

Negatively Associated Random Variables

Emin Karayel

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Abstract

Negative Association is a generalization of independence for random variables, that retains some of the key properties of independent random variables. In particular closure properties, such as composition with monotone functions, as well as, the well-known Chernoff-Hoeffding bounds.

This entry introduces the concept and verifies the most important closure properties, as well as, the concentration inequalities. It also verifies the FKG inequality, which is a generalization of Chebyshev's sum inequality for distributive lattices and a key tool for establishing negative association, but has also many applications beyond the context of negative association, in particular, statistical physics and graph theory.

As an example, permutation distributions are shown to be negatively associated, from which many more sets of negatively random variables can be derived, such as, e.g., n-subsets, or the balls-into-bins process.

Finally, the entry derives a correct false-positive rate for Bloom filters using the library.

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1 Preliminary Definitions and Lemmas

```
theory Negative-Association-Util
imports
  Concentration-Inequalities. Concentration-Inequalities-Preliminary
  Universal-Hash-Families. Universal-Hash-Families-More-Product-PMF
begin
```

```
abbreviation (input) flip ::  $\langle ('a \Rightarrow 'b \Rightarrow 'c) \Rightarrow 'b \Rightarrow 'a \Rightarrow 'c \rangle$  where
   $\langle flip f x y \equiv f y x \rangle$ 
```

Additional introduction rules for boundedness:

```
lemma bounded-const-min:
  fixes f :: ' $a \Rightarrow real$ 
  assumes bdd-below ( $f \cdot M$ )
  shows bounded ( $(\lambda x. min c (f x)) \cdot M$ )
  ⟨proof⟩
```

```
lemma bounded-prod:
  fixes f :: ' $i \Rightarrow 'a \Rightarrow real$ 
  assumes finite I
  assumes  $\bigwedge i. i \in I \implies bounded (f i \cdot T)$ 
  shows bounded ( $(\lambda x. (\prod i \in I. f i x)) \cdot T$ )
  ⟨proof⟩
```

```
lemma bounded-vec-mult-comp:
  fixes f g :: ' $a \Rightarrow real$ 
  assumes bounded ( $f \cdot T$ ) bounded ( $g \cdot T$ )
  shows bounded ( $(\lambda x. (f x) *_R (g x)) \cdot T$ )
  ⟨proof⟩
```

```
lemma bounded-max:
  fixes f :: ' $a \Rightarrow real$ 
  assumes bounded ( $(\lambda x. f x) \cdot T$ )
  shows bounded ( $(\lambda x. max c (f x)) \cdot T$ )
  ⟨proof⟩
```

```
lemma bounded-of-bool: bounded (range of-bool) ⟨proof⟩
```

```
lemma bounded-range-imp:
  assumes bounded (range f)
  shows bounded ( $(\lambda \omega. f (h \omega)) \cdot S$ )
  ⟨proof⟩
```

The following allows to state integrability and conditions about the integral simultaneously, e.g. *has-int-that M f* ($\lambda x. x \leq c$) says f is integrable on M and the integral smaller or equal to c.

```
definition has-int-that where
```

has-int-that $M f P = (\text{integrable } M f \wedge (P (\int \omega. f \omega \partial M)))$

lemma *true-eq-iff*: $P \implies \text{True} = P$ $\langle \text{proof} \rangle$

lemma *le-trans*: $y \leq z \implies x \leq y \longrightarrow x \leq (z :: 'a :: \text{order})$ $\langle \text{proof} \rangle$

lemma *has-int-that-mono*:

assumes $\bigwedge x. P x \longrightarrow Q x$

shows *has-int-that* $M f P \leq \text{has-int-that } M f Q$

$\langle \text{proof} \rangle$

lemma *has-int-thatD*:

assumes *has-int-that* $M f P$

shows *integrable* $M f P$ ($\text{integral}^L M f$)

$\langle \text{proof} \rangle$

This is useful to specify which components a functional depends on.

definition *depends-on* :: $(('a \Rightarrow 'b) \Rightarrow 'c) \Rightarrow 'a \text{ set} \Rightarrow \text{bool}$

where *depends-on* $f I = (\forall x y. \text{restrict } x I = \text{restrict } y I \longrightarrow f x = f y)$

lemma *depends-onI*:

assumes $\bigwedge x. f x = f (\lambda i. \text{if } i \in I \text{ then } (x i) \text{ else undefined})$

shows *depends-on* $f I$

$\langle \text{proof} \rangle$

lemma *depends-on-comp*:

assumes *depends-on* $f I$

shows *depends-on* $(g \circ f) I$

$\langle \text{proof} \rangle$

lemma *depends-on-comp-2*:

assumes *depends-on* $f I$

shows *depends-on* $(\lambda x. g (f x)) I$

$\langle \text{proof} \rangle$

lemma *depends-onD*:

assumes *depends-on* $f I$

shows $f \omega = f (\lambda i \in I. (\omega i))$

$\langle \text{proof} \rangle$

lemma *depends-onD2*:

assumes *depends-on* $f I$ $\text{restrict } x I = \text{restrict } y I$

shows $f x = f y$

$\langle \text{proof} \rangle$

lemma *depends-on-empty*:

assumes *depends-on* $f \{\}$

shows $f \omega = f \text{ undefined}$

$\langle \text{proof} \rangle$

```

lemma depends-on-mono:
  assumes  $I \subseteq J$  depends-on  $f I$ 
  shows depends-on  $f J$ 
  ⟨proof⟩

abbreviation square-integrable  $M f \equiv$  integrable  $M ((\text{power2} :: \text{real} \Rightarrow \text{real}) \circ f)$ 

There are many results in the field of negative association, where a statement is true for simultaneously monotone or anti-monotone functions. With the below construction, we introduce a mechanism where we can parameterize on the direction of a relation:

datatype RelDirection = Fwd | Rev

definition dir-le :: RelDirection  $\Rightarrow (('d :: \text{order}) \Rightarrow ('d :: \text{order}) \Rightarrow \text{bool})$  (infixl
 $\leq_{\geq 1} 60$ )
  where dir-le  $\eta = (\text{if } \eta = \text{Fwd} \text{ then } (\leq) \text{ else } (\geq))$ 

lemma dir-le[simp]:
   $(\leq_{\geq \text{Fwd}}) = (\leq)$ 
   $(\leq_{\geq \text{Rev}}) = (\geq)$ 
  ⟨proof⟩

definition dir-sign :: RelDirection  $\Rightarrow 'a :: \{\text{one}, \text{uminus}\}$  ( $\pm_1$ )
  where dir-sign  $\eta = (\text{if } \eta = \text{Fwd} \text{ then } 1 \text{ else } -1)$ 

lemma dir-le-refl:  $x \leq_{\geq \eta} x$ 
  ⟨proof⟩

lemma dir-sign[simp]:
   $(\pm_{\text{Fwd}}) = (1)$ 
   $(\pm_{\text{Rev}}) = (-1)$ 
  ⟨proof⟩

lemma conv-rel-to-sign:
  fixes  $f :: 'a :: \text{order} \Rightarrow \text{real}$ 
  shows monotone  $(\leq) (\leq_{\geq \eta}) f = \text{mono} ((*) (\pm_\eta) \circ f)$ 
  ⟨proof⟩

instantiation RelDirection :: times
begin
definition times-RelDirection :: RelDirection  $\Rightarrow$  RelDirection where
  times-RelDirection-def: times-RelDirection  $x y = (\text{if } x = y \text{ then } \text{Fwd} \text{ else } \text{Rev})$ 

instance ⟨proof⟩
end

lemmas rel-dir-mult[simp] = times-RelDirection-def

lemma dir-mult-hom:  $(\pm_\sigma * \tau) = (\pm_\sigma) * ((\pm_\tau) :: \text{real})$ 

```

$\langle proof \rangle$

Additional lemmas about clamp for the specific case on reals.

```
lemma clamp-eqI2:  
  assumes x ∈ {a..b::real}  
  shows x = clamp a b x  
 $\langle proof \rangle$ 
```

```
lemma clamp-eqI:  
  assumes |x| ≤ (a::real)  
  shows x = clamp (-a) a x  
 $\langle proof \rangle$ 
```

```
lemma clamp-real-def:  
  fixes x :: real  
  shows clamp a b x = max a (min x b)  
 $\langle proof \rangle$ 
```

```
lemma clamp-range:  
  assumes a ≤ b  
  shows ∀x. clamp a b x ≥ a ∧ ∀x. clamp a b x ≤ b range (clamp a b) ⊆ {a..b::real}  
 $\langle proof \rangle$ 
```

```
lemma clamp-abs-le:  
  assumes a ≥ (0::real)  
  shows |clamp (-a) a x| ≤ |x|  
 $\langle proof \rangle$ 
```

```
lemma bounded-clamp:  
  fixes a b :: real  
  shows bounded ((clamp a b ∘ f) ` S)  
 $\langle proof \rangle$ 
```

```
lemma bounded-clamp-alt:  
  fixes a b :: real  
  shows bounded ((λx. clamp a b (f x)) ` S)  
 $\langle proof \rangle$ 
```

```
lemma clamp-borel[measurable]:  
  fixes a b :: 'a::{euclidean-space,second-countable-topology}  
  shows clamp a b ∈ borel-measurable borel  
 $\langle proof \rangle$ 
```

```
lemma monotone-clamp:  
  assumes monotone (≤) (≤≥η) f  
  shows monotone (≤) (≤≥η) (λω. clamp a (b::real) (f ω))  
 $\langle proof \rangle$ 
```

This part introduces the term *KL-div* as the Kullback-Leibler divergence

between a pair of Bernoulli random variables. The expression is useful to express some of the Chernoff bounds more concisely [12, Th. 1].

```

lemma radon-nikodym-pmf:
  assumes set-pmf p ⊆ set-pmf q
  defines f ≡ (λx. ennreal (pmf p x / pmf q x))
  shows
    AE x in measure-pmf q. RN-deriv q p x = f x (is ?R1)
    AE x in measure-pmf p. RN-deriv q p x = f x (is ?R2)
  ⟨proof⟩

lemma KL-divergence-pmf:
  assumes set-pmf q ⊆ set-pmf p
  shows KL-divergence b (measure-pmf p) (measure-pmf q) = (ʃ x. log b (pmf q x
  / pmf p x) ∂q)
  ⟨proof⟩

definition KL-div :: real ⇒ real ⇒ real where
  KL-div p q = KL-divergence (exp 1) (bernoulli-pmf q) (bernoulli-pmf p)

lemma KL-div-eq:
  assumes q ∈ {0 <.. < 1} p ∈ {0..1}
  shows KL-div p q = p * ln (p/q) + (1-p) * ln ((1-p)/(1-q)) (is ?L = ?R)
  ⟨proof⟩

lemma KL-div-swap:
  assumes q ∈ {0 <.. < 1} p ∈ {0..1}
  shows KL-div p q = KL-div (1-p) (1-q)
  ⟨proof⟩

A few results about independent random variables:

lemma (in prob-space) indep-vars-const:
  assumes ⋀ i. i ∈ I ⇒ c i ∈ space (N i)
  shows indep-vars N (λi -. c i) I
  ⟨proof⟩

lemma indep-vars-map-pmf:
  assumes prob-space.indep-vars (measure-pmf p) (λ-. discrete) (λi. X i ∘ f) I
  shows prob-space.indep-vars (map-pmf f p) (λ-. discrete) X I
  ⟨proof⟩

lemma indep-var-pair-pmf:
  fixes x y :: 'a pmf
  shows prob-space.indep-var (pair-pmf x y) discrete fst discrete snd
  ⟨proof⟩

lemma measure-pair-pmf: measure (pair-pmf p q) (A × B) = measure p A *
  measure q B (is ?L = ?R)
  ⟨proof⟩

```

```
instance bool :: second-countable-topology
⟨proof⟩
```

```
end
```

2 Definition

This section introduces the concept of negatively associated random variables (RVs). The definition follows, as closely as possible, the original description by Joag-Dev and Proschan [13].

However, the following modifications have been made:

Singleton and empty sets of random variables are considered negatively associated. This is useful because it simplifies many of the induction proofs. The second modification is that the RV's don't have to be real valued. Instead the range can be into any linearly ordered space with the borel σ -algebra. This is a major enhancement compared to the original work, as well as results by following authors [6, 7, 8, 14, 17].

```
theory Negative-Association-Definition
imports
  Concentration-Inequalities.Bienaymes-Identity
  Negative-Association-Util
begin

context prob-space
begin

definition neg-assoc :: ('i ⇒ 'a ⇒ 'c :: {linorder-topology}) ⇒ 'i set ⇒ bool
where neg-assoc X I = (
  (∀ i ∈ I. random-variable borel (X i)) ∧
  (∀ (f::nat ⇒ ('i ⇒ 'c) ⇒ real) J. J ⊆ I ∧
    (∀ ℓ<2. bounded (range (f ℓ)) ∧ mono(f ℓ) ∧ depends-on (f ℓ) ([J,I-J]!ℓ) ∧
      f ℓ ∈ PiM ([J,I-J]!ℓ) (λ-. borel) →M borel) —→
    covariance (f 0 ∘ flip X) (f 1 ∘ flip X) ≤ 0))

lemma neg-assocI:
assumes ⋀ i. i ∈ I ⇒ random-variable borel (X i)
assumes ⋀ f g J. J ⊆ I
  ⇒ depends-on f J ⇒ depends-on g (I-J)
  ⇒ mono f ⇒ mono g
  ⇒ bounded (range f::real set) ⇒ bounded (range g)
  ⇒ f ∈ PiM J (λ-. borel) →M borel ⇒ g ∈ PiM (I-J) (λ-. borel) →M borel
  ⇒ covariance (f ∘ flip X) (g ∘ flip X) ≤ 0
shows neg-assoc X I
⟨proof⟩

lemma neg-assocI2:
```

```

assumes [measurable]:  $\bigwedge i. i \in I \implies \text{random-variable borel } (X i)$ 
assumes  $\bigwedge f g J. J \subseteq I$ 
 $\implies \text{depends-on } f J \implies \text{depends-on } g (I-J)$ 
 $\implies \text{mono } f \implies \text{mono } g$ 
 $\implies \text{bounded } (\text{range } f) \implies \text{bounded } (\text{range } g)$ 
 $\implies f \in \text{PiM } J (\lambda\text{-borel}) \rightarrow_M (\text{borel} :: \text{real measure})$ 
 $\implies g \in \text{PiM } (I-J) (\lambda\text{-borel}) \rightarrow_M (\text{borel} :: \text{real measure})$ 
 $\implies (\int \omega. f(\lambda i. X i \omega) * g(\lambda i. X i \omega) \partial M) \leq (\int \omega. f(\lambda i. X i \omega) \partial M) * (\int \omega. g(\lambda i. X i \omega) \partial M)$ 

```

shows neg-assoc $X I$
 $\langle \text{proof} \rangle$

lemma neg-assoc-empty:
neg-assoc $X \{\}$
 $\langle \text{proof} \rangle$

lemma neg-assoc-singleton:
assumes random-variable borel $(X i)$
shows neg-assoc $X \{i\}$
 $\langle \text{proof} \rangle$

lemma neg-assoc-imp-measurable:
assumes neg-assoc $X I$
assumes $i \in I$
shows random-variable borel $(X i)$
 $\langle \text{proof} \rangle$

Even though the assumption was that defining property is true for pairs of monotone functions over the random variables, it is also true for pairs of anti-monotone functions.

lemma neg-assoc-imp-mult-mono-bounded:
fixes $f g :: ('i \Rightarrow 'c:linorder-topology) \Rightarrow \text{real}$
assumes neg-assoc $X I$
assumes $J \subseteq I$
assumes bounded $(\text{range } f)$ bounded $(\text{range } g)$
assumes monotone $(\leq) (\leq_{\geq\eta}) f$ monotone $(\leq) (\leq_{\geq\eta}) g$
assumes depends-on $f J$ depends-on $g (I-J)$
assumes [measurable]: $f \in \text{borel-measurable } (\text{Pi}_M J (\lambda\text{-borel}))$
assumes [measurable]: $g \in \text{borel-measurable } (\text{Pi}_M (I-J) (\lambda\text{-borel}))$
shows
 $\text{covariance } (f \circ \text{flip } X) (g \circ \text{flip } X) \leq 0$
 $(\int \omega. f(\lambda i. X i \omega) * g(\lambda i. X i \omega) \partial M) \leq \text{expectation } (\lambda x. f(\lambda y. X y x)) * \text{expectation } (\lambda x. g(\lambda y. X y x))$
 $(\mathbf{is} ?L \leq ?R)$
 $\langle \text{proof} \rangle$

lemma lim-min-n: $(\lambda n. \min (\text{real } n) x) \longrightarrow x$
 $\langle \text{proof} \rangle$

lemma *lim-clamp-n*: $(\lambda n. \text{clamp}(-\text{real } n) (\text{real } n) x) \longrightarrow x$
 $\langle \text{proof} \rangle$

lemma *neg-assoc-imp-mult-mono*:

```
fixes f g :: ('i ⇒ 'c::linorder-topology) ⇒ real
assumes neg-assoc X I
assumes J ⊆ I
assumes square-integrable M (f ∘ flip X) square-integrable M (g ∘ flip X)
assumes monotone (≤) (≤≥η) f monotone (≤) (≤≥η) g
assumes depends-on f J depends-on g (I-J)
assumes [measurable]: f ∈ borel-measurable (Pi_M J (λ-. borel))
assumes [measurable]: g ∈ borel-measurable (Pi_M (I-J) (λ-. borel))
shows (∫ ω. f (λi. X i ω) * g (λi. X i ω) ∂M) ≤ (∫ x. f(λy. X y x) ∂M) * (∫ x. g(λy. X y x) ∂M)
(is ?L ≤ ?R)
⟨proof⟩
```

Property P4 [13]

lemma *neg-assoc-subset*:

```
assumes J ⊆ I
assumes neg-assoc X I
shows neg-assoc X J
⟨proof⟩
```

lemma *neg-assoc-imp-mult-mono-nonneg*:

```
fixes f g :: ('i ⇒ 'c::linorder-topology) ⇒ real
assumes neg-assoc X I J ⊆ I
assumes range f ⊆ {0..} range g ⊆ {0..}
assumes integrable M (f ∘ flip X) integrable M (g ∘ flip X)
assumes monotone (≤) (≤≥η) f monotone (≤) (≤≥η) g
assumes depends-on f J depends-on g (I-J)
assumes f ∈ borel-measurable (Pi_M J (λ-. borel)) g ∈ borel-measurable (Pi_M (I-J) (λ-. borel))
shows has-int-that M (λω. f (flip X ω) * g (flip X ω))
(λr. r ≤ expectation (f ∘ flip X) * expectation (g ∘ flip X))
⟨proof⟩
```

Property P2 [13]

lemma *neg-assoc-imp-prod-mono*:

```
fixes f :: 'i ⇒ ('c::linorder-topology) ⇒ real
assumes finite I
assumes neg-assoc X I
assumes ∀i. i ∈ I ⇒ integrable M (λω. f i (X i ω))
assumes ∀i. i ∈ I ⇒ monotone (≤) (≤≥η) (f i)
assumes ∀i. i ∈ I ⇒ range (f i) ⊆ {0..}
assumes ∀i. i ∈ I ⇒ f i ∈ borel-measurable borel
shows has-int-that M (λω. (∏ i∈I. f i (X i ω))) (λr. r ≤ (∏ i∈ I. expectation
(λω. f i (X i ω))))
⟨proof⟩
```

Property P5 [13]

```

lemma neg-assoc-compose:
  fixes f :: 'j ⇒ ('i ⇒ ('c::linorder-topology)) ⇒ ('d ::linorder-topology)
  assumes finite I
  assumes neg-assoc X I
  assumes ⋀j. j ∈ J ⇒ deps j ⊆ I
  assumes ⋀j1 j2. j1 ∈ J ⇒ j2 ∈ J ⇒ j1 ≠ j2 ⇒ deps j1 ∩ deps j2 = {}
  assumes ⋀j. j ∈ J ⇒ monotone (≤) (≤≥η) (f j)
  assumes ⋀j. j ∈ J ⇒ depends-on (f j) (deps j)
  assumes ⋀j. j ∈ J ⇒ f j ∈ borel-measurable (PiM (deps j) (λ-. borel))
  shows neg-assoc (λj ω. f j (λi. X i ω)) J
  ⟨proof⟩

lemma neg-assoc-compose-simple:
  fixes f :: 'i ⇒ ('c::linorder-topology) ⇒ ('d ::linorder-topology)
  assumes finite I
  assumes neg-assoc X I
  assumes ⋀i. i ∈ I ⇒ monotone (≤) (≤≥η) (f i)
  assumes [measurable]: ⋀i. i ∈ I ⇒ f i ∈ borel-measurable borel
  shows neg-assoc (λi ω. f i (X i ω)) I
  ⟨proof⟩

lemma covariance-distr:
  fixes f g :: 'b ⇒ real
  assumes [measurable]: φ ∈ M →M N f ∈ borel-measurable N g ∈ borel-measurable N
  shows prob-space.covariance (distr M N φ) f g = covariance (f ∘ φ) (g ∘ φ) (is ?L = ?R)
  ⟨proof⟩

lemma neg-assoc-iff-distr:
  assumes [measurable]: ⋀i. i ∈ I ⇒ X i ∈ borel-measurable M
  shows neg-assoc X I ↔
    prob-space.neg-assoc (distr M (PiM I (λ-. borel)) (λω. λi∈I. X i ω)) (flip id) I
    (is ?L ↔ ?R)
  ⟨proof⟩

lemma neg-assoc-cong:
  assumes finite I
  assumes [measurable]: ⋀i. i ∈ I ⇒ Y i ∈ borel-measurable M
  assumes neg-assoc X I ⋀i. i ∈ I ⇒ AE ω in M. X i ω = Y i ω
  shows neg-assoc Y I
  ⟨proof⟩

lemma neg-assoc-reindex-aux:
  assumes inj-on h I
  assumes neg-assoc X (h ` I)
  shows neg-assoc (λk. X (h k)) I
  ⟨proof⟩

```

```

lemma neg-assoc-reindex:
  assumes inj-on h I finite I
  shows neg-assoc X (h ` I)  $\longleftrightarrow$  neg-assoc ( $\lambda k. X (h k)$ ) I (is ?L  $\longleftrightarrow$  ?R)
  (proof)

```

```

lemma measurable-compose-merge-1:
  assumes depends-on h K
  assumes h  $\in$  PiM K M'  $\rightarrow_M$  N K  $\subseteq$  I  $\cup$  J
  assumes ( $\lambda x. \text{restrict} (\text{fst} (f x)) (K \cap I)$ )  $\in$  A  $\rightarrow_M$  PiM (K  $\cap$  I) M'
  assumes ( $\lambda x. \text{restrict} (\text{snd} (f x)) (K \cap J)$ )  $\in$  A  $\rightarrow_M$  PiM (K  $\cap$  J) M'
  shows ( $\lambda x. h(\text{merge} I J (f x))$ )  $\in$  A  $\rightarrow_M$  N
  (proof)

```

```

lemma measurable-compose-merge-2:
  assumes depends-on h K h  $\in$  PiM K M'  $\rightarrow_M$  N K  $\subseteq$  I  $\cup$  J
  assumes ( $\lambda x. \text{restrict} (f x) (K \cap I)$ )  $\in$  A  $\rightarrow_M$  PiM (K  $\cap$  I) M'
  assumes ( $\lambda x. \text{restrict} (g x) (K \cap J)$ )  $\in$  A  $\rightarrow_M$  PiM (K  $\cap$  J) M'
  shows ( $\lambda x. h(\text{merge} I J (f x, g x))$ )  $\in$  A  $\rightarrow_M$  N
  (proof)

```

```

lemma neg-assoc-combine:
  fixes I I1 I2 :: 'i set
  fixes X :: 'i  $\Rightarrow$  'a  $\Rightarrow$  ('b::linorder-topology)
  assumes finite I I1  $\cup$  I2 = I I1  $\cap$  I2 = {}
  assumes indep-var (PiM I1 ( $\lambda$ . borel)) ( $\lambda \omega. \lambda i \in I1. X i \omega$ ) (PiM I2 ( $\lambda$ . borel))
  ( $\lambda \omega. \lambda i \in I2. X i \omega$ )
  assumes neg-assoc X I1
  assumes neg-assoc X I2
  shows neg-assoc X I
  (proof)

```

Property P7 [13]

```

lemma neg-assoc-union:
  fixes I :: 'i set
  fixes p :: 'j  $\Rightarrow$  'i set
  fixes X :: 'i  $\Rightarrow$  'a  $\Rightarrow$  ('b::linorder-topology)
  assumes finite I  $\bigcup$  (p ` J) = I
  assumes indep-vars ( $\lambda j. \text{PiM} (p j) (\lambda \omega. \text{borel})$ ) ( $\lambda j \omega. \lambda i \in p j. X i \omega$ ) J
  assumes  $\bigwedge j. j \in J \implies$  neg-assoc X (p j)
  assumes disjoint-family-on p J
  shows neg-assoc X I
  (proof)

```

Property P5 [13]

```

lemma indep-imp-neg-assoc:
  assumes finite I
  assumes indep-vars ( $\lambda \omega. \text{borel}$ ) X I
  shows neg-assoc X I

```

```

⟨proof⟩
end

lemma neg-assoc-map-pmf:
  shows measure-pmf.neg-assoc (map-pmf f p) X I = measure-pmf.neg-assoc p (λi
  ω. X i (f ω)) I
    (is ?L  $\longleftrightarrow$  ?R)
  ⟨proof⟩
end

```

3 Chernoff-Hoeffding Bounds

This section shows that all the well-known Chernoff-Hoeffding bounds hold also for negatively associated random variables. The proofs follow the derivations by Hoeffding [11], as well as, Motwani and Raghavan [16, Ch. 4], with the modification that the crucial steps, where the classic proofs use independence, are replaced with the application of Property P2 for negatively associated RV's.

```

theory Negative-Association-Chernoff-Bounds
imports
  Negative-Association-Definition
  Concentration-Inequalities.McDiarmid-Inequality
  Weighted-Arithmetic-Geometric-Mean.Weighted-Arithmetic-Geometric-Mean
begin

context prob-space
begin

context
  fixes I :: 'i set
  fixes X :: 'i ⇒ 'a ⇒ real
  assumes na-X: neg-assoc X I
  assumes fin-I: finite I
begin

private lemma transfer-to-clamped-vars:
  assumes (forall i ∈ I. AE ω in M. X i ω ∈ {a i..b i}) ∧ a i ≤ b i
  assumes X-def: X = (λi. clamp (a i) (b i) ∘ X i)
  shows neg-assoc X I (is ?A)
    and ∀i. i ∈ I ⇒ expectation (X i) = expectation (X i)
    and P(ω in M. (∑ i ∈ I. X i ω) ≤≥_η c) = P(ω in M. (∑ i ∈ I. X i ω) ≤≥_η
    c) (is ?C)
    and ∀i ω. i ∈ I ⇒ X i ω ∈ {a i..b i}
    and ∀i S. i ∈ I ⇒ bounded (X i ` S)
    and ∀i. i ∈ I ⇒ expectation (X i) ∈ {a i..b i}

```

$\langle proof \rangle$

lemma *ln-one-plus-x-lower-bound*:
assumes $x \geq (0::real)$
shows $2*x/(2+x) \leq \ln(1+x)$
 $\langle proof \rangle$

Based on Theorem 4.1 by Motwani and Raghavan [16].

theorem *multiplicative-chernoff-bound-upper*:
assumes $\delta > 0$
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i))$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \geq (1+\delta) * \mu) \leq (\exp(\delta / ((1+\delta) \text{powr} (1+\delta))))$
 $\text{powr } \mu \text{ (is } ?L \leq ?R)$
and $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \geq (1+\delta) * \mu) \leq \exp(-(\delta^2) * \mu / (2+\delta))$
(is $- \leq ?R1)$
 $\langle proof \rangle$

lemma *ln-one-minus-x-lower-bound*:
assumes $x \in \{(0::real)..<1\}$
shows $(x^2/2 - x)/(1-x) \leq \ln(1-x)$
 $\langle proof \rangle$

Based on Theorem 4.2 by Motwani and Raghavan [16].

theorem *multiplicative-chernoff-bound-lower*:
assumes $\delta \in \{0<..<1\}$
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i))$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \leq (1-\delta) * \mu) \leq (\exp(-\delta / ((1-\delta) \text{powr} (1-\delta))))$
 $\text{powr } \mu \text{ (is } ?L \leq ?R)$
and $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \leq (1-\delta) * \mu) \leq (\exp(-(\delta^2) * \mu / 2))$ **(is** $- \leq ?R1)$
 $\langle proof \rangle$

theorem *multiplicative-chernoff-bound-two-sided*:
assumes $\delta \in \{0<..<1\}$
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i))$
shows $\mathcal{P}(\omega \text{ in } M. |(\sum i \in I. X i \omega) - \mu| \geq \delta * \mu) \leq 2 * (\exp(-(\delta^2) * \mu / 3))$ **(is** $?L \leq ?R)$
 $\langle proof \rangle$

lemma *additive-chernoff-bound-upper-aux*:
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\} I \neq \{\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i)) / \text{real}(\text{card } I)$
assumes $\delta \in \{0<..<1-\mu\} \mu \in \{0<..<1\}$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \geq (\mu + \delta) * \text{real}(\text{card } I)) \leq \exp(-\text{real}(\text{card } I) * \text{KL-div}(\mu + \delta, \mu))$
(is $?L \leq ?R)$

$\langle proof \rangle$

lemma additive-chernoff-bound-upper-aux-2:
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\} I \neq \{\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i)) / \text{real}(\text{card } I)$
assumes $\mu \in \{0 <.. < 1\}$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \geq \text{real}(\text{card } I)) \leq \exp(-\text{real}(\text{card } I) * KL\text{-div}_{1 \mu})$
(is $?L \leq ?R$)
 $\langle proof \rangle$

Based on Theorem 1 by Hoeffding [11].

lemma additive-chernoff-bound-upper:
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\} I \neq \{\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i)) / \text{real}(\text{card } I)$
assumes $\delta \in \{0..1 - \mu\} \mu \in \{0 <.. < 1\}$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \geq (\mu + \delta) * \text{real}(\text{card } I)) \leq \exp(-\text{real}(\text{card } I) * KL\text{-div}_{(\mu + \delta) \mu})$
(is $?L \leq ?R$)
 $\langle proof \rangle$

Based on Theorem 2 by Hoeffding [11].

lemma hoeffding-bound-upper:
assumes $\bigwedge i. i \in I \implies a i \leq b i$
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{a..b\}$
defines $n \equiv \text{real}(\text{card } I)$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i))$
assumes $\delta \geq 0 (\sum i \in I. (b i - a i)^2) > 0$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \geq \mu + \delta * n) \leq \exp(-2 * (n * \delta)^2 / (\sum i \in I. (b i - a i)^2))$
(is $?L \leq ?R$)
 $\langle proof \rangle$

end

Dual and two-sided versions of Theorem 1 and 2 by Hoeffding [11].

lemma additive-chernoff-bound-lower:
assumes neg-assoc $X I$ finite I
assumes $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{0..1\} I \neq \{\}$
defines $\mu \equiv (\sum i \in I. \text{expectation}(X i)) / \text{real}(\text{card } I)$
assumes $\delta \in \{0..\mu\} \mu \in \{0 <.. < 1\}$
shows $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \leq (\mu - \delta) * \text{real}(\text{card } I)) \leq \exp(-\text{real}(\text{card } I) * KL\text{-div}_{(\mu - \delta) \mu})$
(is $?L \leq ?R$)
 $\langle proof \rangle$

lemma hoeffding-bound-lower:
assumes neg-assoc $X I$ finite I
assumes $\bigwedge i. i \in I \implies a i \leq b i$

```

assumes  $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{a..b\}$ 
defines  $n \equiv \text{real}(\text{card } I)$ 
defines  $\mu \equiv (\sum i \in I. \text{expectation}(X i))$ 
assumes  $\delta \geq 0 (\sum i \in I. (b i - a i)^2) > 0$ 
shows  $\mathcal{P}(\omega \text{ in } M. (\sum i \in I. X i \omega) \leq \mu - \delta * n) \leq \exp(-2 * (n * \delta)^2 / (\sum i \in I. (b i - a i)^2))$ 
(is ?L ≤ ?R)
⟨proof⟩

lemma hoeffding-bound-two-sided:
assumes neg-assoc X I finite I
assumes  $\bigwedge i. i \in I \implies a i \leq b i$ 
assumes  $\bigwedge i. i \in I \implies AE \omega \text{ in } M. X i \omega \in \{a..b\} I \neq \{\}$ 
defines  $n \equiv \text{real}(\text{card } I)$ 
defines  $\mu \equiv (\sum i \in I. \text{expectation}(X i))$ 
assumes  $\delta \geq 0 (\sum i \in I. (b i - a i)^2) > 0$ 
shows  $\mathcal{P}(\omega \text{ in } M. |(\sum i \in I. X i \omega) - \mu| \geq \delta * n) \leq 2 * \exp(-2 * (n * \delta)^2 / (\sum i \in I. (b i - a i)^2))$ 
(is ?L ≤ ?R)
⟨proof⟩

end

end

```

4 The FKG inequality

The FKG inequality [9] is a generalization of Chebyshev's less known other inequality. It is sometimes referred to as Chebyshev's sum inequality. Although there is also a continuous version, which can be stated as:

$$E[fg] \geq E[f]E[g]$$

where f, g are continuous simultaneously monotone or simultaneously antimonotone functions on the Lebesgue probability space $[a, b] \subseteq \mathbb{R}$. (Ef denotes the expectation of the function.)

Note that the inequality is also true for totally ordered discrete probability spaces, for example: $\{1, \dots, n\}$ with uniform probabilities.

The FKG inequality is essentially a generalization of the above to not necessarily totally ordered spaces, but finite distributive lattices.

The proof follows the derivation in the book by Alon and Spencer [2, Ch. 6].

```

theory Negative-Association-FKG-Inequality
imports
  Negative-Association-Util
  Birkhoff-Finite-Distributive-Lattices.Birkhoff-Finite-Distributive-Lattices

```

begin

theorem *four-functions-helper*:
fixes $\varphi :: nat \Rightarrow 'a set \Rightarrow real$
assumes *finite I*
assumes $\bigwedge i. i \in \{0..3\} \implies \varphi i \in Pow I \rightarrow \{0..\}$
assumes $\bigwedge A B. A \subseteq I \implies B \subseteq I \implies \varphi 0 A * \varphi 1 B \leq \varphi 2 (A \cup B) * \varphi 3 (A \cap B)$
shows $(\sum A \in Pow I. \varphi 0 A) * (\sum B \in Pow I. \varphi 1 B) \leq (\sum C \in Pow I. \varphi 2 C) * (\sum D \in Pow I. \varphi 3 D)$
(proof)

The following is the Ahlswede-Daykin inequality [1] also referred to by Alon and Spencer as the four functions theorem [2, Th. 6.1.1].

theorem *four-functions*:
fixes $\alpha \beta \gamma \delta :: 'a set \Rightarrow real$
assumes *finite I*
assumes $\alpha \in Pow I \rightarrow \{0..\} \beta \in Pow I \rightarrow \{0..\} \gamma \in Pow I \rightarrow \{0..\} \delta \in Pow I \rightarrow \{0..\}$
assumes $\bigwedge A B. A \subseteq I \implies B \subseteq I \implies \alpha A * \beta B \leq \gamma (A \cup B) * \delta (A \cap B)$
assumes $M \subseteq Pow I N \subseteq Pow I$
shows $(\sum A \in M. \alpha A) * (\sum B \in N. \beta B) \leq (\sum C | \exists A \in M. \exists B \in N. C = A \cup B. \gamma C) * (\sum D | \exists A \in M. \exists B \in N. D = A \cap B. \delta D)$
(is ?L ≤ ?R)
(proof)

Using Birkhoff's Representation Theorem [3, 5] it is possible to generalize the previous to finite distributive lattices [2, Cor. 6.1.2].

lemma *four-functions-in-lattice*:
fixes $\alpha \beta \gamma \delta :: 'a :: finite-distrib-lattice \Rightarrow real$
assumes *range α ⊆ {0..} range β ⊆ {0..} range γ ⊆ {0..} range δ ⊆ {0..}*
assumes $\bigwedge x y. \alpha x * \beta y \leq \gamma (x \sqcup y) * \delta (x \sqcap y)$
shows $(\sum x \in M. \alpha x) * (\sum y \in N. \beta y) \leq (\sum c | \exists x \in M. \exists y \in N. c = x \sqcup y. \gamma c) * (\sum d | \exists x \in M. \exists y \in N. d = x \sqcap y. \delta d)$
(is ?L ≤ ?R)
(proof)

theorem *fkg-inequality*:
fixes $\mu :: 'a :: finite-distrib-lattice \Rightarrow real$
assumes *range μ ⊆ {0..} range f ⊆ {0..} range g ⊆ {0..}*
assumes $\bigwedge x y. \mu x * \mu y \leq \mu (x \sqcup y) * \mu (x \sqcap y)$
assumes *mono f mono g*
shows $(\sum x \in UNIV. \mu x * f x) * (\sum x \in UNIV. \mu x * g x) \leq (\sum x \in UNIV. \mu x * f x * g x) * sum \mu UNIV$
(is ?L ≤ ?R)
(proof)

theorem *fkg-inequality-gen*:
fixes $\mu :: 'a :: finite-distrib-lattice \Rightarrow real$

```

assumes range  $\mu \subseteq \{0..\}$ 
assumes  $\bigwedge x y. \mu x * \mu y \leq \mu (x \sqcup y) * \mu (x \sqcap y)$ 
assumes monotone  $(\leq) (\leq_{\geq\tau}) f$  monotone  $(\leq) (\leq_{\geq\sigma}) g$ 
shows  $(\sum x \in UNIV. \mu x * f x) * (\sum x \in UNIV. \mu x * g x) \leq_{\geq\tau * \sigma} (\sum x \in UNIV. \mu x * f x * g x) * \text{sum } \mu UNIV$ 
  (is ?L  $\leq_{\geq\tau} ?R$ )
  ⟨proof⟩

theorem fkg-inequality-pmf:
  fixes M :: ('a :: finite-distrib-lattice) pmf
  fixes f g :: 'a ⇒ real
  assumes  $\bigwedge x y. \text{pmf } M x * \text{pmf } M y \leq \text{pmf } M (x \sqcup y) * \text{pmf } M (x \sqcap y)$ 
  assumes monotone  $(\leq) (\leq_{\geq\tau}) f$  monotone  $(\leq) (\leq_{\geq\sigma}) g$ 
  shows  $(\int x. f x \partial M) * (\int x. g x \partial M) \leq_{\geq\tau * \sigma} (\int x. f x * g x \partial M)$ 
    (is ?L  $\leq_{\geq\tau} ?R$ )
  ⟨proof⟩

end

```

5 Preliminary Results on Lattices

This entry establishes a few missing lemmas for the set-based theory of lattices from “HOL-Algebra”. In particular, it introduces the sublocale for distributive lattices.

More crucially, a transfer theorem which can be used in conjunction with the Types-To-Sets mechanism to be able to work with locally defined finite distributive lattices.

This is being needed for the verification of the negative association of permutation distributions in Section 6.

```

theory Negative-Association-More-Lattices
  imports HOL-Algebra.Lattice
begin

```

Lemma 1 Birkhoff Lattice Theory, p.8, L3

```

lemma (in lattice) meet-assoc-law:
  assumes x ∈ carrier L y ∈ carrier L z ∈ carrier L
  shows x ∩ (y ∩ z) = (x ∩ y) ∩ z
  ⟨proof⟩

```

Lemma 1 Birkhoff Lattice Theory, p.8, L3

```

lemma (in lattice) join-assoc-law:
  assumes x ∈ carrier L y ∈ carrier L z ∈ carrier L
  shows x ∪ (y ∪ z) = (x ∪ y) ∪ z
  ⟨proof⟩

```

Lemma 1 Birkhoff Lattice Theory, p.8, L4

lemma (in lattice) absorbtion-law:
assumes $x \in \text{carrier } L$ $y \in \text{carrier } L$
shows $x \sqcap (x \sqcup y) = x$ $x \sqcup (x \sqcap y) = x$
 $\langle proof \rangle$

Theorem 9 Birkhoff Lattice Theory, p.11

lemma (in lattice) distrib-laws-equiv:
defines $\text{meet-distrib} \equiv (\forall x y z. \{x,y,z\} \subseteq \text{carrier } L \longrightarrow (x \sqcap (y \sqcup z)) = (x \sqcap y) \sqcup (x \sqcap z))$
defines $\text{join-distrib} \equiv (\forall x y z. \{x,y,z\} \subseteq \text{carrier } L \longrightarrow (x \sqcup (y \sqcap z)) = (x \sqcup y) \sqcap (x \sqcup z))$
shows $\text{meet-distrib} \longleftrightarrow \text{join-distrib}$
 $\langle proof \rangle$

lemma (in lattice) lub-unique-set:
assumes $\text{is-lub } L z S$
shows $z = \bigcup S$
 $\langle proof \rangle$

lemma (in lattice) lub-unique:
assumes $\text{is-lub } L z \{x,y\}$
shows $z = x \sqcup y$
 $\langle proof \rangle$

lemma (in lattice) glb-unique-set:
assumes $\text{is-glb } L z S$
shows $z = \bigcap S$
 $\langle proof \rangle$

lemma (in lattice) glb-unique:
assumes $\text{is-glb } L z \{x,y\}$
shows $z = x \sqcap y$
 $\langle proof \rangle$

lemma (in lattice) inf-lower:
assumes $S \subseteq \text{carrier } L s \in S \text{ finite } S$
shows $\bigcap S \sqsubseteq s$
 $\langle proof \rangle$

lemma (in lattice) sup-upper:
assumes $S \subseteq \text{carrier } L s \in S \text{ finite } S$
shows $s \sqsubseteq \bigcup S$
 $\langle proof \rangle$

locale distrib-lattice = lattice +
assumes max-distrib:
 $x \in \text{carrier } L \implies y \in \text{carrier } L \implies z \in \text{carrier } L \implies (x \sqcap (y \sqcup z)) = (x \sqcap y) \sqcup (x \sqcap z)$
begin

```

lemma min-distrib:
  assumes  $x \in \text{carrier } L$   $y \in \text{carrier } L$   $z \in \text{carrier } L$ 
  shows  $(x \sqcup (y \sqcap z)) = (x \sqcup y) \sqcap (x \sqcup z)$ 
   $\langle proof \rangle$ 

end

locale finite-ne-distrib-lattice = distrib-lattice +
  assumes non-empty-carrier:  $\text{carrier } L \neq \{\}$ 
  assumes finite-carrier: finite ( $\text{carrier } L$ )
begin

lemma bounded-lattice-axioms-1:  $\exists x. \text{least } L x (\text{carrier } L)$ 
   $\langle proof \rangle$ 

lemma bounded-lattice-axioms-2:  $\exists x. \text{greatest } L x (\text{carrier } L)$ 
   $\langle proof \rangle$ 

sublocale bounded-lattice
   $\langle proof \rangle$ 

lemma inf-empty:  $\sqcap \{\} = \top$ 
   $\langle proof \rangle$ 

lemma inf-closed:  $S \subseteq \text{carrier } L \implies \sqcap S \in \text{carrier } L$ 
   $\langle proof \rangle$ 

lemma inf-insert:
  assumes  $x \in \text{carrier } L$   $S \subseteq \text{carrier } L$ 
  shows  $\sqcap (\text{insert } x S) = x \sqcap (\sqcap S)$ 
   $\langle proof \rangle$ 

lemma sup-empty:  $\sqcup \{\} = \perp$ 
   $\langle proof \rangle$ 

lemma sup-closed:  $S \subseteq \text{carrier } L \implies \sqcup S \in \text{carrier } L$ 
   $\langle proof \rangle$ 

lemma sup-insert:
  assumes  $x \in \text{carrier } L$   $S \subseteq \text{carrier } L$ 
  shows  $\sqcup (\text{insert } x S) = x \sqcup (\sqcup S)$ 
   $\langle proof \rangle$ 

lemma inf-carrier:  $\sqcap (\text{carrier } L) = \perp$ 
   $\langle proof \rangle$ 

lemma sup-carrier:  $\sqcup (\text{carrier } L) = \top$ 
   $\langle proof \rangle$ 

```

```

lemma transfer-to-type:
  assumes finite (carrier L) type-definition Rep Abs (carrier L)
  defines inf' ≡ (λM. Abs (Π Rep ‘ M)))
  defines sup' ≡ (λM. Abs (Σ Rep ‘ M)))
  defines join' ≡ (λx y. Abs (Rep x ∩ Rep y))
  defines le' ≡ (λx y. (Rep x ⊑ Rep y))
  defines less' ≡ (λx y. (Rep x ⊂ Rep y))
  defines meet' ≡ (λx y. (Abs (Rep x ∙ Rep y)))
  defines bot' ≡ (Abs ⊥ :: 'c)
  defines top' ≡ Abs ⊤
  shows class.finite-distrib-lattice inf' sup' join' le' less' meet' bot' top'
  ⟨proof⟩

end

end

```

6 Permutation Distributions

One of the fundamental examples for negatively associated random variables are permutation distributions.

Let x_1, \dots, x_n be n (not-necessarily) distinct values from a totally ordered set, then we choose a permutation $\sigma : \{0, \dots, n-1\} \rightarrow \{0, \dots, n-1\}$ uniformly at random. Then the random variables defined by $X_i(\sigma) = x_{\sigma(i)}$ are negatively associated.

An important special case is the case where x consists of 1 one and $(n-1)$ zeros, modelling randomly putting a ball into one of n bins. Of course the process can be repeated independently, the resulting distribution is also referred to as the balls into bins process. Because of the closure properties established before, it is possible to conclude that the number of hits of each bin in such a process are also negatively associated random variables.

In this section, we will derive that permutation distributions are negatively associated. The proof follows Dubashi [8, Th. 10] closely. A very short proof was presented in the work by Joag-Dev [13], however after close inspection that proof seemed to missing a lot of details. In fact, I don't think it is correct.

```

theory Negative-Association-Permutation-Distributions
  imports
    Negative-Association-Definition
    Negative-Association-FKG-Inequality
    Negative-Association-More-Lattices
    Finite-Fields.Finite-Fields-More-PMF
    HOL-Types-To-Sets.Types-To-Sets

```

*Executable-Randomized-Algorithms.Randomized-Algorithm
Twelvefold-Way.Card-Bijections*

begin

The following introduces a lattice for n-element subsets of a finite set (with size larger or equal to n.) A subset x is smaller or equal to y , if the smallest element of x is smaller or equal to the smallest element of y , the second smallest element of x is smaller or equal to the second smallest element of y , etc.)

The lattice is introduced without name by Dubashi [?, Example 7].

definition *le-ordered-set-lattice* :: ('*a::linorder*) *set* \Rightarrow '*a set* \Rightarrow *bool*
where *le-ordered-set-lattice S T* = *list-all2* (\leq) (*sorted-list-of-set S*) (*sorted-list-of-set T*)

definition *ordered-set-lattice* :: ('*a :: linorder*) *set* \Rightarrow *nat* \Rightarrow '*a set gorder*
where *ordered-set-lattice S n* =
 (carrier = {*T*. *T* \subseteq *S* \wedge finite *T* \wedge card *T* = *n*},
 eq = (=),
 le = *le-ordered-set-lattice*)

definition *osl-repr* :: ('*a :: linorder*) *set* \Rightarrow *nat* \Rightarrow '*a set* \Rightarrow '*a
where osl-repr S n e* = ($\lambda i \in \{.. < n\}$. *sorted-list-of-set e ! i*)

lemma *osl-carr-sorted-list-of-set*:
assumes finite *S n* \leq card *S*
assumes *s* \in carrier (*ordered-set-lattice S n*)
defines *t* \equiv *sorted-list-of-set s*
shows finite *s* card *s* = *n* *s* \subseteq *S* length *t* = *n* set *t* = *s* sorted-wrt ($<$) *t*
{proof}

lemma *ordered-set-lattice-carrier-intro*:
assumes finite *S n* \leq card *S*
assumes set *s* \subseteq *S* distinct *s* length *s* = *n*
shows set *s* \in carrier (*ordered-set-lattice S n*)
{proof}

lemma *osl-list-repr-inj*:
assumes finite *S n* \leq card *S*
assumes *s* \in carrier (*ordered-set-lattice S n*)
assumes *t* \in carrier (*ordered-set-lattice S n*)
assumes $\bigwedge i$. osl-repr *S n s i* = osl-repr *S n t i*
shows *s* = *t*
{proof}

lemma *osl-leD*:
assumes finite *S n* \leq card *S*
assumes *e* \in carrier (*ordered-set-lattice S n*)
assumes *f* \in carrier (*ordered-set-lattice S n*)

```

shows  $e \sqsubseteq_{\text{ordered-set-lattice } S n} f \longleftrightarrow (\forall i. \text{osl-repr } S n e i \leq \text{osl-repr } S n f i)$  (is
? $L = ?R$ )
⟨proof⟩

lemma ordered-set-lattice-partial-order:
  fixes  $S :: ('a :: \text{linorder}) \text{ set}$ 
  assumes  $\text{finite } S n \leq \text{card } S$ 
  shows partial-order (ordered-set-lattice  $S n$ )
⟨proof⟩

lemma map2-max-mono:
  fixes  $xs :: ('a :: \text{linorder}) \text{ list}$ 
  assumes  $\text{length } xs = \text{length } ys$ 
  assumes  $\text{sorted-wrt } (<) \text{ xs sorted-wrt } (<) \text{ ys}$ 
  shows  $\text{sorted-wrt } (<) (\text{map2 max } xs \text{ } ys)$ 
⟨proof⟩

lemma map2-min-mono:
  fixes  $xs :: ('a :: \text{linorder}) \text{ list}$ 
  assumes  $\text{length } xs = \text{length } ys$ 
  assumes  $\text{sorted-wrt } (<) \text{ xs sorted-wrt } (<) \text{ ys}$ 
  shows  $\text{sorted-wrt } (<) (\text{map2 min } xs \text{ } ys)$ 
⟨proof⟩

lemma ordered-set-lattice-carrier-finite-ne:
  assumes  $\text{finite } S n \leq \text{card } S$ 
  shows  $\text{carrier } (\text{ordered-set-lattice } S n) \neq \{\}$   $\text{finite } (\text{carrier } (\text{ordered-set-lattice } S n))$ 
⟨proof⟩

lemma ordered-set-lattice-lattice:
  fixes  $S :: ('a :: \text{linorder}) \text{ set}$ 
  assumes  $\text{finite } S n \leq \text{card } S$ 
  shows finite-ne-distrib-lattice (ordered-set-lattice  $S n$ )
⟨proof⟩

lemma insort-eq:
  fixes  $xs :: ('a :: \text{linorder}) \text{ list}$ 
  assumes sorted  $xs$ 
  shows  $\exists ys \text{ zs}. \text{insort } e \text{ } xs = ys @ e \# zs \wedge ys @ zs = xs \wedge \text{set } ys \subseteq \{.. < e\} \wedge \text{set } zs \subseteq \{e..\}$ 
⟨proof⟩

lemma list-all2-insort:
  fixes  $xs \text{ } ys :: ('a :: \text{linorder}) \text{ list}$ 
  assumes  $\text{length } xs = \text{length } ys$   $\text{sorted } xs \text{ sorted } ys$ 
  shows  $\text{list-all2 } (\leq) \text{ } xs \text{ } ys \longleftrightarrow \text{list-all2 } (\leq) (\text{insort } e \text{ } xs) (\text{insort } e \text{ } ys)$ 
⟨proof⟩

```

```

lemma le-ordered-set-lattice-diff:
  fixes x y :: ('a :: linorder) set
  assumes finite x finite y card x = card y
  shows le-ordered-set-lattice x y  $\longleftrightarrow$  le-ordered-set-lattice (x - y) (y - x)
  ⟨proof⟩

lemma ordered-set-lattice-carrier:
  assumes T ∈ carrier (ordered-set-lattice S n)
  shows finite T card T = n T ⊆ S
  ⟨proof⟩

lemma ordered-set-lattice-dual:
  assumes finite S n ≤ card S
  defines L ≡ ordered-set-lattice S n
  defines M ≡ ordered-set-lattice S (card S - n)
  shows
     $\bigwedge x. x \in \text{carrier } L \implies (S-x) \in \text{carrier } M$ 
     $\bigwedge x. x \in \text{carrier } M \implies (S-x) \in \text{carrier } L$ 
     $\bigwedge x y. x \in \text{carrier } L \wedge y \in \text{carrier } L \implies x \sqsubseteq_L y \longleftrightarrow (S-y) \sqsubseteq_M (S-x)$ 
  ⟨proof⟩

lemma bij-betw-ord-set-lattice-pairs:
  assumes finite S n ≤ card S
  defines L ≡ ordered-set-lattice S n
  assumes x ∈ carrier L y ∈ carrier L x  $\sqsubseteq_L$  y
  shows  $\exists \varphi. \text{bij-betw } \varphi x y \wedge \text{strict-mono-on } x \varphi \wedge (\forall e. \varphi e \geq e)$ 
  ⟨proof⟩

definition bij-pmf I F = pmf-of-set {f. bij-betw f I F  $\wedge$  f ∈ extensional I}

lemma card-bijections':
  assumes finite A finite B card A = card B
  shows card {f. bij-betw f A B  $\wedge$  f ∈ extensional A} = fact (card A) (is ?L = ?R)
  ⟨proof⟩

lemma bij-betw-non-empty-finite:
  assumes finite I finite F card I = card F
  shows
    finite {f. bij-betw f I F  $\wedge$  f ∈ extensional I} (is ?T1)
    {f. bij-betw f I F  $\wedge$  f ∈ extensional I} ≠ {} (is ?T2)
  ⟨proof⟩

lemma bij-pmf:
  assumes finite I finite F card I = card F
  shows
    set-pmf (bij-pmf I F) = {f. bij-betw f I F  $\wedge$  f ∈ extensional I}
    finite (set-pmf (bij-pmf I F))
  ⟨proof⟩

```

```

lemma expectation-ge-eval-at-point:
  assumes  $\bigwedge y. y \in \text{set-pmf } p \implies f y \geq (0::\text{real})$ 
  assumes integrable p f
  shows pmf p x * f x  $\leq (\int x. f x \partial p)$  (is ?L  $\leq$  ?R)
  ⟨proof⟩

lemma split-bij-pmf:
  assumes finite I finite F card I = card F J ⊆ I
  shows bij-pmf I F =
    do {
      S ← pmf-of-set {S. card S = card J ∧ S ⊆ F};
      φ ← bij-pmf J S;
      ψ ← bij-pmf (I - J) (F - S);
      return-pmf (merge J (I - J) (φ, ψ))
    } (is ?L = ?R)
  ⟨proof⟩

lemma map-bij-pmf:
  assumes finite I finite F card I = card F inj-on φ F
  shows map-pmf (λf. (λx ∈ I. φ(f x))) (bij-pmf I F) = bij-pmf I (φ ` F)
  ⟨proof⟩

lemma pmf-of-multiset-eq-pmf-of-setI:
  assumes c > 0 x ≠ {#}
  assumes  $\bigwedge i. i \in y \implies \text{count } x i = c$ 
  assumes  $\bigwedge i. i \in \# x \implies i \in y$ 
  shows pmf-of-multiset x = pmf-of-set y
  ⟨proof⟩

lemma card-multi-bij:
  assumes finite J
  assumes I = ∪(A ` J) disjoint-family-on A J
  assumes  $\bigwedge j. j \in J \implies \text{finite } (A j) \wedge \text{finite } (B j) \wedge \text{card } (A j) = \text{card } (B j)$ 
  shows card {f. (∀j ∈ J. bij-betw f (A j) (B j)) ∧ f ∈ extensional I} = (∏i ∈ J.
  fact (card (A i)))
    (is card ?L = ?R)
  ⟨proof⟩

lemma map-bij-pmf-non-inj:
  fixes I :: 'a set
  fixes F :: 'b set
  fixes φ :: 'b ⇒ 'c
  assumes finite I finite F card I = card F
  defines q ≡ {f. f ∈ extensional I ∧ {#f x. x ∈ # mset-set I#} = {#φ x. x ∈ # mset-set F#}}
  shows map-pmf (λf. (λx ∈ I. φ(f x))) (bij-pmf I F) = pmf-of-set q (is ?L = -)
  ⟨proof⟩

```

```

lemmas fkg-inequality-pmf-internalized = fkg-inequality-pmf[unoverload-type 'a]

lemma permutation-distributions-are-neg-associated:
  fixes F :: ('a :: linorder-topology) set
  fixes I :: 'b set
  assumes finite F finite I card I = card F
  shows measure-pmf.neg-assoc (bij-pmf I F) (λi ω. ω i) I
  ⟨proof⟩

lemma multiset-permutation-distributions-are-neg-associated:
  fixes F :: ('a :: linorder-topology) multiset
  fixes I :: 'b set
  assumes finite I card I = size F
  defines p ≡ pmf-of-set {φ. φ ∈ extensional I ∧ image-mset φ (mset-set I) = F}
  shows measure-pmf.neg-assoc p (λi ω. ω i) I
  ⟨proof⟩

lemma n-subsets-prob:
  assumes d ≤ card S finite S s ∈ S
  shows
    measure-pmf.prob (pmf-of-set {a. a ⊆ S ∧ card a = d}) {ω. s ∉ ω} = (1 −
    real d / card S)
    measure-pmf.prob (pmf-of-set {a. a ⊆ S ∧ card a = d}) {ω. s ∈ ω} = real
    d / card S
  ⟨proof⟩

lemma n-subsets-distribution-neg-assoc:
  assumes finite S k ≤ card S
  defines p ≡ pmf-of-set {T. T ⊆ S ∧ card T = k}
  shows measure-pmf.neg-assoc p (∈) S
  ⟨proof⟩

end

```

7 Application: Bloom Filters

The false positive probability of Bloom Filters is a case where negative association is really useful. Traditionally it is derived only approximately. Bloom [4] first derives the expected number of bits set to true given the number of elements inserted, then the false positive probability is computed, pretending that the expected number of bits is the actual number of bits. Both Blooms original derivation and Mitzenmacher and Upfal [15] use this method.

A more correct approach would be to derive a tail bound for the number of set bits and derive a false-positive probability based on that, which unfortunately leads to a complex formula.

An exact result has later been derived using combinatorial methods by Gopinathan and Sergey [10]. However their formula is less useful, as it consists of a sum with Stirling numbers and binomial coefficients.

It is however easy to see that the original bound derived by Bloom is a correct upper bound for the false positive probability using negative association. (This is pointed out by Bao et al. [?].)

In this section, we derive the same bound using this library as an example for the applicability of this library.

```

theory Negative-Association-Bloom-Filters
  imports Negative-Association-Permutation-Distributions
begin

fun bloom-filter-pmf where
  bloom-filter-pmf 0 d N = return-pmf {} |
  bloom-filter-pmf (Suc n) d N = do {
    h ← bloom-filter-pmf n d N;
    a ← pmf-of-set {a. a ⊆ {..<(N::nat)} ∧ card a = d};
    return-pmf (a ∪ h)
  }

lemma bloom-filter-neg-assoc:
  assumes d ≤ N
  shows measure-pmf.neg-assoc (bloom-filter-pmf n d N) (λi ω. i ∈ ω) {..<N}
  ⟨proof⟩

lemma bloom-filter-cell-prob:
  assumes d ≤ N i < N
  shows measure (bloom-filter-pmf n d N) {ω. i ∈ ω} = 1 - (1 - real d / real N) ^ n
  ⟨proof⟩

lemma bloom-filter-false-positive-prob:
  assumes d ≤ N T ⊆ {..<N} card T = d
  shows measure (bloom-filter-pmf n d N) {ω. T ⊆ ω} ≤ (1 - (1 - real d / real
  N) ^ n) ^ d
  (is ?L ≤ ?R)
  ⟨proof⟩

end

```

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