Universal Hash Families

Emin Karayel

March 17, 2025

Abstract

A k-universal hash family is a probability space of functions, which have uniform distribution and form k-wise independent random variables.

They can often be used in place of classic (or cryptographic) hash functions and allow the rigorous analysis of the performance of randomized algorithms and data structures that rely on hash functions.

In 1981 Wegman and Carter [4] introduced a generic construction for such families with arbitrary k using polynomials over a finite field. This entry contains a formalization of them and establishes the property of k-universality.

To be useful the formalization also provides an explicit construction of finite fields using the factor ring of integers modulo a prime. Additionally, some generic results about independent families are shown that might be of independent interest.

1 Introduction and Definition

theory Universal-Hash-Families

imports *HOL*-*Probability.Independent-Family* **begin**

Universal hash families are commonly used in randomized algorithms and data structures to randomize the input of algorithms, such that probabilistic methods can be employed without requiring any assumptions about the input distribution.

If we regard a family of hash functions from a domain D to a finite range R as a uniform probability space, then the family is k-universal if:

- For each $x \in D$ the evaluation of the functions at x forms a uniformly distributed random variable on R.
- The evaluation random variables for k or fewer distinct domain elements form an independent family of random variables.

This definition closely follows the definition from Vadhan [3, §3.5.5], with the minor modification that independence is required not only for exactly k, but also for *fewer* than k distinct domain elements. The correction is due to the fact that in the corner case where D has fewer than k elements, the second part of their definition becomes void. In the formalization this helps avoid an unnecessary assumption in the theorems.

The following definition introduces the notion of k-wise independent random variables:

definition (in prob-space) k-wise-indep-vars where k-wise-indep-vars k M' X I = $(\forall J \subseteq I. \ card \ J \leq k \longrightarrow finite \ J \longrightarrow indep-vars \ M' \ X \ J)$ lemma (in prob-space) k-wise-indep-vars-subset: assumes k-wise-indep-vars $k M' \ X \ I$ assumes finite Jassumes card $J \leq k$ shows indep-vars $M' \ X \ J$ $\langle proof \rangle$ lemma (in prob-space) k-wise-indep-subset: assumes k-wise-indep-vars $k M' \ X' \ I$ shows k-wise-indep-vars $k M' \ X' \ J$ $\langle proof \rangle$

Similarly for a finite non-empty set A the predicate *uniform-on* XA indicates that the random variable is uniformly distributed on A:

definition (in prob-space) uniform-on X A = (distr M (count-space UNIV) X = uniform-measure (count-space UNIV) $A \land A \neq \{\} \land$ finite $A \land$ random-variable (count-space UNIV) X)

```
lemma (in prob-space) uniform-onD:

assumes uniform-on X A

shows prob {\omega \in space \ M. \ X \ \omega \in B} = card (A \cap B) / card A

{proof}
```

With the two previous definitions it is possible to define the k-universality condition for a family of hash functions from D to R:

definition (in prob-space) k-universal $k \ X \ D \ R = ($ k-wise-indep-vars $k \ (\lambda$ -. count-space UNIV) $X \ D \land$ $(\forall i \in D. uniform-on \ (X \ i) \ R))$

Note: The definition is slightly more generic then the informal specification from above. This is because usually a family is formed by a single function with a variable seed parameter. Instead of choosing a random function from a probability space, a random seed is chosen from the probability space which parameterizes the hash function.

The following section contains some preliminary results about independent families of random variables. Section 3 introduces the Carter-Wegman hash family, which is an explicit construction of k-universal families for arbitrary k using polynomials over finite fields. The last section contains a proof that the factor ring of the integers modulo a prime ideal is a finite field, followed by an isomorphic construction of prime fields over an initial segment of the natural numbers.

end

2 Preliminary Results

```
theory Universal-Hash-Families-More-Independent-Families
imports
Universal-Hash-Families
HOL-Probability.Stream-Space
HOL-Probability.Probability-Mass-Function
begin
lemma set-comp-image-cong:
```

assumes $\bigwedge x. P x \Longrightarrow f x = h (g x)$ shows $\{f x \mid x. P x\} = h ` \{g x \mid x. P x\}$ $\langle proof \rangle$

lemma (in prob-space) k-wise-indep-vars-compose: assumes k-wise-indep-vars k M' X I assumes $\bigwedge i. i \in I \implies Y i \in measurable (M' i) (N i)$ shows k-wise-indep-vars k N ($\lambda i x. Y i (X i x)$) I $\langle proof \rangle$

```
lemma (in prob-space) k-wise-indep-vars-triv:
assumes indep-vars N T I
shows k-wise-indep-vars k N T I
\langle proof \rangle
```

The following two lemmas are of independent interest, they help infer independence of events and random variables on distributions. (Candidates for *HOL–Probability.Independent-Family*).

lemma (in prob-space) indep-sets-distr: fixes A assumes random-variable N f defines $F \equiv (\lambda i. (\lambda a. f - `a \cap space M) `A i)$ assumes indep-F: indep-sets F I assumes sets-A: $\bigwedge i. i \in I \Longrightarrow A i \subseteq sets N$ shows prob-space.indep-sets (distr M N f) A I $\langle proof \rangle$ **lemma** range-inter: range $((\cap) F) = Pow F$ $\langle proof \rangle$

The singletons and the empty set form an intersection stable generator of a countable discrete σ -algebra:

```
lemma sigma-sets-singletons-and-empty:

assumes countable M

shows sigma-sets M (insert {} ((\lambda k. {k}) 'M)) = Pow M

\langle proof \rangle
```

In some of the following theorems, the premise M = measure-pmf p is used. This allows stating theorems that hold for pmfs more concisely, for example, instead of measure-pmf.prob $p \ A \leq measure-pmf.prob \ p \ B$ we can just write $M = measure-pmf \ p \Longrightarrow prob \ A \leq prob \ B$ in the locale prob-space.

```
lemma prob-space-restrict-space:

assumes [simp]:M = measure-pmf p

shows prob-space (restrict-space M (set-pmf p))

\langle proof \rangle
```

The abbreviation below is used to specify the discrete σ -algebra on UNIV as a measure space. It is used in places where the existing definitions, such as *indep-vars*, expect a measure space even though only a *measurable* space is really needed, i.e., in cases where the property is invariant with respect to the actual measure.

```
hide-const (open) discrete
```

```
abbreviation discrete \equiv count-space UNIV
```

```
lemma (in prob-space) indep-vars-restrict-space:

assumes [simp]:M = measure-pmf p

assumes

prob-space.indep-vars (restrict-space M (set-pmf p)) (\lambda-. discrete) X I

shows indep-vars (\lambda-. discrete) X I

\langle proof \rangle

lemma (in prob-space) measure-pmf-eq:

assumes M = measure-pmf p

assumes \Lambda x. x \in set-pmf p \Longrightarrow (x \in P) = (x \in Q)

shows prob P = prob Q

\langle proof \rangle
```

The following lemma is an intro rule for the independence of random variables defined on pmfs. In that case it is possible, to check the independence of random variables point-wise.

The proof relies on the fact that the support of a pmf is countable and the σ -algebra of such a set can be generated by singletons.

lemma (in prob-space) indep-vars-pmf: **assumes** [simp]:M = measure-pmf p **assumes** $\bigwedge a \ J. \ J \subseteq I \implies finite \ J \implies$ prob $\{\omega. \ \forall i \in J. \ X \ i \ \omega = a \ i\} = (\prod i \in J. \ prob \ \{\omega. \ X \ i \ \omega = a \ i\})$ **shows** indep-vars (λ -. discrete) $X \ I$ $\langle proof \rangle$

```
lemma (in prob-space) split-indep-events:

assumes M = measure-pmf p

assumes indep-vars (\lambda i. discrete) X' I

assumes K \subseteq I finite K

shows prob {\omega. \forall x \in K. P x (X' x \omega)} = (\prod x \in K. prob {\omega. P x (X' x \omega)})

\langle proof \rangle
```

```
lemma pmf-of-set-eq-uniform:

assumes finite A \ A \neq \{\}

shows measure-pmf (pmf-of-set A) = uniform-measure discrete A \ \langle proof \rangle
```

```
lemma (in prob-space) uniform-onI:

assumes M = measure-pmf p

assumes finite A \ A \neq \{\}

assumes \bigwedge a. prob \{\omega. X \ \omega = a\} = indicator A \ a \ / card A

shows uniform-on X A

\langle proof \rangle
```

 \mathbf{end}

3 Carter-Wegman Hash Family

theory Carter-Wegman-Hash-Family

imports

Interpolation-Polynomials-HOL-Algebra.Interpolation-Polynomial-Cardinalities Universal-Hash-Families-More-Independent-Families

begin

The Carter-Wegman hash family is a generic method to obtain k-universal hash families for arbitrary k. (There are faster solutions, such as tabulation hashing, which are limited to a specific k. See for example [2].)

The construction was described by Wegman and Carter [4], it is a hash family between the elements of a finite field and works by choosing randomly a polynomial over the field with degree less than k. The hash function is

the evaluation of a such a polynomial.

Using the property that the fraction of polynomials interpolating a given set of $s \leq k$ points is 1 / real (card (carrier R))^s, which is shown in [1], it is possible to obtain both that the hash functions are k-wise independent and uniformly distributed.

In the following two locales are introduced, the main reason for both is to make the statements of the theorems and proofs more concise. The first locale *poly-hash-family* fixes a finite ring R and the probability space of the polynomials of degree less than k. Because the ring is not a field, the family is not yet k-universal, but it is still possible to state a few results such as the fact that the range of the hash function is a subset of the carrier of the ring.

The second locale *carter-wegman-hash-family* is an extension of the former with the assumption that R is a field with which the k-universality follows. The reason for using two separate locales is to support use cases, where the ring is only probably a field. For example if it is the set of integers modulo an approximate prime, in such a situation a subset of the properties of an algorithm using approximate primes would need to be verified even if R is only a ring.

definition (in ring) hash $x \omega = eval \omega x$

```
locale poly-hash-family = ring +
fixes k :: nat
assumes finite-carrier[simp]: finite (carrier R)
assumes k-ge-0: k > 0
begin
```

definition space where space = bounded-degree-polynomials R kdefinition M where M = measure-pmf (pmf-of-set space)

lemma finite-space[simp]:finite space $\langle proof \rangle$

lemma non-empty-bounded-degree-polynomials[simp]:space \neq {} $\langle proof \rangle$

This is to add *carrier-not-empty* to the simp set in the context of *poly-hash-family*:

```
lemma non-empty-carrier[simp]: carrier R \neq \{\} \langle proof \rangle
```

sublocale prob-space M $\langle proof \rangle$

lemma hash-range[simp]: **assumes** $\omega \in space$ **assumes** $x \in carrier R$

```
shows hash x \omega \in carrier R
  \langle proof \rangle
lemma hash-range-2:
  assumes \omega \in space
 shows (\lambda x. hash \ x \ \omega) ' carrier R \subseteq carrier R
  \langle proof \rangle
lemma integrable-M[simp]:
  fixes f :: 'a \ list \Rightarrow 'c::\{banach, second-countable-topology\}
 shows integrable M f
    \langle proof \rangle
end
locale carter-wegman-hash-family = poly-hash-family +
 assumes field-R: field R
begin
sublocale field
  \langle proof \rangle
abbreviation field-size \equiv card (carrier R)
lemma poly-cards:
  assumes K \subseteq carrier R
 assumes card K \leq k
 assumes y \in K \subseteq (carrier R)
  shows
    card {\omega \in \text{space.} (\forall k \in K. \text{ eval } \omega \ k = y \ k)} = field-size \widehat{(k-\text{card } K)}
  \langle proof \rangle
lemma poly-cards-single:
 assumes x \in carrier R
 assumes y \in carrier R
 shows card \{\omega \in space. eval \ \omega \ x = y\} = field-size(k-1)
  \langle proof \rangle
lemma hash-prob:
  assumes K \subseteq carrier R
  assumes card K \leq k
 assumes y ' K \subseteq carrier R
 shows
    prob {\omega. (\forall x \in K. hash x \omega = y x)} = 1/(real field-size)^card K
\langle proof \rangle
lemma prob-single:
 assumes x \in carrier R \ y \in carrier R
 shows prob {\omega. hash x \omega = y} = 1/(real field-size)
  \langle proof \rangle
```

```
lemma prob-range:

assumes [simp]:x \in carrier R

shows prob {\omega. hash x \ \omega \in A} = card (A \cap carrier R) / field-size

\langle proof \rangle
```

```
lemma indep:

assumes J \subseteq carrier R

assumes card J \leq k

shows indep-vars (\lambda-. discrete) hash J

\langle proof \rangle
```

```
lemma k-wise-indep:
k-wise-indep-vars k (\lambda-. discrete) hash (carrier R)
\langle proof \rangle
```

```
lemma inj-if-degree-1:

assumes \omega \in space

assumes degree \omega = 1

shows inj-on (\lambda x. hash \ x \ \omega) (carrier R)

\langle proof \rangle
```

```
lemma uniform:

assumes i \in carrier R

shows uniform-on (hash i) (carrier R)

\langle proof \rangle
```

This the main result of this section - the Carter-Wegman hash family is k-universal.

```
theorem k-universal:
k-universal k hash (carrier R) (carrier R)
\langle proof \rangle
```

\mathbf{end}

```
lemma poly-hash-familyI:
assumes ring R
assumes finite (carrier R)
assumes 0 < k
shows poly-hash-family R k
\langle proof \rangle
```

```
lemma carter-wegman-hash-familyI:

assumes field F

assumes finite (carrier F)

assumes 0 < k

shows carter-wegman-hash-family F k

\langle proof \rangle
```

```
lemma hash-k-wise-indep:
 assumes field F \wedge finite (carrier F)
 assumes 1 \leq n
 shows
   prob-space.k-wise-indep-vars (pmf-of-set (bounded-degree-polynomials F n)) n
   (\lambda-. pmf-of-set (carrier F)) (ring.hash F) (carrier F)
\langle proof \rangle
lemma hash-prob-single:
 assumes field F \wedge finite (carrier F)
 assumes x \in carrier F
 assumes 1 \leq n
 assumes y \in carrier F
 shows
   \mathcal{P}(\omega \text{ in pmf-of-set (bounded-degree-polynomials } F n). ring.hash F x \omega = y)
     = 1/(real (card (carrier F)))
\langle proof \rangle
```

 \mathbf{end}

4 Indexed Products of Probability Mass Functions

theory Universal-Hash-Families-More-Product-PMF imports Concentration-Inequalities.Concentration-Inequalities-Preliminary Finite-Fields.Finite-Fields-More-Bijections Universal-Hash-Families-More-Independent-Families

begin

hide-const (open) Isolated.discrete

This section introduces a restricted version of Pi-pmf where the default value is *undefined* and contains some additional results about that case in addition to HOL-Probability.Product-PMF

abbreviation prod-pmf where prod-pmf I $M \equiv Pi$ -pmf I undefined M

lemma *measure-pmf-cong*:

assumes $\bigwedge x. \ x \in set\text{-pmf } p \implies x \in P \iff x \in Q$ **shows** measure (measure-pmf p) $P = measure (measure-pmf p) \ Q$ $\langle proof \rangle$

lemma pmf-mono:

assumes $\bigwedge x. \ x \in set\text{-pmf } p \implies x \in P \implies x \in Q$ **shows** measure (measure-pmf p) $P \leq measure$ (measure-pmf p) Q $\langle proof \rangle$

lemma pmf-add: **assumes** $\bigwedge x. \ x \in P \implies x \in set\text{-pmf } p \implies x \in Q \lor x \in R$ shows measure $p \ P \le measure \ p \ Q + measure \ p \ R$ $\langle proof \rangle$ lemma pmf-prod-pmf: assumes finite Ishows $pmf \ (prod-pmf \ I \ M) \ x = (if \ x \in extensional \ I \ then \ \prod i \in I. \ (pmf \ (M \ i)))$ $(x \ i) \ else \ 0)$ $\langle proof \rangle$ lemma PiE-defaut-undefined-eq: PiE- $dflt \ I \ undefined \ M = PiE \ I \ M$ $\langle proof \rangle$ lemma set-prod-pmf: assumes finite Ishows set- $pmf \ (prod-pmf \ I \ M) = PiE \ I \ (set-pmf \ \circ \ M)$ $\langle proof \rangle$

A more general version of *measure-Pi-pmf-Pi*.

lemma prob-prod-pmf': **assumes** finite I **assumes** $J \subseteq I$ **shows** measure (measure-pmf (Pi-pmf I d M)) (Pi J A) = ($\prod i \in J$. measure (M i) (A i)) $\langle proof \rangle$

lemma prob-prod-pmf-slice: **assumes** finite I **assumes** $i \in I$ **shows** measure (measure-pmf (prod-pmf I M)) { ω . P (ω i)} = measure (M i) { ω . P ω } $\langle proof \rangle$

definition restrict-dfl where restrict-dfl $f A d = (\lambda x. if x \in A then f x else d)$

lemma pi-pmf-decompose: assumes finite I shows Pi-pmf I d $M = map-pmf \ (\lambda \omega. restrict-dfl \ (\lambda i. \omega \ (f \ i) \ i) \ I \ d) \ (Pi-pmf \ (f \ `I) \ (\lambda -. \ d) \ (\lambda j. \ Pi-pmf \ (f \ -` \{j\} \cap I) \ d \ M))$ $\langle proof \rangle$

lemma restrict-dfl-iter: restrict-dfl (restrict-dfl f I d) $J d = restrict-dfl f (I \cap J) d$

 $\langle proof \rangle$

lemma indep-vars-restrict': **assumes** finite I **shows** prob-space.indep-vars (Pi-pmf I d M) (λ -. discrete) ($\lambda i \omega$. restrict-dfl ω ($f - \{i\} \cap I$) d) ($f \cdot I$) (proof) **lemma** indep-vars-restrict-intro': **assumes** finite I **assumes** $\bigwedge i \ \omega. \ i \in J \implies X' \ i \ \omega = X' \ i \ (restrict-dfl \ \omega \ (f - `\{i\} \cap I) \ d)$ **assumes** $J = f \ `I$ **assumes** $\bigwedge \omega \ i. \ i \in J \implies X' \ i \ \omega \in space \ (M' \ i)$ **shows** prob-space.indep-vars (measure-pmf (Pi-pmf I \ d \ p)) M' (\lambda i \ \omega. \ X' \ i \ \omega) J $\langle proof \rangle$

lemma

fixes $f :: 'b \Rightarrow ('c :: \{second-countable-topology, banach, real-normed-field\})$ assumes finite I assumes $i \in I$ assumes integrable (measure-pmf (M i)) f shows integrable-Pi-pmf-slice: integrable (Pi-pmf I d M) ($\lambda x. f(x i)$) and expectation-Pi-pmf-slice: integral^L (Pi-pmf I d M) ($\lambda x. f(x i)$) = integral^L (M i) f (proof)

This is an improved version of *expectation-prod-Pi-pmf*. It works for general normed fields instead of non-negative real functions .

lemma expectation-prod-Pi-pmf: **fixes** $f :: 'a \Rightarrow 'b \Rightarrow ('c :: \{second-countable-topology, banach, real-normed-field\})$ **assumes** finite I **assumes** $\bigwedge i. i \in I \implies integrable (measure-pmf (M i)) (f i)$ **shows** $integral^{L} (Pi-pmf I d M) (\lambda x. (\prod i \in I. f i (x i))) = (\prod i \in I. integral^{L} (M i) (f i))$ $\langle proof \rangle$

lemma variance-prod-pmf-slice:

fixes $f :: 'a \Rightarrow real$ assumes $i \in I$ finite Iassumes integrable (measure-pmf (M i)) ($\lambda \omega$. $f \omega^2$) shows prob-space.variance (Pi-pmf I d M) ($\lambda \omega$. $f (\omega i)$) = prob-space.variance (M i) f(proof)

lemma Pi-pmf-bind-return: **assumes** finite I **shows** Pi-pmf I d (λi . M i \gg = (λx . return-pmf (f i x))) = Pi-pmf I d' M \gg = (λx . return-pmf (λi . if $i \in I$ then f i (x i) else d)) (proof)

 $\begin{array}{l} \textbf{lemma } pmf\text{-}of\text{-}set\text{-}prod\text{-}eq\text{:}}\\ \textbf{assumes } A \neq \{\} \ finite \ A\\ \textbf{assumes } B \neq \{\} \ finite \ B\\ \textbf{shows } \ pmf\text{-}of\text{-}set \ (A \times B) = pair\text{-}pmf \ (pmf\text{-}of\text{-}set \ A) \ (pmf\text{-}of\text{-}set \ B)\\ \langle proof \rangle \end{array}$

```
\begin{array}{l} \textbf{lemma split-pmf-mod-div':}\\ \textbf{assumes } a > (0::nat)\\ \textbf{assumes } b > 0\\ \textbf{shows } map-pmf \ (\lambda x. \ (x \ mod \ a, \ x \ div \ a)) \ (pmf-of-set \ \{..<a*b\}) = pmf-of-set\\ (\{..<a\} \times \{..<b\})\\ \langle proof \rangle\\ \end{array}\begin{array}{l} \textbf{lemma \ split-pmf-mod-div:}\\ \textbf{assumes } a > (0::nat)\\ \textbf{assumes } b > 0\\ \textbf{shows } map-pmf \ (\lambda x. \ (x \ mod \ a, \ x \ div \ a)) \ (pmf-of-set \ \{..<a*b\}) = \\ pair-pmf \ (pmf-of-set \ \{..<a*b\}) = \\ pair-pmf \ (pmf-of-set \ \{..<a*b\}) = \\ proof \rangle \end{array}
```

 \mathbf{end}

5 Pseudorandom Objects

theory Pseudorandom-Objects imports Universal-Hash-Families-More-Product-PMF begin

This section introduces a combinator library for pseudorandom objects [3]. These can be thought of as PRNGs but with rigorous mathematical properties, which can be used to in algorithms to reduce their randomness usage.

Such an object represents a non-empty multiset, with an efficient mechanism to sample from it. They have a natural interpretation as a probability space (each element is selected with a probability proportional to its occurrence count in the multiset).

The following section will introduce a construction of k-independent hash families as a pseudorandom object. The AFP entry Expander_Graphs then follows up with expander walks as pseudorandom objects.

record 'a pseudorandom-object = pro-last :: nat pro-select :: nat \Rightarrow 'a

definition pro-size where pro-size S = pro-last S + 1definition sample-pro where sample-pro S = map-pmf (pro-select S) (pmf-of-set $\{0..pro-last S\}$)

declare [[coercion sample-pro]]

abbreviation pro-set where pro-set $S \equiv set\text{-pmf}$ (sample-pro S)

lemma sample-pro-alt: sample-pro S = map-pmf (pro-select S) (pmf-of-set {..<pro-size S})

 $\langle proof \rangle$

lemma pro-size-gt-0: pro-size S > 0 $\langle proof \rangle$ **lemma** set-sample-pro: pro-set $S = \text{pro-select } S \text{ '} \{..<\text{pro-size } S\}$ $\langle proof \rangle$ **lemma** *set-pmf-of-set-sample-size*[*simp*]: set-pmf $(pmf-of-set \{..< pro-size S\}) = \{..< pro-size S\}$ $\langle proof \rangle$ **lemma** pro-select-in-set: pro-select S (x mod pro-size S) \in pro-set S $\langle proof \rangle$ **lemma** finite-pro-set: finite (pro-set S) $\langle proof \rangle$ **lemma** *integrable-sample-pro*[*simp*]: fixes $f :: 'a \Rightarrow 'c::{banach, second-countable-topology}$ **shows** integrable (measure-pmf (sample-pro S)) f $\langle proof \rangle$ definition *list-pro* :: 'a *list* \Rightarrow 'a *pseudorandom-object* where $list-pro \ ls = (| \ pro-last = length \ ls - 1, \ pro-select = (!) \ ls |)$ lemma *list-pro*: assumes $xs \neq []$ shows sample-pro (list-pro xs) = pmf-of-multiset (mset xs) (is ?L = ?R) $\langle proof \rangle$ lemma *list-pro-2*: assumes $xs \neq []$ distinct xsshows sample-pro (list-pro xs) = pmf-of-set (set xs) (is ?L = ?R) $\langle proof \rangle$ lemma *list-pro-size*: assumes $xs \neq []$ **shows** pro-size (list-pro xs) = length xs $\langle proof \rangle$ lemma *list-pro-set*: assumes $xs \neq []$ **shows** pro-set (list-pro xs) = set xs

 $\langle proof \rangle$

definition *nat-pro* :: *nat* \Rightarrow *nat pseudorandom-object* **where** nat-pro n = (pro-last = n-1, pro-select = id)lemma *nat-pro-size*: assumes $n > \theta$ **shows** pro-size $(nat-pro \ n) = n$ $\langle proof \rangle$ **lemma** *nat-pro*: assumes $n > \theta$ **shows** sample-pro (nat-pro n) = pmf-of-set $\{.. < n\}$ $\langle proof \rangle$ **lemma** *nat-pro-set*: assumes $n > \theta$ shows pro-set $(nat-pro \ n) = \{.. < n\}$ $\langle proof \rangle$ **fun** count-zeros :: $nat \Rightarrow nat \Rightarrow nat$ where count-zeros $0 \ k = 0$ count-zeros (Suc n) $k = (if odd \ k \ then \ 0 \ else \ 1 + count-zeros \ n \ (k \ div \ 2))$ **lemma** count-zeros-iff: $j \leq n \Longrightarrow$ count-zeros $n \ k \geq j \longleftrightarrow 2 \ j \ dvd \ k$ $\langle proof \rangle$ lemma count-zeros-max: count-zeros n k \leq n $\langle proof \rangle$ **definition** geom-pro :: $nat \Rightarrow nat pseudorandom-object$ where geom-pro $n = (| pro-last = 2\hat{n} - 1, pro-select = count-zeros n |)$ **lemma** geom-pro-size: pro-size (geom-pro n) = $2\hat{n}$ $\langle proof \rangle$ **lemma** geom-pro-range: pro-set (geom-pro n) \subseteq {...n} $\langle proof \rangle$ **lemma** geom-pro-prob: measure (sample-pro (geom-pro n)) { ω . $\omega \ge j$ } = of-bool ($j \le n$) / 2^j (is ?L = (R) $\langle proof \rangle$ **lemma** geom-pro-prob-single: measure (sample-pro (geom-pro n)) $\{j\} \leq 1 / 2\hat{j}$ (is $?L \leq ?R$)

 $\langle proof \rangle$

definition prod-pro :: 'a pseudorandom-object \Rightarrow 'b pseudorandom-object \Rightarrow ('a \times 'b) pseudorandom-object where prod-pro P Q = () pro-last = pro-size P * pro-size Q - 1, pro-select = (λk . (pro-select P (k mod pro-size P), pro-select Q (k div pro-size P)))))

lemma prod-pro-size: pro-size (prod-pro P Q) = pro-size P * pro-size Q $\langle proof \rangle$

lemma *prod-pro*:

sample-pro (prod-pro P Q) = pair-pmf (sample-pro P) (sample-pro Q) (is ?L = ?R)(proof)

lemma prod-pro-set: pro-set (prod-pro P Q) = pro-set $P \times \text{pro-set } Q$ $\langle \text{proof} \rangle$

 \mathbf{end}

6 K-Independent Hash Families as Pseudorandom Objects

theory Pseudorandom-Objects-Hash-Families imports Pseudorandom-Objects Finite-Fields.Find-Irreducible-Poly Carter-Wegman-Hash-Family Universal-Hash-Families-More-Product-PMF begin

hide-const (open) Numeral-Type.mod-ring hide-const (open) Divisibility.prime hide-const (open) Isolated.discrete

definition hash-space' ::

('a, 'b) idx-ring-enum-scheme \Rightarrow nat \Rightarrow ('c, 'd) pseudorandom-object-scheme \Rightarrow (nat \Rightarrow 'c) pseudorandom-object where hash-space' R k S = ((pro-last = idx-size R $^{\sim}k-1$, pro-select = (λx i. pro-select S

 $(idx-enum-inv \ R \ (poly-eval \ R \ (poly-enum \ R \ k \ x) \ (idx-enum \ R \ i)) \ mod \ pro-size$ S)))) **lemma** hash-prob-single': assumes field F finite (carrier F) **assumes** $x \in carrier F$ assumes $1 \leq n$ **shows** measure (pmf-of-set (bounded-degree-polynomials F(n)) { ω . ring.hash F(x) $\omega = y\} =$ of-bool $(y \in carrier F)/(real (card (carrier F)))$ (is ?L = ?R) $\langle proof \rangle$ **lemma** hash-k-wise-indep': **assumes** field $F \wedge finite$ (carrier F) assumes 1 < n**shows** prob-space.k-wise-indep-vars (pmf-of-set (bounded-degree-polynomials F n)) n $(\lambda$ -. discrete) (ring.hash F) (carrier F) $\langle proof \rangle$ **lemma** hash-space': fixes R :: ('a, 'b) idx-ring-enum-scheme assumes $enum_C R$ field_C R assumes pro-size $S \, dvd \, order \, (ring-of R)$ assumes $I \subseteq \{..<order (ring-of R)\}$ card $I \leq k$ shows map-pmf (λf . ($\lambda i \in I$. f i)) (sample-pro (hash-space' R k S)) = prod-pmf I $(\lambda$ -. sample-pro S) (is ?L = ?R) $\langle proof \rangle$ **lemma** hash-space'-range: pro-select (hash-space' $R \ k \ S$) $i \ j \in pro-set \ S$ $\langle proof \rangle$ definition hash-pro :: $nat \Rightarrow nat \Rightarrow ('a, 'b)$ pseudorandom-object-scheme $\Rightarrow (nat \Rightarrow 'a)$ pseudorandom-object where hash-pro k d S = (let (p,j) = split-power (pro-size S);l = max j (floorlog p (d-1))in hash-space' (GF $(p\hat{l})$) k S) definition hash-pro-spmf :: $nat \Rightarrow nat \Rightarrow ('a, 'b) \ pseudorandom-object-scheme \Rightarrow (nat \Rightarrow 'a) \ pseudoran-object-scheme \Rightarrow (na$ dom-object spmf where hash-pro-spmf k d S = $do \{$ let (p,j) = split-power (pro-size S);

 $let \ l = max \ j \ (floorlog \ p \ (d-1));$ $R \leftarrow GF_R \ (p\ l);$ $return-spmf \ (hash-space' \ R \ k \ S)$ }

 $\begin{array}{l} \textbf{definition } hash-pro-pmf ::\\ nat \Rightarrow nat \Rightarrow ('a,'b) \ pseudorandom-object-scheme \Rightarrow (nat \Rightarrow 'a) \ pseudorandom-object pmf\\ \textbf{where } hash-pro-pmf \ k \ d \ S = map-pmf \ the \ (hash-pro-spmf \ k \ d \ S) \\ \textbf{syntax}\\ -FLIPBIND \qquad :: ('a \Rightarrow 'b) \Rightarrow 'c \Rightarrow 'b \ (\textbf{infixr} <=<<>54) \end{array}$

syntax-consts -FLIPBIND == Monad-Syntax.bind

translations

-FLIPBIND $f g \implies f$

context

fixes S
fixes d :: nat
fixes k :: nat
assumes size-prime-power: is-prime-power (pro-size S)
begin

private definition p where p = fst (split-power (pro-size S)) private definition j where j = snd (split-power (pro-size S)) private definition l where l = max j (floorlog p (d-1))

private lemma split-power: (p,j) = split-power (pro-size S) $\langle proof \rangle$ **lemma** hash-sample-space-alt: hash-pro k d S = hash-space' (GF (p^{γ})) k S $\langle proof \rangle$ **lemma** p-prime : prime p and j-gt-0: j > 0 $\langle proof \rangle$ **lemma** l-gt-0: l > 0 $\langle proof \rangle$ **lemma** prime-power: is-prime-power (p^{γ}) $\langle proof \rangle$

lemma hash-in-hash-pro-spmf: hash-pro k d $S \in$ set-spmf (hash-pro-spmf k d S) $\langle proof \rangle$

lemma lossless-hash-pro-spmf: lossless-spmf (hash-pro-spmf k d S) $\langle proof \rangle$

lemma hashp-eq-hash-pro-spmf: set-pmf (hash-pro-pmf k d S) = set-spmf (hash-pro-spmf k d S)

 $\langle proof \rangle$

lemma hashp-in-hash-pro-spmf:

assumes $x \in set\text{-}pmf$ (hash-pro-pmf k d S) shows $x \in set\text{-}spmf$ (hash-pro-spmf k d S) $\langle proof \rangle$

lemma hash-pro-in-hash-pro-pmf: hash-pro k d $S \in$ set-pmf (hash-pro-pmf k d S) $\langle proof \rangle$

lemma hash-pro-spmf-distr: **assumes** $s \in set$ -spmf (hash-pro-spmf k d S) **assumes** $I \subseteq \{..<d\}$ card $I \leq k$ **shows** map-pmf ($\lambda f.$ ($\lambda i \in I.$ f i)) (sample-pro s) = prod-pmf I (λ -. sample-pro S) $\langle proof \rangle$

lemma hash-pro-spmf-component: **assumes** $s \in set$ -spmf (hash-pro-spmf k d S) **assumes** i < d k > 0 **shows** map-pmf ($\lambda f. f i$) (sample-pro s) = sample-pro S (**is** ?L = ?R) $\langle proof \rangle$

lemma hash-pro-spmf-indep: **assumes** $s \in set$ -spmf (hash-pro-spmf k d S) **assumes** $I \subseteq \{..< d\}$ card $I \leq k$ **shows** prob-space.indep-vars (sample-pro s) (λ -. discrete) ($\lambda i \ \omega. \ \omega i$) I $\langle proof \rangle$

```
lemma hash-pro-spmf-k-indep:

assumes s \in set-spmf (hash-pro-spmf k d S)

shows prob-space.k-wise-indep-vars (sample-pro s) k (\lambda-. discrete) (\lambda i \ \omega. \ \omega i)

{..<d}

\langle proof \rangle lemma hash-pro-spmf-size-aux:

assumes s \in set-spmf (hash-pro-spmf k d S)

shows pro-size s = (p \ l) \ k (is ?L = ?R)

\langle proof \rangle
```

lemma floorlog-alt-def: floorlog b $a = (if \ 1 < b \ then \ nat \ \lceil log \ (real \ b) \ (real \ a+1) \rceil \ else \ 0) \langle proof \rangle$

lemma hash-pro-spmf-size: assumes $s \in set$ -spmf (hash-pro-spmf k d S) assumes (p',j') = split-power (pro-size S) shows pro-size $s = (p' \widehat{(max j' (floorlog p' (d-1)))})\widehat{k}$ $\langle proof \rangle$

lemma hash-pro-spmf-size': **assumes** $s \in set$ -spmf (hash-pro-spmf k d S) d > 0 **assumes** (p',j') = split-power (pro-size S)**shows** pro-size $s = (p' \cap (k*max j' (nat \lceil log p' d \rceil)))$

$\langle proof \rangle$

```
lemma hash-pro-spmf-size-prime-power:

assumes s \in set-spmf (hash-pro-spmf k d S)

assumes k > 0

shows is-prime-power (pro-size s)

\langle proof \rangle
```

```
lemma hash-pro-smpf-range:

assumes s \in set-spmf (hash-pro-spmf k d S)

shows pro-select s i q \in pro-set S

\langle proof \rangle
```

```
lemmas hash-pro-size' = hash-pro-spmf-size'[OF hash-in-hash-pro-spmf]
lemmas hash-pro-size = hash-pro-spmf-size[OF hash-in-hash-pro-spmf]
lemmas hash-pro-size-prime-power = hash-pro-spmf-size-prime-power[OF hash-in-hash-pro-spmf]
lemmas hash-pro-distr = hash-pro-spmf-distr[OF hash-in-hash-pro-spmf]
lemmas hash-pro-component = hash-pro-spmf-component[OF hash-in-hash-pro-spmf]
lemmas hash-pro-indep = hash-pro-spmf-indep[OF hash-in-hash-pro-spmf]
lemmas hash-pro-k-indep = hash-pro-spmf-k-indep[OF hash-in-hash-pro-spmf]
lemmas hash-pro-range = hash-pro-spmf-range[OF hash-in-hash-pro-spmf]
```

```
lemmas hash-pro-pmf-size' = hash-pro-spmf-size'[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-size = hash-pro-spmf-size[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-size-prime-power = hash-pro-spmf-size-prime-power[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-distr = hash-pro-spmf-distr[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-component = hash-pro-spmf-component[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-indep = hash-pro-spmf-indep[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-k-indep = hash-pro-spmf-k-indep[OF hashp-in-hash-pro-spmf]
lemmas hash-pro-pmf-k-indep = hash-pro-spmf-k-indep[OF hashp-in-hash-pro-spmf]
```

 \mathbf{end}

```
open-bundle pseudorandom-object-syntax
begin
notation hash-pro (\langle \mathcal{H} \rangle)
notation hash-pro-spmf (\langle \mathcal{H}_S \rangle)
notation hash-pro-pmf (\langle \mathcal{H}_P \rangle)
notation list-pro (\langle \mathcal{L} \rangle)
notation nat-pro (\langle \mathcal{L} \rangle)
notation geom-pro (\langle \mathcal{G} \rangle)
notation prod-pro (infixr (\times_P \rangle 65)
end
```

 \mathbf{end}

References

- E. Karayel. Interpolation polynomials (in hol-algebra). Archive of Formal Proofs, Jan. 2022. https://isa-afp.org/entries/Interpolation_ Polynomials_HOL_Algebra.html, Formal proof development.
- [2] M. Thorup and Y. Zhang. Tabulation based 5-universal hashing and linear probing. In *Proceedings of the Meeting on Algorithm Engineering* & *Experiments*, ALENEX '10, pages 62–76, USA, 2010. Society for Industrial and Applied Mathematics.
- [3] S. P. Vadhan. Pseudorandomness. Foundations and Trends Din Theoretical Computer Science, 7(1-3):1–336, 2012.
- [4] M. N. Wegman and J. L. Carter. New hash functions and their use in authentication and set equality. *Journal of Computer and System Sciences*, 22(3):265–279, 1981.