le-disc [simp]: 0 ≤ l and
  r-bounded: bounded (range r)
begin

```

This extension to the basic MDPs is formalized with another locale. It assumes the existence of a reward function  $r$  which takes a state-action pair to a real number. We assume that the function is bounded  $r$ -bounded.

Furthermore, we fix a discounting factor  $l$ , where  $0 \leq l \wedge l < 1$ .

### 7.1 Util

#### 7.1.1 Basic Properties of rewards

```

lemma r-bfun: r ∈ bfun
  ⟨proof⟩

```

```

lemma r-bounded': bounded (r ' X)
  ⟨proof⟩

```

```

definition r_M = (⊔ sa. |r sa|)

```

```

lemma abs-r-le-r_M: |r sa| ≤ r_M
  ⟨proof⟩

```

```

lemma abs-r_M-eq-r_M [simp]: |r_M| = r_M
  ⟨proof⟩

```

```

lemma r_M-nonneg: 0 ≤ r_M
  ⟨proof⟩

```

**lemma** *measurable-r-nth* [*measurable*]:  $(\lambda t. r (t !! i)) \in \text{borel-measurable } S$

*<proof>*

**lemma** *integrable-r-nth* [*simp*]: *integrable*  $(\mathcal{T} p s) (\lambda t. r (t !! i))$

*<proof>*

**lemma** *expectation-abs-r-le*: *measure-pmf.expectation*  $d (\lambda a. |r (s, a)|) \leq r_M$

*<proof>*

**lemma** *abs-exp-r-le*:  $| \text{measure-pmf.expectation } d r | \leq r_M$

*<proof>*

### 7.1.2 Infinite discounted sums

**lemma** *abs-disc-eq*[*simp*]:  $|l \hat{ } i * x| = l \hat{ } i * |x|$

*<proof>*

**lemma** *norm-l-pow-eq*[*simp*]:  $\text{norm } (l \hat{ } t *_{R} F) = l \hat{ } t * \text{norm } F$

*<proof>*

## 7.2 Total Reward for Single Traces

**abbreviation**  *$\nu$ -trace-fin*  $t N \equiv \sum_{i < N}. l \hat{ } i * r (t !! i)$

**abbreviation**  *$\nu$ -trace*  $t \equiv \sum_{i}. l \hat{ } i * r (t !! i)$

**lemma** *abs- $\nu$ -trace-fin-le*:  $|\nu\text{-trace-fin } t N| \leq (\sum_{i < N}. l \hat{ } i * r_M)$

*<proof>*

**lemma** *measurable-suminf-reward*[*measurable*]:  *$\nu$ -trace*  $\in \text{borel-measurable } S$

*<proof>*

**lemma** *integrable- $\nu$ -trace-fin*: *integrable*  $(\mathcal{T} p s) (\lambda t. \nu\text{-trace-fin } t N)$

*<proof>*

**context**

**fixes**  $p :: ('s, 'a) \text{pol}$

**begin**

## 7.3 Expected Finite-Horizon Discounted Reward

**definition**  *$\nu$ -fin*  $n s = \int t. \nu\text{-trace-fin } t n \partial \mathcal{T} p s$

**lemma** *abs- $\nu$ -fin-le*:  $|\nu\text{-fin } N s| \leq (\sum_{i < N}. l \hat{ } i * r_M)$

*<proof>*

**lemma**  *$\nu$ -fin-bfun*:  $(\lambda s. \nu\text{-fin } N s) \in \text{bfun}$

$\langle proof \rangle$

**lift-definition**  $\nu_b\text{-fin} :: \text{nat} \Rightarrow 's \Rightarrow_b \text{real}$  is  $\nu\text{-fin}$   
 $\langle proof \rangle$

**lemma**  $\nu\text{-fin}\text{-Suc}[simp]: \nu\text{-fin} (\text{Suc } n) s = \nu\text{-fin } n s + l \hat{\wedge} n * \int t. r$   
 $(t !! n) \partial \mathcal{T} p s$   
 $\langle proof \rangle$

**lemma**  $\nu\text{-fin}\text{-zero}[simp]: \nu\text{-fin } 0 s = 0$   
 $\langle proof \rangle$

**lemma**  $\nu\text{-fin}\text{-eq}\text{-Pn}: \nu\text{-fin } n s = (\sum i < n. l \hat{\wedge} i * \text{measure}\text{-pmf}.\text{expectation}$   
 $(Pn' p s i) r)$   
 $\langle proof \rangle$   
**end**

## 7.4 Expected Total Discounted Reward

**definition**  $\nu p s = \text{lim} (\lambda n. \nu\text{-fin } p n s)$

**lemmas**  $\nu\text{-eq}\text{-lim} = \nu\text{-def}$

**lemma**  $\nu\text{-eq}\text{-Pn}: \nu p s = (\sum i. l \hat{\wedge} i * \text{measure}\text{-pmf}.\text{expectation} (Pn' p$   
 $s i) r)$   
 $\langle proof \rangle$

## 7.5 Reward of a Decision Rule

**context**

**fixes**  $d :: ('s, 'a) \text{dec}$

**begin**

**abbreviation**  $r\text{-dec } s \equiv \int a. r (s, a) \partial d s$

**lemma**  $\text{abs}\text{-}r\text{-dec}\text{-le}: |r\text{-dec } s| \leq r_M$   
 $\langle proof \rangle$

**lemma**  $r\text{-dec}\text{-eq}\text{-}r\text{-}K0: r\text{-dec } s = \text{measure}\text{-pmf}.\text{expectation} (K0' d s) r$   
 $\langle proof \rangle$

**lemma**  $r\text{-dec}\text{-bfun}: r\text{-dec} \in \text{bfun}$   
 $\langle proof \rangle$

**lift-definition**  $r\text{-dec}_b :: 's \Rightarrow_b \text{real}$  is  $r\text{-dec}$   
 $\langle proof \rangle$

**declare**  $r\text{-dec}_b.\text{rep}\text{-eq}[simp] \text{ bfun}.\text{Bfun}\text{-inverse}[simp]$

**lemma**  $\text{norm}\text{-}r\text{-dec}\text{-le}: \text{norm } r\text{-dec}_b \leq r_M$   
 $\langle proof \rangle$



**end**

**lemma** *r-dec-det* [*simp*]:  $r\text{-dec } (mk\text{-dec-det } d) s = r (s, d s)$   
*<proof>*

## 7.6 Transition Probability Matrix for MDPs

**context**

**fixes**  $p :: nat \Rightarrow ('s, 'a) dec$

**begin**

**definition**  $\mathcal{P}_X n = push\text{-exp } (\lambda s. Xn' (mk\text{-markovian } p) s n)$

**lemma**  $\mathcal{P}_X\text{-}0$ [*simp*]:  $\mathcal{P}_X 0 = id$   
*<proof>*

**lemma**  $\mathcal{P}_X\text{-bounded-linear}$ [*simp*]: *bounded-linear* ( $\mathcal{P}_X t$ )  
*<proof>*

**lemma** *norm- $\mathcal{P}_X$*  [*simp*]: *onorm* ( $\mathcal{P}_X t$ ) = 1  
*<proof>*

**lemma** *norm- $\mathcal{P}_X$ -apply*[*simp*]: *norm* ( $\mathcal{P}_X n x$ )  $\leq norm x$   
*<proof>*

**lemma**  $\mathcal{P}_X\text{-bound-r}$ : *norm* ( $\mathcal{P}_X t (r\text{-dec}_b (p t))$ )  $\leq r_M$   
*<proof>*

**lemma**  $\mathcal{P}_X\text{-bounded-r}$ : *bounded* (*range* ( $\lambda t. (\mathcal{P}_X t (r\text{-dec}_b (p t)))$ ))  
*<proof>*

**end**

**lemma**  $\nu\text{-fin-elm}$ :  $\nu\text{-fin } (mk\text{-markovian } p) n s = (\sum i < n. l\hat{i} * \mathcal{P}_X p i (r\text{-dec}_b (p i)) s)$   
*<proof>*

**lemma**  $\nu_b\text{-fin-eq-}\mathcal{P}_X$ :  $\nu_b\text{-fin } (mk\text{-markovian } p) n = (\sum i < n. l\hat{i} *_R \mathcal{P}_X p i (r\text{-dec}_b (p i)))$   
*<proof>*

**lemma**  $\nu\text{-fin-eq-}\mathcal{P}_X$ :  $\nu\text{-fin } (mk\text{-markovian } p) n = (\sum i < n. l\hat{i} *_R \mathcal{P}_X p i (r\text{-dec}_b (p i)))$   
*<proof>*

$\mathcal{P}_1 d v$  defines for each state the expected value of  $v$  after taking a single step in the MDP according to the decision rule  $d$ .

**context**

**fixes**  $d :: ('s, 'a) dec$

**begin**

**lift-definition**  $\mathcal{P}_1 :: ('s \Rightarrow_b \text{real}) \Rightarrow_L ('s \Rightarrow_b \text{real})$  is push-exp (K-st d)

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1\text{-bfun-one}$  [simp]:  $\mathcal{P}_1 1 = 1$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1\text{-pow-bfun-one}$  [simp]:  $(\mathcal{P}_1 \text{~~} t) 1 = 1$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1\text{-pow}$ :  $\text{blinfun-apply } (\mathcal{P}_1 \text{~~} n) = \text{blinfun-apply } \mathcal{P}_1 \text{~~} n$

$\langle \text{proof} \rangle$

**lemma**  $\text{norm-}\mathcal{P}_1$  [simp]:  $\text{norm } \mathcal{P}_1 = 1$

$\langle \text{proof} \rangle$

**end**

**lemma**  $\mathcal{P}_X\text{-Suc}$ :  $\mathcal{P}_X p (\text{Suc } n) v = \mathcal{P}_1 (p 0) ((\mathcal{P}_X (\lambda n. p (\text{Suc } n)) n) v)$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_X\text{-Suc}'$ :  $\mathcal{P}_X p (\text{Suc } n) v = \mathcal{P}_X p n (\mathcal{P}_1 (p n) v)$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_X\text{-const}$ :  $\mathcal{P}_X (\lambda-. d) n = \mathcal{P}_1 d \text{~~} n$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_X\text{-sconst}$ :  $\mathcal{P}_X (\lambda-. p) n = \mathcal{P}_1 p \text{~~} n$

$\langle \text{proof} \rangle$

**lemma**  $\text{norm-}\mathcal{P}\text{-}n$ [simp]:  $\text{onorm } (\mathcal{P}_1 d \text{~~} n) = 1$

$\langle \text{proof} \rangle$

**lemma**  $\text{norm-}\mathcal{P}_1\text{-pow}$  [simp]:  $\text{norm } (\mathcal{P}_1 d \text{~~} t) = 1$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_X\text{-Suc-n-elem}$ :  $\mathcal{P}_X p n (\mathcal{P}_1 (p n) v) = \mathcal{P}_X p (\text{Suc } n) v$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1\text{-eq-}\mathcal{P}_X\text{-one}$ :  $\text{blinfun-apply } (\mathcal{P}_1 (p 0)) = \mathcal{P}_X p 1$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1\text{-pos}$ :  $0 \leq u \implies 0 \leq \mathcal{P}_1 d u$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1\text{-nonneg}$ :  $\text{nonneg-blinfun } (\mathcal{P}_1 d)$

$\langle \text{proof} \rangle$

**lemma**  $\mathcal{P}_1$ -*n-pos*:  $0 \leq u \implies 0 \leq (\mathcal{P}_1 d \hat{\sim} n) u$   
 ⟨proof⟩

**lemma**  $\mathcal{P}_1$ -*n-nonneg*: *nonneg-blinfun*  $(\mathcal{P}_1 d \hat{\sim} n)$   
 ⟨proof⟩

**lemma**  $\mathcal{P}_1$ -*n-disc-pos*:  $0 \leq u \implies 0 \leq (l \hat{\sim} n *_R \mathcal{P}_1 d \hat{\sim} n) u$   
 ⟨proof⟩

**lemma**  $\mathcal{P}_1$ -*sum-pos*:  $0 \leq u \implies 0 \leq (\sum t \leq n. l \hat{\sim} t *_R (\mathcal{P}_1 d \hat{\sim} t)) u$   
 ⟨proof⟩

**lemma**  $\mathcal{P}_1$ -*sum-ge*:  
**assumes**  $0 \leq u$   
**shows**  $u \leq (\sum t \leq n. l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t) u$   
 ⟨proof⟩

## 7.7 The Bellman Operator

**definition**  $L d v \equiv r\text{-dec}_b d + l *_R \mathcal{P}_1 d v$

**lemma** *norm-L-le*: *norm*  $(L d v) \leq r_M + l * \text{norm } v$   
 ⟨proof⟩

**lemma** *abs-L-le*:  $|L d v s| \leq r_M + l * \text{norm } v$   
 ⟨proof⟩

### 7.7.1 Bellman Operator for Single Actions

**abbreviation**  $L_a a v s \equiv r (s, a) + l * \text{measure-pmf.expectation } (K (s, a)) v$

**lemma**  $L_a$ -*le*:  
**fixes**  $v :: 's \Rightarrow_b \text{real}$   
**shows**  $|L_a a v s| \leq r_M + l * \text{norm } v$   
 ⟨proof⟩

**lemma**  $L_a$ -*bounded*:  
*bounded*  $(\text{range } (\lambda a. L_a a (\text{apply-bfun } v) s))$   
 ⟨proof⟩

**lemma**  $L_a$ -*int*:  
**fixes**  $d :: 'a \text{ pmf}$  **and**  $v :: 's \Rightarrow_b \text{real}$   
**shows**  $(\int a. L_a a v s \partial d) = (\int a. r (s, a) \partial d) + l * \int a. \int s'. v s'$   
 $\partial K (s, a) \partial d$   
 ⟨proof⟩

**lemma**  $L$ -*eq- $L_a$* :  $L d v s = \text{measure-pmf.expectation } (d s) (\lambda a. L_a a v s)$   
 ⟨proof⟩

**lemma**  $L$ -eq- $L_a$ -det:  $L (mk\text{-}dec\text{-}det\ d) v\ s = L_a (d\ s) v\ s$   
 $\langle proof \rangle$

**lemma**  $L_a$ -eq- $L$ :  $measure\text{-}pmf.\text{expectation}\ p (\lambda a. L_a\ a\ (apply\text{-}bfun\ v)\ s) =$   
 $L (\lambda t. \text{if } t = s \text{ then } p \text{ else } return\text{-}pmf\ (SOME\ a. a \in A\ t)) v\ s$   
 $\langle proof \rangle$

**lemma**  $L$ -le:  $L\ d\ v\ s \leq r_M + l * norm\ v$   
 $\langle proof \rangle$

**lemma**  $L_a$ -le':  $L_a\ a\ (apply\text{-}bfun\ v)\ s \leq r_M + l * norm\ v$   
 $\langle proof \rangle$

## 7.8 Optimality Equations

**definition**  $\mathcal{L} (v :: 's \Rightarrow_b\ real) s = (\bigsqcup d \in D_R. L\ d\ v\ s)$

**lemma**  $\mathcal{L}$ -bfun:  $\mathcal{L}\ v \in bfun$   
 $\langle proof \rangle$

**lift-definition**  $\mathcal{L}_b :: ('s \Rightarrow_b\ real) \Rightarrow 's \Rightarrow_b\ real$  is  $\mathcal{L}$   
 $\langle proof \rangle$

**lemma**  $L$ -bounded[simp, intro]:  $bounded\ (range\ (\lambda p. L\ p\ v\ s))$   
 $\langle proof \rangle$

**lemma**  $L$ -bounded'[simp, intro]:  $bounded\ ((\lambda p. L\ p\ v\ s) \text{ ` } X)$   
 $\langle proof \rangle$

**lemma**  $L$ -bdd-above[simp, intro]:  $bdd\text{-}above\ ((\lambda p. L\ p\ v\ s) \text{ ` } X)$   
 $\langle proof \rangle$

**lemma**  $L$ -le- $\mathcal{L}_b$ :  $is\text{-}dec\ d \implies L\ d\ v \leq \mathcal{L}_b\ v$   
 $\langle proof \rangle$

### 7.8.1 Equivalences involving $\mathcal{L}_b$

**lemma**  $SUP$ -step- $MR$ -eq:  
 $\mathcal{L}\ v\ s = (\bigsqcup pa \in \{pa. set\text{-}pmf\ pa \subseteq A\ s\}. (\int a. L_a\ a\ v\ s\ \partial measure\text{-}pmf\ pa))$   
 $\langle proof \rangle$

**lemma**  $\mathcal{L}_b$ -eq- $SUP$ - $L_a$ :  $\mathcal{L}_b\ v\ s = (\bigsqcup p \in \{p. set\text{-}pmf\ p \subseteq A\ s\}. \int a. L_a\ a\ v\ s\ \partial measure\text{-}pmf\ p)$   
 $\langle proof \rangle$

**lemma**  $SUP$ -step-det-eq:  $(\bigsqcup d \in D_D. L (mk\text{-}dec\text{-}det\ d) v\ s) = (\bigsqcup a \in A\ s. L_a\ a\ v\ s)$

$\langle proof \rangle$

**lemma** *integrable- $L_a$* : *integrable (measure-pmf x) ( $\lambda a. L_a a$  (apply-bfun v) s)*

$\langle proof \rangle$

**lemma** *SUP- $L_a$ -eq-det*:

**fixes**  $v :: 's \Rightarrow_b \text{real}$

**shows**  $(\bigsqcup p \in \{p. \text{set-pmf } p \subseteq A \text{ s}\}. \int a. L_a a v s \partial \text{measure-pmf } p) = (\bigsqcup a \in A \text{ s}. L_a a v s)$

$\langle proof \rangle$

**lemma**  *$\mathcal{L}$ -eq-SUP-det*:  $\mathcal{L} v s = (\bigsqcup d \in D_D. L (\text{mk-dec-det } d) v s)$

$\langle proof \rangle$

**lemma**  *$\mathcal{L}_b$ -eq-SUP-det*:  $\mathcal{L}_b v s = (\bigsqcup d \in D_D. L (\text{mk-dec-det } d) v s)$

$\langle proof \rangle$

## 7.9 Monotonicity

**lemma**  *$\mathcal{P}_X$ -mono[*intro*]*:  $a \leq b \implies \mathcal{P}_X p n a \leq \mathcal{P}_X p n b$

$\langle proof \rangle$

**lemma**  *$\mathcal{P}_1$ -mono[*intro*]*:  $a \leq b \implies \mathcal{P}_1 p a \leq \mathcal{P}_1 p b$

$\langle proof \rangle$

**lemma**  *$L$ -mono[*intro*]*:  $u \leq v \implies L d u \leq L d v$

$\langle proof \rangle$

**lemma**  *$\mathcal{L}_b$ -mono[*intro*]*:  $u \leq v \implies \mathcal{L}_b u \leq \mathcal{L}_b v$

$\langle proof \rangle$

**lemma** *step-mono*:

**assumes**  $\mathcal{L}_b v \leq v d \in D_R$

**shows**  $L d v \leq v$

$\langle proof \rangle$

**lemma** *step-mono-elem-det*:

**assumes**  $v \leq \mathcal{L}_b v e > 0$

**shows**  $\exists d \in D_D. v \leq L (\text{mk-dec-det } d) v + e *_R 1$

$\langle proof \rangle$

**lemma** *step-mono-elem*:

**assumes**  $v \leq \mathcal{L}_b v e > 0$

**shows**  $\exists d \in D_R. v \leq L d v + e *_R 1$

$\langle proof \rangle$

**lemma**  *$\mathcal{P}_X$ - $L$ -le*:

**assumes**  $\mathcal{L}_b v \leq v p \in \Pi_{MR}$

**shows**  $\mathcal{P}_X p n (L (p n) v) \leq \mathcal{P}_X p n v$   
 $\langle \text{proof} \rangle$

**end**

**locale** *MDP-reward-disc* = *MDP-reward* *A K r l*

**for**

*A* **and**

*K* :: '*s* :: countable × '*a* :: countable ⇒ '*s* pmf **and**

*r l* +

**assumes**

*disc-lt-one* [*simp*]:  $l < 1$

**begin**

**definition** *is-opt-act* *v s* = *is-arg-max* ( $\lambda a. L_a a v s$ ) ( $\lambda a. a \in A s$ )

**abbreviation** *opt-acts* *v s*  $\equiv \{a. \text{is-opt-act } v s a\}$

**lemma** *summable-disc* [*intro*, *simp*]: *summable* ( $\lambda i. l \hat{\ } i * x$ )  
 $\langle \text{proof} \rangle$

**lemma** *summable-r-disc*[*intro*, *simp*]:

*summable* ( $\lambda i. |l \hat{\ } i * r (sa i)|$ )

*summable* ( $\lambda i. l \hat{\ } i * |r (sa i)|$ )

*summable* ( $\lambda i. l \hat{\ } i * r (sa i)$ )

$\langle \text{proof} \rangle$

**lemma** *summable-norm-disc-I*[*intro*]:

**assumes** *summable* ( $\lambda t. (l \hat{\ } t * \text{norm } F)$ )

**shows** *summable* ( $\lambda t. \text{norm } (l \hat{\ } t *_{R} F)$ )

$\langle \text{proof} \rangle$

**lemma** *summable-norm-disc-I'*[*intro*]:

**assumes** *summable* ( $\lambda t. (l \hat{\ } t * \text{norm } (F t))$ )

**shows** *summable* ( $\lambda t. \text{norm } (l \hat{\ } t *_{R} F t)$ )

$\langle \text{proof} \rangle$

**lemma** *summable-discI* [*intro*]:

**assumes** *bounded* (*range* *F*)

**shows** *summable* ( $\lambda t. l \hat{\ } t * \text{norm } (F t)$ )

$\langle \text{proof} \rangle$

**lemma** *summable-disc-reward* [*intro*]:

**assumes** *bounded* (*range* (*F* :: *nat* ⇒ '*b* :: *banach*))

**shows** *summable* ( $\lambda t. l \hat{\ } t *_{R} (F t)$ )

$\langle \text{proof} \rangle$

**lemma** *summable-norm-bfun-disc*: *summable* ( $\lambda t. l \hat{\ } t * \text{norm } (\text{apply-bfun } f t)$ )

$\langle \text{proof} \rangle$

**lemma** *summable-bfun-disc* [*simp*]: *summable* ( $\lambda t. l \hat{\wedge} t * (\text{apply-bfun } f \ t)$ )  
 ⟨*proof*⟩

**lemma** *norm-bfun-disc-le*:  $\text{norm } f \leq B \implies (\sum x. l \hat{\wedge} x * \text{norm } (\text{apply-bfun } f \ x)) \leq (\sum x. l \hat{\wedge} x * B)$   
 ⟨*proof*⟩

**lemma** *norm-bfun-disc-le'*:  $\text{norm } f \leq B \implies (\sum x. l \hat{\wedge} x * (\text{apply-bfun } f \ x)) \leq (\sum x. l \hat{\wedge} x * B)$   
 ⟨*proof*⟩

**lemma** *sum-disc-lim-l*:  $(\sum x. l \hat{\wedge} x * B) = B / (1-l)$   
 ⟨*proof*⟩

**lemma** *sum-disc-bound*:  $(\sum x. l \hat{\wedge} x * \text{apply-bfun } f \ x) \leq (\text{norm } f) / (1-l)$   
 ⟨*proof*⟩

**lemma** *sum-disc-bound'*:  
**fixes**  $f :: \text{nat} \Rightarrow 'b \Rightarrow_b \text{real}$   
**assumes**  $h: \forall n. \text{norm } (f \ n) \leq B$   
**shows**  $\text{norm } (\sum x. l \hat{\wedge} x *_R f \ x) \leq B / (1-l)$   
 ⟨*proof*⟩

**lemma** *abs- $\nu$ -trace-le*:  $|\nu\text{-trace } t| \leq (\sum i. l \hat{\wedge} i * r_M)$   
 ⟨*proof*⟩

**lemma** *integrable- $\nu$ -trace*: *integrable* ( $\mathcal{T} \ p \ s$ )  $\nu\text{-trace}$   
 ⟨*proof*⟩

**context**  
**fixes**  $p :: ('s, 'a) \text{pol}$   
**begin**

**lemma**  *$\nu$ -eq- $\nu$ -trace*:  $\nu \ p \ s = \int t. \nu\text{-trace } t \ \partial \mathcal{T} \ p \ s$   
 ⟨*proof*⟩

**lemma** *abs- $\nu$ -le*:  $|\nu \ p \ s| \leq (\sum i. l \hat{\wedge} i * r_M)$   
 ⟨*proof*⟩

**lemma**  *$\nu$ -le*:  $\nu \ p \ s \leq (\sum i. l \hat{\wedge} i * r_M)$   
 ⟨*proof*⟩

**lemma**  *$\nu$ -bfun*:  $\nu \ p \in \text{bfun}$   
 ⟨*proof*⟩

**lift-definition**  $\nu_b :: 's \Rightarrow_b \text{real is } \nu p$   
 ⟨proof⟩

**lemma** *norm- $\nu$ -le*:  $\text{norm } \nu_b \leq r_M / (1-l)$   
 ⟨proof⟩  
**end**

**lemma**  *$\nu$ -as-markovian*:  $\nu (\text{mk-markovian } (\text{as-markovian } p (\text{return-pmf } s))) s = \nu p s$   
 ⟨proof⟩

**lemma**  *$\nu_b$ -as-markovian*:  $\nu_b (\text{mk-markovian } (\text{as-markovian } p (\text{return-pmf } s))) s = \nu_b p s$   
 ⟨proof⟩

## 7.10 Optimal Reward

**definition**  *$\nu$ -MD*  $s \equiv \bigsqcup p \in \Pi_{MD}. \nu (\text{mk-markovian-det } p) s$   
**definition**  *$\nu$ -opt*  $s \equiv \bigsqcup p \in \Pi_{HR}. \nu p s$

**lemma**  *$\nu$ -opt-bfun*:  $\nu\text{-opt} \in \text{bfun}$   
 ⟨proof⟩

**lift-definition**  $\nu_b\text{-opt} :: 's \Rightarrow_b \text{real is } \nu\text{-opt}$   
 ⟨proof⟩

**lemma**  *$\nu_b$ -opt-eq*:  $\nu_b\text{-opt } s = (\bigsqcup p \in \Pi_{HR}. \nu_b p s)$   
 ⟨proof⟩

**lemma**  *$\nu$ -le- $\nu$ -opt* [*intro*]:  
**assumes** *is-policy*  $p$   
**shows**  $\nu p s \leq \nu\text{-opt } s$   
 ⟨proof⟩

**lemma**  *$\nu_b$ -le-opt* [*intro*]:  $p \in \Pi_{HR} \implies \nu_b p \leq \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  *$\nu_b$ -le-opt-MD* [*intro*]:  $p \in \Pi_{MD} \implies \nu_b (\text{mk-markovian-det } p) \leq \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  *$\nu_b$ -le-opt-DD* [*intro*]: *is-dec-det*  $d \implies \nu_b (\text{mk-stationary-det } d) \leq \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  *$\nu_b$ -le-opt-DR* [*intro*]: *is-dec*  $d \implies \nu_b (\text{mk-stationary } d) \leq \nu_b\text{-opt}$   
 ⟨proof⟩



**lemma**  $\nu_b$ -opt-eq-MR:  $\nu_b$ -opt  $s = (\bigsqcup p \in \Pi_{MR}. \nu_b (mk\text{-markovian } p)$   
 $s)$   
 $\langle proof \rangle$

**lemma** *summable-norm-disc-reward'*[simp]: *summable* ( $\lambda t. l \hat{\sim} t * norm$   
 $(\mathcal{P}_X p t (r\text{-dec}_b (p t)))$ )  
 $\langle proof \rangle$

**lemma** *summable-disc-reward- $\mathcal{P}_X$*  [simp]: *summable* ( $\lambda t. l \hat{\sim} t *_R \mathcal{P}_X p$   
 $t (r\text{-dec}_b (p t))$ )  
 $\langle proof \rangle$

**lemma** *disc-reward-tendsto*:  
 $(\lambda n. \sum t < n. l \hat{\sim} t *_R \mathcal{P}_X p t (r\text{-dec}_b (p t))) \longrightarrow (\sum t. l \hat{\sim} t *_R \mathcal{P}_X$   
 $p t (r\text{-dec}_b (p t)))$   
 $\langle proof \rangle$

**lemma**  $\nu$ -eq- $\mathcal{P}_X$ :  $\nu (mk\text{-markovian } p) = (\sum i. l \hat{\sim} i *_R \mathcal{P}_X p i (r\text{-dec}_b$   
 $(p i)))$   
 $\langle proof \rangle$

**lemma**  $\nu_b$ -eq- $\mathcal{P}_X$ :  $\nu_b (mk\text{-markovian } p) = (\sum i. l \hat{\sim} i *_R \mathcal{P}_X p i (r\text{-dec}_b$   
 $(p i)))$   
 $\langle proof \rangle$

**lemma**  $\nu_b$ -fin-tendsto- $\nu_b$ : ( $\nu_b$ -fin (*mk-markovian*  $p$ ))  $\longrightarrow \nu_b$  (*mk-markovian*  
 $p$ )  
 $\langle proof \rangle$

**lemma** *norm- $\mathcal{P}_1$ -l-less*: *norm* ( $l *_R \mathcal{P}_1 d$ )  $< 1$   
 $\langle proof \rangle$

**lemma** *disc- $\mathcal{P}_1$ -tendsto*: ( $\lambda n. (\sum t \leq n. l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t)$ )  $\longrightarrow (\sum t.$   
 $l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t)$   
 $\langle proof \rangle$

**lemma** *disc- $\mathcal{P}_1$ -lim*: *lim* ( $\lambda n. (\sum t \leq n. l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t)$ ) = ( $\sum t. l \hat{\sim} t$   
 $*_R \mathcal{P}_1 d \hat{\sim} t)$   
 $\langle proof \rangle$

**lemma** *convergent-disc- $\mathcal{P}_1$* : *convergent* ( $\lambda n. (\sum t \leq n. l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t)$ )  
 $\langle proof \rangle$

**lemma**  $\mathcal{P}_1$ -suminf-ge:  
**assumes**  $0 \leq u$  **shows**  $u \leq (\sum t. l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t) u$   
 $\langle proof \rangle$

**lemma**  $\mathcal{P}_1$ -suminf-pos:  
**assumes**  $0 \leq u$   
**shows**  $0 \leq (\sum t. l \hat{\sim} t *_R \mathcal{P}_1 d \hat{\sim} t) u$

$\langle proof \rangle$

**lemma** *lemma-6-1-2-b*:

**assumes**  $v \leq u$

**shows**  $(\sum t. l \hat{t} *_R \mathcal{P}_1 d \sim t) v \leq (\sum t. l \hat{t} *_R \mathcal{P}_1 d \sim t) u$

$\langle proof \rangle$

**lemma**  *$\nu$ -stationary*:  $\nu_b (mk\text{-stationary } d) = (\sum t. l \hat{t} *_R (\mathcal{P}_1 d \sim t)) (r\text{-dec}_b d)$

$\langle proof \rangle$

**lemma**  *$\nu$ -stationary-inv*:  $\nu_b (mk\text{-stationary } d) = inv_L (id\text{-blinfun } - l *_R \mathcal{P}_1 d) (r\text{-dec}_b d)$

$\langle proof \rangle$

The value of a markovian policy can be expressed in terms of  $L$ .

**lemma**  *$\nu$ -step*:  $\nu_b (mk\text{-markovian } p) = L (p 0) (\nu_b (mk\text{-markovian } (\lambda n. p (Suc n))))$

$\langle proof \rangle$

**lemma**  *$L$ - $\nu$ -fix*:  $\nu_b (mk\text{-stationary } d) = L d (\nu_b (mk\text{-stationary } d))$

$\langle proof \rangle$

**lemma**  *$L$ -fix- $\nu$* :

**assumes**  $L p v = v$

**shows**  $v = \nu_b (mk\text{-stationary } p)$

$\langle proof \rangle$

**lemma**  *$L$ - $\nu$ -fix-iff*:  $L d v = v \longleftrightarrow v = \nu_b (mk\text{-stationary } d)$

$\langle proof \rangle$

## 7.11 Properties of Solutions of the Optimality Equations

**abbreviation**  $\mathcal{P}_d p n v \equiv l \hat{n} *_R \mathcal{P}_X p n v$

**lemma**  *$\mathcal{P}_d$ -lim*:  $(\lambda n. (\mathcal{P}_d p n v)) \longrightarrow 0$

$\langle proof \rangle$

**lemma**  *$\mathcal{L}$ -dec-ge-opt*:

**assumes**  $\mathcal{L}_b v \leq v$

**shows**  $\nu_b\text{-opt} \leq v$

$\langle proof \rangle$

**lemma**  *$\mathcal{L}$ -inc-le-opt*:

**assumes**  $v \leq \mathcal{L}_b v$   
**shows**  $v \leq \nu_b\text{-opt}$   
 ⟨proof⟩  
**lemma**  $\mathcal{L}\text{-fix-imp-opt}$ :  
**assumes**  $v = \mathcal{L}_b v$   
**shows**  $v = \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  $\text{bounded-P}$ :  $\text{bounded } (P_1 \text{ ' } X)$   
 ⟨proof⟩

## 7.12 Solutions to the Optimality Equation

### 7.12.1 $\mathcal{L}_b$ and $L$ are Contraction Mappings

**declare**  $\text{bounded-apply-blinfun}$ [intro]  $\text{bounded-apply-bfun}$ '[intro]

**lemma**  $\text{contraction-L}$ :  $\text{dist } (\mathcal{L}_b v) (\mathcal{L}_b u) \leq l * \text{dist } v u$   
 ⟨proof⟩

**lemma**  $\text{is-contraction-L}$ :  $\text{is-contraction } \mathcal{L}_b$   
 ⟨proof⟩

**lemma**  $\text{contraction-L}$ :  $\text{dist } (L p v) (L p u) \leq l * \text{dist } v u$   
 ⟨proof⟩

**lemma**  $\text{is-contraction-L}$ :  $\text{is-contraction } (L p)$   
 ⟨proof⟩

### 7.12.2 Existence of a Fixpoint of $\mathcal{L}_b$

**lemma**  $\mathcal{L}_b\text{-conv}$ :  
 $\exists! v. \mathcal{L}_b v = v$   $(\lambda n. (\mathcal{L}_b \text{ } \overset{\sim}{n}) v) \longrightarrow (\text{THE } v. \mathcal{L}_b v = v)$   
 ⟨proof⟩

**lemma**  $\mathcal{L}_b\text{-fix-iff-opt}$  [simp]:  $\mathcal{L}_b v = v \longleftrightarrow v = \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  $\nu_b\text{-opt-fix}$ :  $\nu_b\text{-opt} = (\text{THE } v. \mathcal{L}_b v = v)$   
 ⟨proof⟩

**lemma**  $\mathcal{L}_b\text{-opt}$  [simp]:  $\mathcal{L}_b \nu_b\text{-opt} = \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  $\mathcal{L}_b\text{-lim}$ :  $(\lambda n. (\mathcal{L}_b \text{ } \overset{\sim}{n}) v) \longrightarrow \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  $\text{thm-6-2-6}$ :  $\nu_b p = \nu_b\text{-opt} \longleftrightarrow \mathcal{L}_b (\nu_b p) = \nu_b p$   
 ⟨proof⟩

**lemma** *thm-6-2-6'*:  $\nu p = \nu\text{-opt} \iff \mathcal{L}_b(\nu_b p) = \nu_b p$   
 ⟨proof⟩

### 7.13 Existence of Optimal Policies

**definition**  $\nu$ -improving  $v d \iff (\forall s. \text{is-arg-max } (\lambda d. (L d v) s) (\lambda d. d \in D_R) d)$

**lemma**  $\nu$ -improving-iff:  $\nu$ -improving  $v d \iff d \in D_R \wedge (\forall d' \in D_R. \forall s. L d' v s \leq L d v s)$   
 ⟨proof⟩

**lemma**  $\nu$ -improving-D-MR[dest]:  $\nu$ -improving  $v d \implies d \in D_R$   
 ⟨proof⟩

**lemma**  $\nu$ -improving-ge:  $\nu$ -improving  $v d \implies d' \in D_R \implies L d' v s \leq L d v s$   
 ⟨proof⟩

**lemma**  $\nu$ -improving-imp- $\mathcal{L}_b$ :  $\nu$ -improving  $v d \implies \mathcal{L}_b v = L d v$   
 ⟨proof⟩

**lemma**  $\mathcal{L}_b$ -imp- $\nu$ -improving:  
 assumes  $d \in D_R \mathcal{L}_b v = L d v$   
 shows  $\nu$ -improving  $v d$   
 ⟨proof⟩

**lemma**  $\nu$ -improving-alt:  
 assumes  $d \in D_R$   
 shows  $\nu$ -improving  $v d \iff \mathcal{L}_b v = L d v$   
 ⟨proof⟩

**definition**  $\nu$ -conserving  $d = \nu$ -improving  $(\nu_b\text{-opt}) d$

**lemma**  $\nu$ -conserving-iff:  $\nu$ -conserving  $d \iff d \in D_R \wedge (\forall d' \in D_R. \forall s. L d' \nu_b\text{-opt} s \leq L d \nu_b\text{-opt} s)$   
 ⟨proof⟩

**lemma**  $\nu$ -conserving-ge:  $\nu$ -conserving  $d \implies d' \in D_R \implies L d' \nu_b\text{-opt} s \leq L d \nu_b\text{-opt} s$   
 ⟨proof⟩

**lemma**  $\nu$ -conserving-imp- $\mathcal{L}_b$  [simp]:  $\nu$ -conserving  $d \implies L d \nu_b\text{-opt} = \nu_b\text{-opt}$   
 ⟨proof⟩

**lemma**  $\mathcal{L}_b$ -imp- $\nu$ -conserving:  
 assumes  $d \in D_R \mathcal{L}_b \nu_b\text{-opt} = L d \nu_b\text{-opt}$   
 shows  $\nu$ -conserving  $d$

$\langle proof \rangle$

**lemma**  $\nu$ -conserving-alt:

**assumes**  $d \in D_R$

**shows**  $\nu$ -conserving  $d \iff \mathcal{L}_b \nu_b\text{-opt} = L d \nu_b\text{-opt}$

$\langle proof \rangle$

**lemma**  $\nu$ -conserving-alt':

**assumes**  $d \in D_R$

**shows**  $\nu$ -conserving  $d \iff L d \nu_b\text{-opt} = \nu_b\text{-opt}$

$\langle proof \rangle$

### 7.13.1 Conserving Decision Rules are Optimal

**theorem** *ex-improving-imp-conserving*:

**assumes**  $\bigwedge v. \exists d. \nu$ -improving  $v$  (*mk-dec-det*  $d$ )

**shows**  $\exists d. \nu$ -conserving (*mk-dec-det*  $d$ )

$\langle proof \rangle$

**theorem** *conserving-imp-opt[simp]*:

**assumes**  $\nu$ -conserving (*mk-dec-det*  $d$ )

**shows**  $\nu_b$  (*mk-stationary-det*  $d$ ) =  $\nu_b\text{-opt}$

$\langle proof \rangle$

**lemma** *conserving-imp-opt'*:

**assumes**  $\exists d. \nu$ -conserving (*mk-dec-det*  $d$ )

**shows**  $\exists d \in D_D. (\nu_b$  (*mk-stationary-det*  $d$ )) =  $\nu_b\text{-opt}$

$\langle proof \rangle$

**theorem** *improving-att-imp-det-opt*:

**assumes**  $\bigwedge v. \exists d. \nu$ -improving  $v$  (*mk-dec-det*  $d$ )

**shows**  $\nu_b\text{-opt } s = (\bigsqcup d \in D_D. \nu_b$  (*mk-stationary-det*  $d$ )  $s$ )

$\langle proof \rangle$

**lemma**  $\mathcal{L}_b$ -sup-att-dec:

**assumes**  $d \in D_R \mathcal{L}_b v = L d v$

**shows**  $\exists d' \in D_D. \mathcal{L}_b v = L$  (*mk-dec-det*  $d'$ )  $v$

$\langle proof \rangle$

**lemma**  $\mathcal{L}_b$ -sup-att-dec':

**assumes**  $d \in D_R \mathcal{L}_b v = L d v$

**shows**  $\exists d' \in D_D. \nu$ -improving  $v$  (*mk-dec-det*  $d'$ )

$\langle proof \rangle$

### 7.13.2 Deterministic Decision Rules are Optimal

**lemma** *opt-imp-opt-dec-det*:

**assumes**  $p \in \Pi_{HR} \nu_b p = \nu_b\text{-opt}$

**shows**  $\exists d \in D_D. \nu_b$  (*mk-stationary-det*  $d$ ) =  $\nu_b\text{-opt}$

$\langle proof \rangle$

### 7.13.3 Optimal Decision Rules for Finite Action Spaces

**lemma** *ex-opt-act*:

**assumes**  $\bigwedge s. finite (A s)$

**shows**  $\exists a \in A s. L_a a (v :: - \Rightarrow_b -) s = \mathcal{L}_b v s$

$\langle proof \rangle$

**lemma** *ex-opt-dec-det*:

**assumes**  $\bigwedge s. finite (A s)$

**shows**  $\exists d \in D_D. L (mk-dec-det d) (v :: - \Rightarrow_b -) = \mathcal{L}_b v$

$\langle proof \rangle$

**lemma** *thm-6-2-10*:

**assumes**  $\bigwedge s. finite (A s)$

**shows**  $\exists d \in D_D. \nu_b-opt = \nu_b (mk-stationary-det d)$

$\langle proof \rangle$

### 7.13.4 Existence of Epsilon-Optimal Policies

**lemma** *ex-det-eps*:

**assumes**  $0 < e$

**shows**  $\exists d \in D_D. \mathcal{L}_b v \leq L (mk-dec-det d) v + e *_R 1$

$\langle proof \rangle$

**lemma** *thm-6-2-11*:

**assumes**  $eps > 0$

**shows**  $\exists d \in D_D. \nu_b-opt \leq \nu_b (mk-stationary-det d) + eps *_R 1$

$\langle proof \rangle$

**lemma** *ex-det-dist-eps*:

**assumes**  $0 < (e :: real)$

**shows**  $\exists d \in D_D. dist (\mathcal{L}_b v) (L (mk-dec-det d) v) \leq e$

$\langle proof \rangle$

**lemma** *less-imp-ex-add-le*:  $(x :: real) < y \implies \exists eps > 0. x + eps \leq y$

$\langle proof \rangle$

**lemma**  *$\nu_b-opt-le-det$* :  $\nu_b-opt s \leq (\bigsqcup d \in D_D. \nu_b (mk-stationary-det d) s)$

$\langle proof \rangle$

**lemma**  *$\nu_b-opt-eq-det$* :  $\nu_b-opt s = (\bigsqcup d \in D_D. \nu_b (mk-stationary-det d) s)$

$\langle proof \rangle$

**lemma** *lemma-6-3-1-a*:

**assumes**  $v0 \in bfun$

**shows** *uniform-limit UNIV*  $(\lambda n. ((\lambda v. \mathcal{L} (Bfun v)) \overset{\sim}{\sim} n) v0) \nu\text{-opt}$   
*sequentially*

$\langle proof \rangle$

**lemma** *dist-Suc-tendsto-zero*:

**assumes**  $(\lambda n. f n) \longrightarrow (y::\text{real-normed-vector})$

**shows**  $(\lambda n. \text{dist } (f n) (f (Suc n))) \longrightarrow 0$

$\langle proof \rangle$

**lemma** *dist- $\mathcal{L}_b$ -tendsto*:  $(\lambda n. \text{dist } ((\mathcal{L}_b \overset{\sim}{\sim} n) v) ((\mathcal{L}_b \overset{\sim}{\sim} (Suc n)) v))$   
 $\longrightarrow 0$

$\langle proof \rangle$

**definition** *max-L-ex s v*  $\equiv \text{has-arg-max } (\lambda a. L_a a v s) (A s)$

**lemma**  *$\nu_b$ -fin-zero[simp]*:  $\nu_b\text{-fin } p \ 0 = 0$

$\langle proof \rangle$

**lemma**  *$\nu_b$ -fin-Suc[simp]*:

$\nu_b\text{-fin } (mk\text{-stationary } d) (Suc n) = \nu_b\text{-fin } (mk\text{-stationary } d) n + ((l$   
 $*_R \mathcal{P}_1 d) \overset{\sim}{\sim} n) (r\text{-dec}_b d)$

$\langle proof \rangle$

**lemma**  *$\nu_b$ -fin-eq*:  $\nu_b\text{-fin } (mk\text{-stationary } d) n = (\sum i < n. ((l *_R \mathcal{P}_1$   
 $d) \overset{\sim}{\sim} i)) (r\text{-dec}_b d)$

$\langle proof \rangle$

**lemma** *L-iter*:  $(L d \overset{\sim}{\sim} m) v = \nu_b\text{-fin } (mk\text{-stationary } d) m + ((l *_R$   
 $\mathcal{P}_1 d) \overset{\sim}{\sim} m) v$

$\langle proof \rangle$

**lemma** *bounded-stationary- $\nu_b$ -fin*: *bounded*  $((\lambda x. (\nu_b\text{-fin } (mk\text{-stationary}$   
 $x) N) s) ' X)$

$\langle proof \rangle$

**lemma** *bounded-disc- $\mathcal{P}_1$* : *bounded*  $((\lambda x. (((l *_R \mathcal{P}_1 x) \overset{\sim}{\sim} m) v) s) ' X)$

$\langle proof \rangle$

**lemma** *bounded-disc- $\mathcal{P}_1'$* : *bounded*  $((\lambda x. ((\mathcal{P}_1 x \overset{\sim}{\sim} m) v) s) ' X)$

$\langle proof \rangle$

**lemma** *L-iter-le- $\mathcal{L}_b$* : *is-dec*  $d \implies (L d \overset{\sim}{\sim} n) v \leq (\mathcal{L}_b \overset{\sim}{\sim} n) v$

$\langle proof \rangle$

**end**

## 7.14 More Restrictive MDP Locales

**locale** *MDP-fin-acts* = *discrete-MDP* +  
**assumes**  $\bigwedge s. \text{finite } (A \ s)$

**locale** *MDP-att- $\mathcal{L}$*  = *MDP-reward-disc* *A K r l*  
**for**  
*A* **and**  
*K* :: 's :: countable  $\times$  'a :: countable  $\Rightarrow$  's pmf **and**  
*r* **and** *l* +  
**assumes** *Sup-att: max-L-ex* (*s* :: 's) *v*

**begin**

**theorem**  *$\mathcal{L}_b$ -eq-argmax- $L_a$* :

**fixes** *v* :: 's  $\Rightarrow_b$  real

**assumes** *is-arg-max* ( $\lambda a. L_a \ a \ v \ s$ ) ( $\lambda a. a \in A \ s$ ) *a*

**shows**  $\mathcal{L}_b \ v \ s = L_a \ a \ v \ s$

*<proof>*

**lemma**  *$L_a$ -le-arg-max*:  $a \in A \ s \Rightarrow L_a \ a \ v \ s \leq L_a \ (\text{arg-max-on } (\lambda a. L_a \ a \ v \ s) \ (A \ s)) \ v \ s$

*<proof>*

**lemma** *arg-max-on-in*: *has-arg-max* *f Q*  $\Rightarrow$  *arg-max-on* *f Q*  $\in Q$

*<proof>*

**lemma**  *$\mathcal{L}_b$ -eq- $L_a$ -max*:  $\mathcal{L}_b \ v \ s = L_a \ (\text{arg-max-on } (\lambda a. L_a \ a \ v \ s) \ (A \ s)) \ v \ s$

*<proof>*

**lemma** *ex-opt-det*:  $\exists d \in D_D. \mathcal{L}_b \ v = L \ (\text{mk-dec-det } d) \ v$

*<proof>*

**lemma** *ex-improving-det*:  $\exists d \in D_D. \nu\text{-improving } v \ (\text{mk-dec-det } d)$

*<proof>*

**end**

**locale** *MDP-act* = *discrete-MDP* *A K* **for** *A* :: 's :: countable  $\Rightarrow$  'a :: countable  
*set* **and** *K* +

**fixes** *arb-act* :: 'a set  $\Rightarrow$  'a

**assumes** *arb-act-in[simp]*:  $X \neq \{\}$   $\Rightarrow$  *arb-act*  $X \in X$

**locale** *MDP-act-disc* = *MDP-act* *A K* + *MDP-att- $\mathcal{L}$*  *A K r l*

**for** *A* :: 's :: countable  $\Rightarrow$  'a :: countable *set* **and** *K r l*

**begin**

**lemma** *is-opt-act-some*: *is-opt-act* *v s* (*arb-act* (*opt-acts* *v s*))

*<proof>*

**lemma** *some-opt-acts-in-A*: *arb-act* (*opt-acts* *v s*)  $\in A \ s$



*<proof>*

**lemma**  *$\nu$ -improving-opt-acts:  $\nu$ -improving  $v0$  (mk-dec-det ( $\lambda s.$  arb-act  
(opt-acts (apply-bfun  $v0$ )  $s$ )))*  
*<proof>*

**end**

**locale** *MDP-finite-type = MDP-reward-disc  $A K r l$*   
**for**  *$A$  and  $K :: 's :: finite \times 'a :: finite \Rightarrow 's pmf$  and  $r l$*

**end**

## References

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- [2] M. L. Puterman. *Markov Decision Processes: Discrete Stochastic Dynamic Programming*. Wiley Series in Probability and Statistics. Wiley, 1994.