# Formalization of Randomized Approximation Algorithms for Frequency Moments

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#### March 17, 2025

#### Abstract

In 1999 Alon et. al. introduced the still active research topic of approximating the frequency moments of a data stream using randomized algorithms with minimal space usage. This includes the problem of estimating the cardinality of the stream elements—the zeroth frequency moment. But, also higher-order frequency moments that provide information about the skew of the data stream. (The k-th frequency moment of a data stream is the sum of the k-th powers of the occurrence counts of each element in the stream.) This entry formalizes three randomized algorithms for the approximation of  $F_0$ ,  $F_2$  and  $F_k$  for  $k \geq 3$  based on [1, 2] and verifies their expected accuracy, success probability and space usage.

#### Contents

1	Preliminary Results	2
2	Frequency Moments	5
3	Ranks, $k$ smallest element and elements	6
4	Landau Symbols	8
5	Probability Spaces	10
6	Frequency Moment 0	11
7	Frequency Moment 2	15
8	Frequency Moment $k$	20
9	Tutorial on the use of Pseudorandom-Objects	25

A	Informal proof of correctness for the $F_0$ algorithm A.1 Case $F_0 \ge t$	<b>30</b> 31 33
1	Preliminary Results	
ir	porty Frequency-Moments-Preliminary-Results mports HOL. Transcendental HOL—Computational-Algebra. Primes HOL—Library. Extended-Real HOL—Library. Multiset HOL—Library. Sublist Prefix-Free-Code-Combinators. Prefix-Free-Code-Combinators Bertrands-Postulate. Bertrand Expander-Graphs. Expander-Graphs-Multiset-Extras gin	
$\operatorname{Th}$	is section contains various preliminary results.	
fi as sl	nma card-ordered-pairs: <b>xes</b> $M$ :: ('a ::linorder) set <b>ssumes</b> finite $M$ <b>hows</b> $2 * card \{(x,y) \in M \times M. \ x < y\} = card \ M * (card \ M - 1)$	
	<b>nma</b> ereal-mono: $x \le y \Longrightarrow ereal \ x \le ereal \ y$ proof $\rangle$	
	nma $abs\text{-}ge\text{-}iff$ : $((x::real) \le abs\ y) = (x \le y \lor x \le -y)$ $proof \rangle$	
(2	nma $count$ -list- $gr$ -1: $x \in set \ xs) = (count$ -list $xs \ x \ge 1)$ $proof \rangle$	
	<b>nma</b> count-list-append: count-list $(xs@ys)$ $v = count-list xs v + count-list y proof$	ıs v
as as sl	nma $count$ -list-lt-suffix: ssumes $suffix \ a \ b$ ssumes $x \in \{b \ ! \ i   \ i. \ i < \ length \ b - \ length \ a\}$ hows $count$ -list $a \ x < count$ -list $b \ x$	
as	nma $suffix-drop-drop$ : ssumes $x \ge y$ hows $suffix (drop \ x \ a) (drop \ y \ a)$ $suffix (drop \ x \ a)$	

```
lemma count-list-card: count-list xs \ x = card \ \{k. \ k < length \ xs \land xs \ ! \ k = x\}
\langle proof \rangle
lemma card-gr-1-iff:
  assumes finite S x \in S y \in S x \neq y
 shows card S > 1
  \langle proof \rangle
lemma count-list-ge-2-iff:
  assumes y < z
 assumes z < length xs
 assumes xs ! y = xs ! z
 shows count-list xs(xs!y) > 1
\langle proof \rangle
Results about multisets and sorting
\mathbf{lemmas}\ disj\text{-}induct\text{-}mset = disj\text{-}induct\text{-}mset
lemma prod-mset-conv:
  fixes f :: 'a \Rightarrow 'b :: \{comm-monoid-mult\}
  shows prod-mset (image-mset f(A) = prod(\lambda x. f(x^*)(count(A|x))) (set-mset A)
\langle proof \rangle
There is a version sum-list-map-eq-sum-count but it doesn't work if the
function maps into the reals.
\mathbf{lemma}\ \mathit{sum-list-eval}:
  fixes f :: 'a \Rightarrow 'b :: \{ring, semiring-1\}
 shows sum-list (map\ f\ xs) = (\sum x \in set\ xs.\ of\text{-nat}\ (count\text{-}list\ xs\ x) * f\ x)
\langle proof \rangle
lemma prod-list-eval:
  fixes f :: 'a \Rightarrow 'b :: \{ring, semiring-1, comm-monoid-mult\}
  shows prod-list (map\ f\ xs) = (\prod x \in set\ xs.\ (f\ x) \cap (count-list\ xs\ x))
lemma sorted-sorted-list-of-multiset: sorted (sorted-list-of-multiset M)
  \langle proof \rangle
lemma count-mset: count (mset xs) a = count-list xs a
  \langle proof \rangle
lemma swap-filter-image: filter-mset g (image-mset fA) = image-mset f (filter-mset
(g \circ f) A)
  \langle proof \rangle
lemma list-eq-iff:
  assumes mset \ xs = mset \ ys
```

assumes sorted xs

```
assumes sorted ys
  shows xs = ys
  \langle proof \rangle
\mathbf{lemma}\ sorted\text{-}list\text{-}of\text{-}multiset\text{-}image\text{-}commute}:
  assumes mono f
  shows sorted-list-of-multiset (image-mset f(M) = map(f(sorted-list-of-multiset))
\langle proof \rangle
Results about rounding and floating point numbers
lemma round-down-ge:
  x \leq round\text{-}down\ prec\ x + 2\ powr\ (-prec)
  \langle proof \rangle
lemma truncate-down-ge:
  x \le truncate-down\ prec\ x + abs\ x * 2\ powr\ (-prec)
\langle proof \rangle
{f lemma}\ truncate	ext{-}down	ext{-}pos:
  assumes x \ge \theta
  shows x * (1 - 2 powr (-prec)) < truncate-down prec x
  \langle proof \rangle
lemma truncate-down-eq:
  assumes truncate-down \ r \ x = truncate-down \ r \ y
  shows abs(x-y) \le max(abs x)(abs y) * 2 powr(-real r)
\langle proof \rangle
definition rat-of-float :: float \Rightarrow rat where
  rat-of-float f = of-int (mantissa\ f) *
    (if exponent f \ge 0 then 2 ^ (nat (exponent f)) else 1 / 2 ^ (nat (-exponent
f)))
lemma real-of-rat-of-float: real-of-rat (rat-of-float x) = real-of-float x
\langle proof \rangle
lemma log-est: log 2 (real n + 1) \leq n
\langle proof \rangle
\mathbf{lemma}\ truncate\text{-}mantissa\text{-}bound:
  abs (|x*2 powr (real r - real-of-int | log 2 |x||)|) \le 2 (r+1) (is ?lhs \le -)
\langle proof \rangle
{f lemma}\ truncate	ext{-}float	ext{-}bit	ext{-}count:
  bit-count (F_e (float-of (truncate-down r x))) \le 10 + 4 * real r + 2*log 2 (2 + 2)
|log 2||x||
  (is ?lhs \le ?rhs)
\langle proof \rangle
```

```
definition prime-above :: nat \Rightarrow nat where prime-above n = (SOME x. x \in \{n..(2*n+2)\} \land prime x)
```

The term prime-above n returns a prime between n and 2\*n+2. Because of Bertrand's postulate there always is such a value. In a refinement of the algorithms, it may make sense to replace this with an algorithm, that finds such a prime exactly or approximately.

The definition is intentionally inexact, to allow refinement with various algorithms, without modifying the high-level mathematical correctness proof.

```
lemma ex-subset:

assumes \exists x \in A. P x

assumes A \subseteq B

shows \exists x \in B. P x

\langle proof \rangle

lemma

shows prime-above-prime: prime (prime-above n)

and prime-above-range: prime-above n \in \{n...(2*n+2)\}

\langle proof \rangle

lemma prime-above-min: prime-above n \geq 2

\langle proof \rangle

lemma prime-above-lower-bound: prime-above n \geq n

\langle proof \rangle

lemma prime-above-upper-bound: prime-above n \leq 2*n+2

\langle proof \rangle

lemma prime-above-upper-bound: prime-above n \leq 2*n+2

\langle proof \rangle
```

#### 2 Frequency Moments

```
theory Frequency-Moments
imports
Frequency-Moments-Preliminary-Results
Finite-Fields.Finite-Fields-Mod-Ring-Code
Interpolation-Polynomials-HOL-Algebra.Interpolation-Polynomial-Cardinalities
begin
```

This section contains a definition of the frequency moments of a stream and a few general results about frequency moments..

```
definition F where F \ k \ xs = (\sum x \in set \ xs. \ (rat\text{-}of\text{-}nat \ (count\text{-}list \ xs \ x)^k)) lemma F\text{-}ge\text{-}0: F \ k \ as \ge 0
```

```
\langle proof \rangle
lemma F-gr-\theta:
 assumes as \neq []
  shows F k as > 0
\langle proof \rangle
definition P_e :: nat \Rightarrow nat \Rightarrow nat \ list \Rightarrow bool \ list \ option \ \mathbf{where}
  P_e \ p \ n \ f = (if \ p > 1 \ \land f \in bounded\text{-}degree\text{-}polynomials} \ (ring\text{-}of \ (mod\text{-}ring \ p)) \ n
    ([0..< n] \rightarrow_e Nb_e p) \ (\lambda i \in \{..< n\}. \ ring.coeff \ (ring-of \ (mod-ring \ p)) \ f \ i) \ else
None)
lemma poly-encoding:
  is-encoding (P_e \ p \ n)
\langle proof \rangle
\mathbf{lemma}\ bounded\text{-}degree\text{-}polynomial\text{-}bit\text{-}count:
 assumes p > 1
 assumes x \in bounded-degree-polynomials (ring-of (mod-ring p)) n
 shows bit-count (P_e \ p \ n \ x) \le ereal \ (real \ n * (log \ 2 \ p + 1))
\langle proof \rangle
end
3
      Ranks, k smallest element and elements
theory K-Smallest
 imports
    Frequency-Moments-Preliminary-Results
    Interpolation	ext{-}Polynomials	ext{-}HOL	ext{-}Algebra. Interpolation	ext{-}Polynomial	ext{-}Cardinalities
begin
This section contains definitions and results for the selection of the k smallest
elements, the k-th smallest element, rank of an element in an ordered set.
definition rank-of :: 'a :: linorder \Rightarrow 'a set \Rightarrow nat where rank-of x S = card \{y\}
\in S. \ y < x
The function rank-of returns the rank of an element within a set.
lemma rank-mono:
  assumes finite S
 shows x \leq y \Longrightarrow rank\text{-}of \ x \ S \leq rank\text{-}of \ y \ S
```

lemma rank-mono-2: assumes finite S

 $\langle proof \rangle$ 

shows  $S' \subseteq S \Longrightarrow rank\text{-}of \ x \ S' \le rank\text{-}of \ x \ S$ 

```
lemma rank-mono-commute:
 assumes finite S
 assumes S \subseteq T
 assumes strict-mono-on Tf
 assumes x \in T
 shows rank-of x S = rank-of (f x) (f S)
\langle proof \rangle
definition least where least k S = \{y \in S. \text{ rank-of } y S < k\}
The function K-Smallest least returns the k smallest elements of a finite set.
lemma rank-strict-mono:
 assumes finite S
 shows strict-mono-on S (\lambda x. rank-of x S)
\langle proof \rangle
lemma rank-of-image:
  assumes finite S
 shows (\lambda x. \ rank\text{-}of \ x \ S) \ 'S = \{\theta..< card \ S\}
\langle proof \rangle
lemma card-least:
 assumes finite S
 shows card (least k S) = min k (card S)
lemma least-subset: least k S \subseteq S
  \langle proof \rangle
lemma least-mono-commute:
 assumes finite S
 assumes strict-mono-on S f
  shows f ' least k S = least k (f ' S)
\langle proof \rangle
lemma least-eq-iff:
 assumes finite B
 assumes A \subseteq B
 assumes \bigwedge x. \ x \in B \Longrightarrow rank \text{-} of \ x \ B < k \Longrightarrow x \in A
 shows least k A = least k B
\langle proof \rangle
lemma least-insert:
 assumes finite S
  shows least k (insert x (least k S)) = least k (insert x S) (is ?lhs = ?rhs)
\langle proof \rangle
```

```
definition count-le where count-le x M = size \{ \# y \in \# M. \ y \leq x \# \}
definition count-less where count-less x M = size \{ \# y \in \# M. \ y < x \# \}
definition nth-mset :: nat \Rightarrow ('a :: linorder) multiset <math>\Rightarrow 'a where
  nth-mset\ k\ M = sorted-list-of-multiset\ M\ !\ k
lemma nth-mset-bound-left:
  assumes k < size M
 assumes count-less x M \leq k
 shows x \leq nth-mset k M
\langle proof \rangle
\mathbf{lemma}\ nth\text{-}mset\text{-}bound\text{-}left\text{-}excl\text{:}
 assumes k < size M
 assumes count-le x M < k
 shows x < nth-mset k M
\langle proof \rangle
lemma nth-mset-bound-right:
 assumes k < size M
 assumes count-le x M > k
  shows nth-mset k M \leq x
\langle proof \rangle
{f lemma} nth-mset-commute-mono:
  assumes mono f
 assumes k < size M
 shows f (nth\text{-}mset\ k\ M) = nth\text{-}mset\ k\ (image\text{-}mset\ f\ M)
\langle proof \rangle
lemma nth-mset-max:
  assumes size A > k
 assumes \bigwedge x. x \leq nth-mset k \land A \implies count \land x \leq 1
  shows nth-mset k A = Max (least (k+1) (set-mset A)) and card (least (k+1)
(set\text{-}mset\ A)) = k+1
\langle proof \rangle
end
```

### 4 Landau Symbols

```
theory Landau-Ext

imports

HOL-Library.Landau-Symbols

HOL.Topological-Spaces

begin
```

This section contains results about Landau Symbols in addition to "HOL-Library.Landau".

```
lemma landau-sum:
  assumes eventually (\lambda x. \ g1 \ x \ge (0::real)) F
  assumes eventually (\lambda x. g2 x \geq 0) F
 assumes f1 \in O[F](g1)
 assumes f2 \in O[F](g2)
  shows (\lambda x. f1 \ x + f2 \ x) \in O[F](\lambda x. g1 \ x + g2 \ x)
\langle proof \rangle
lemma landau-sum-1:
  assumes eventually (\lambda x. \ g1 \ x \ge (0::real)) F
 assumes eventually (\lambda x. g2 \ x \ge 0) \ F
 assumes f \in O[F](g1)
 shows f \in O[F](\lambda x. g1 x + g2 x)
\langle proof \rangle
lemma landau-sum-2:
  assumes eventually (\lambda x. \ g1 \ x \geq (0::real)) F
 assumes eventually (\lambda x. g2 x \geq 0) F
 assumes f \in O[F](g2)
  shows f \in O[F](\lambda x. g1 x + g2 x)
\langle proof \rangle
lemma landau-ln-3:
  assumes eventually (\lambda x. (1::real) \leq f x) F
  assumes f \in O[F](g)
  shows (\lambda x. \ln (f x)) \in O[F](g)
\langle proof \rangle
lemma landau-ln-2:
 assumes a > (1::real)
 assumes eventually (\lambda x. \ 1 \leq f x) \ F
  assumes eventually (\lambda x. \ a \leq g \ x) \ F
 assumes f \in O[F](g)
  shows (\lambda x. \ln (f x)) \in O[F](\lambda x. \ln (g x))
\langle proof \rangle
\mathbf{lemma}\ landau\text{-}real\text{-}nat:
  fixes f :: 'a \Rightarrow int
 assumes (\lambda x. \ of\text{-}int \ (f \ x)) \in O[F](g)
  shows (\lambda x. \ real \ (nat \ (f \ x))) \in O[F](g)
\langle proof \rangle
lemma landau-ceil:
 assumes (\lambda -. 1) \in O[F'](g)
 assumes f \in O[F'](g)
  shows (\lambda x. real\text{-}of\text{-}int [f x]) \in O[F'](g)
\langle proof \rangle
```

lemma landau-rat-ceil:

```
assumes (\lambda -. 1) \in O[F'](g)
  assumes (\lambda x. real-of-rat (f x)) \in O[F'](g)
  shows (\lambda x. real-of-int [f x]) \in O[F'](g)
\langle proof \rangle
\mathbf{lemma}\ \mathit{landau}\text{-}\mathit{nat}\text{-}\mathit{ceil}\text{:}
  assumes (\lambda -. 1) \in O[F'](g)
  assumes f \in O[F'](g)
  shows (\lambda x. \ real \ (nat \ \lceil f \ x \rceil)) \in O[F'](g)
  \langle proof \rangle
lemma eventually-prod1':
  assumes B \neq bot
  assumes (\forall_F x in A. P x)
  shows (\forall_F \ x \ in \ A \times_F B. \ P \ (fst \ x))
\langle proof \rangle
lemma eventually-prod2':
  assumes A \neq bot
  assumes (\forall_F x \text{ in } B. P x)
  shows (\forall_F x in A \times_F B. P (snd x))
\langle proof \rangle
lemma sequentially-inf: \forall F \ x \ in \ sequentially. \ n \leq real \ x
  \langle proof \rangle
instantiation rat :: linorder-topology
begin
definition open-rat :: rat \ set \Rightarrow bool
  where open-rat = generate-topology (range (\lambda a. \{... < a\}) \cup range (\lambda a. \{a < ... \}))
instance
  \langle proof \rangle
end
lemma inv-at-right-0-inf:
  \forall_F \ x \ in \ at\text{-right } 0. \ c \leq 1 \ / \ real\text{-of-rat } x
\langle proof \rangle
end
```

### 5 Probability Spaces

Some additional results about probability spaces in addition to "HOL-Probability".

```
theory Probability-Ext
imports
HOL-Probability.Stream-Space
```

```
Concentration-Inequalities. Bienaymes-Identity
    Universal	ext{-}Hash	ext{-}Families. Carter	ext{-}Wegman	ext{-}Hash	ext{-}Family
    Frequency-Moments-Preliminary-Results
begin
context prob-space
begin
lemma pmf-mono:
  assumes M = measure-pmf p
  assumes \bigwedge x. x \in P \Longrightarrow x \in set\text{-pmf } p \Longrightarrow x \in Q
  shows prob P \leq prob Q
\langle proof \rangle
lemma pmf-add:
  assumes M = measure-pmf p
  assumes \bigwedge x. \ x \in P \Longrightarrow x \in \textit{set-pmf} \ p \Longrightarrow x \in Q \lor x \in R
  \mathbf{shows} \ prob \ P \leq prob \ Q + prob \ R
\langle proof \rangle
lemma pmf-add-2:
  assumes M = measure-pmf p
  assumes prob \{\omega. P \omega\} \leq r1
  assumes prob \{\omega. \ Q \ \omega\} \leq r2
  shows prob \{\omega. \ P \ \omega \lor Q \ \omega\} \le r1 + r2 \ (is ?lhs \le ?rhs)
\langle proof \rangle
end
end
```

#### **6** Frequency Moment 0

```
theory Frequency-Moment-0
imports
  Frequency-Moments-Preliminary-Results
  Median-Method.Median
  K-Smallest
  Universal-Hash-Families.Carter-Wegman-Hash-Family
  Frequency-Moments
  Landau-Ext
  Probability-Ext
  Universal-Hash-Families.Universal-Hash-Families-More-Product-PMF
begin
```

This section contains a formalization of a new algorithm for the zero-th frequency moment inspired by ideas described in [2]. It is a KMV-type (k-minimum value) algorithm with a rounding method and matches the space complexity of the best algorithm described in [2].

```
In addition to the Isabelle proof here, there is also an informal hand-written proof in Appendix A.
```

```
type-synonym f0-state = nat \times nat \times nat \times nat \times (nat \Rightarrow nat \ list) \times (nat \Rightarrow nat \ list)
float set)
definition hash where hash p = ring.hash (ring-of (mod-ring p))
fun f0-init :: rat \Rightarrow rat \Rightarrow nat \Rightarrow f0-state pmf where
  f0-init \delta \varepsilon n =
    do \{
       let s = nat \left[ -18 * ln \left( real-of-rat \varepsilon \right) \right];
       let t = nat \lceil 80 / (real-of-rat \delta)^2 \rceil;
       let p = prime-above (max \ n \ 19);
       let r = nat \left( 4 * \lceil log 2 \left( 1 / real-of-rat \delta \right) \rceil + 23 \right);
         h \leftarrow prod\text{-}pmf \ \{...< s\} \ (\lambda\text{-. }pmf\text{-}of\text{-}set \ (bounded\text{-}degree\text{-}polynomials \ (ring\text{-}of\text{-}set)\}
(mod\text{-}ring\ p))\ 2));
       return-pmf (s, t, p, r, h, (\lambda \in \{0... < s\}. \{\}))
    }
fun f0-update :: nat \Rightarrow f0-state \Rightarrow f0-state pmf where
  f0-update x (s, t, p, r, h, sketch) =
    return-pmf (s, t, p, r, h, \lambda i \in \{... < s\}.
       least\ t\ (insert\ (float-of\ (truncate-down\ r\ (hash\ p\ x\ (h\ i))))\ (sketch\ i)))
fun f0-result :: f0-state \Rightarrow rat pmf where
  f0-result (s, t, p, r, h, sketch) = return-pmf (median <math>s (\lambda i \in \{... < s\}).
       (if \ card \ (sketch \ i) < t \ then \ of-nat \ (card \ (sketch \ i)) \ else
         rat-of-nat t* rat-of-nat p / rat-of-float (Max (sketch i)))
    ))
\mathbf{fun}\ \mathit{f0-space-usage} :: (\mathit{nat} \times \mathit{rat} \times \mathit{rat}) \Rightarrow \mathit{real}\ \mathbf{where}
  f0-space-usage (n, \varepsilon, \delta) = (
    let s = nat \left[ -18 * ln \left( real-of-rat \varepsilon \right) \right] in
    let r = nat \left( 4 * \lceil log 2 \left( 1 / real-of-rat \delta \right) \rceil + 23 \right) in
    let t = nat \lceil 80 / (real - of - rat \delta)^2 \rceil in
    6 +
    2 * log 2 (real s + 1) +
    2 * log 2 (real t + 1) +
    2 * log 2 (real n + 21) +
    2 * log 2 (real r + 1) +
    real \ s * (5 + 2 * log 2 (21 + real n) + 1)
    real\ t*(13+4*r+2*log\ 2\ (log\ 2\ (real\ n+13)))))
definition encode-f0-state :: f0-state \Rightarrow bool \ list \ option \ \mathbf{where}
  encode-f0-state =
    N_e \bowtie_e (\lambda s.
    N_e \times_e (
    N_e \bowtie_e (\lambda p.
    N_e \times_e (
```

```
([0..< s] \rightarrow_e (P_e \ p \ 2)) \times_e
    ([\theta .. < s] \rightarrow_e (S_e F_e))))))
lemma inj-on encode-f0-state (dom encode-f0-state)
\langle proof \rangle
context
  fixes \varepsilon \delta :: rat
  fixes n :: nat
  fixes as :: nat \ list
  fixes result
  assumes \varepsilon-range: \varepsilon \in \{0 < .. < 1\}
  assumes \delta-range: \delta \in \{0 < ... < 1\}
  assumes as-range: set as \subseteq \{... < n\}
  defines result \equiv fold (\lambda a state. state \gg f0-update a) as (f0-init \delta \varepsilon n) \gg
f0-result
begin
private definition t where t = nat [80 / (real-of-rat \delta)^2]
private lemma t-gt-0: t > 0 \ \langle proof \rangle definition s where s = nat \ \lceil -(18 * ln) \rceil
(real-of-rat \ \varepsilon))
private lemma s-gt-0: s > 0 \ \langle proof \rangle definition p where p = prime-above \ (max
n 19
private lemma p-prime: Factorial-Ring.prime p
  \langle proof \rangle lemma p-ge-18: p \geq 18
\langle proof \rangle lemma p-qt-0: p > 0 \langle proof \rangle lemma p-qt-1: p > 1 \langle proof \rangle lemma n-le-p:
n \leq p
\langle proof \rangle lemma p-le-n: p \leq 2*n + 40
\langle proof \rangle lemma as-lt-p: \bigwedge x. x \in set \ as \Longrightarrow x < p
  \langle proof \rangle lemma as-subset-p: set as \subseteq \{... < p\}
   \langle proof \rangle definition r where r = nat (4 * \lceil log 2 (1 / real-of-rat \delta) \rceil + 23)
private lemma r-bound: 4 * log 2 (1 / real-of-rat \delta) + 23 \le r
\langle proof \rangle lemma r-ge-23: r \geq 23
\langle proof \rangle lemma two-pow-r-le-1: 0 < 1 - 2 powr - real r
\langle proof \rangle
interpretation carter-wegman-hash-family ring-of (mod-ring p) 2
  rewrites ring.hash (ring-of (mod-ring p)) = Frequency-Moment-0.hash p
  \langle proof \rangle definition tr-hash where tr-hash x \omega = truncate-down r (hash x \omega)
private definition sketch-rv where
  sketch-rv\ \omega = least\ t\ ((\lambda x.\ float-of\ (tr-hash\ x\ \omega))\ `set\ as)
private definition estimate
   where estimate S = (if \ card \ S < t \ then \ of -nat \ (card \ S) \ else \ of -nat \ t * of -nat \ p
/ rat-of-float (Max S)
```

```
private definition sketch-rv' where sketch-rv' \omega = least \ t \ ((\lambda x. \ tr-hash \ x \ \omega))
private definition estimate' where estimate' S = (if \ card \ S < t \ then \ real \ (card
S) else real t * real p / Max S)
private definition \Omega_0 where \Omega_0 = prod\text{-}pmf \{... < s\} \ (\lambda\text{-. }pmf\text{-}of\text{-}set space)
private lemma f0-alg-sketch:
    defines sketch \equiv fold (\lambda a state. state \gg f0-update a) as (f0-init \delta \varepsilon n)
    shows sketch = map-pmf (\lambda x. (s,t,p,r, x, \lambda i \in \{... < s\}. sketch-rv (x i))) \Omega_0
    \langle proof \rangle lemma card-nat-in-ball:
    fixes x :: nat
    fixes q :: real
    assumes q \geq 0
    defines A \equiv \{k. \ abs \ (real \ x - real \ k) \le q \land k \ne x\}
    shows real (card A) \leq 2 * q and finite A
\langle proof \rangle lemma prob-degree-lt-1:
      prob \{\omega.\ degree\ \omega < 1\} \le 1/real\ p
\langle proof \rangle lemma collision-prob:
    assumes c \geq 1
    shows prob \{\omega. \exists x \in set \ as. \exists y \in set \ as. \ x \neq y \land tr-hash \ x \ \omega \leq c \land tr-hash \ x \}
\omega = tr-hash y \omega \} \le
         (5/2) * (real (card (set as)))^2 * c^2 * 2 powr - (real r) / (real p)^2 + 1/real p
(is prob \{\omega. ?l \omega\} \leq ?r1 + ?r2)
\langle proof \rangle lemma of-bool-square: (of\text{-bool }x)^2 = ((of\text{-bool }x)::real)
    \langle proof \rangle definition Q where Q y \omega = card \{x \in set \ as. \ int \ (hash \ x \ \omega) < y\}
private definition m where m = card (set as)
private lemma
    assumes a \geq 0
    assumes a \leq int p
    shows exp-Q: expectation (\lambda \omega. real (Q \ a \ \omega)) = real \ m * (of-int \ a) / p
    and var-Q: variance (\lambda \omega. real (Q \ a \ \omega)) \leq real \ m * (of-int \ a) / p
\langle proof \rangle lemma t-bound: t \leq 81 / (real\text{-}of\text{-}rat \delta)^2
\langle proof \rangle lemma t-r-bound:
    18 * 40 * (real t)^2 * 2 powr (-real r) \le 1
\langle proof \rangle lemma m-eq-F-0: real m = of-rat (F \ 0 \ as)
    \langle proof \rangle lemma estimate'-bounds:
    prob \ \{\omega. \ of\text{-rat} \ \delta * real\text{-}of\text{-}rat \ (F \ 0 \ as) < | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv' \ \omega) - of\text{-}rat \ (F \ 0 \ as) | estimate' \ (sketch\text{-}rv'
|as| \le 1/3
\langle proof \rangle lemma median-bounds:
    \mathcal{P}(\omega \text{ in measure-pmf } \Omega_0. | \text{median s } (\lambda i. \text{ estimate } (\text{sketch-rv } (\omega i))) - F \text{ 0 as} | \leq
\delta * F \ 0 \ as) \ge 1 - real - of - rat \ \varepsilon
\langle proof \rangle
lemma f0-alg-correct':
    \mathcal{P}(\omega \text{ in measure-pmf result. } |\omega - F \text{ 0 as}| \leq \delta * F \text{ 0 as}) \geq 1 - \text{of-rat } \varepsilon
\langle proof \rangle lemma f-subset:
```

```
assumes g 'A \subseteq h 'B
  shows (\lambda x. f(g x)) \cdot A \subseteq (\lambda x. f(h x)) \cdot B
  \langle proof \rangle
lemma f0-exact-space-usage':
  defines \Omega \equiv fold \ (\lambda a \ state. \ state \gg f0-update a) as (f0-init \delta \in n)
  shows AE \omega in \Omega. bit-count (encode-f0-state \omega) \leq f0-space-usage (n, \varepsilon, \delta)
\langle proof \rangle
end
Main results of this section:
theorem f0-alg-correct:
  assumes \varepsilon \in \{0 < .. < 1\}
  assumes \delta \in \{0 < .. < 1\}
  assumes set \ as \subseteq \{..< n\}
  defines \Omega \equiv fold (\lambda a \ state. \ state \gg f0-update a) as (f0-init \delta \in n) \gg f0-result
  shows \mathcal{P}(\omega \text{ in measure-pmf } \Omega. |\omega - F \text{ 0 as}| \leq \delta * F \text{ 0 as}) \geq 1 - \text{ of-rat } \varepsilon
  \langle proof \rangle
theorem f0-exact-space-usage:
  assumes \varepsilon \in \{0 < .. < 1\}
  assumes \delta \in \{0 < .. < 1\}
  assumes set \ as \subseteq \{..< n\}
  defines \Omega \equiv fold (\lambda a \ state. \ state \gg f0-update a) as (f0-init \delta \varepsilon n)
  shows AE \omega in \Omega. bit-count (encode-f0-state \omega) \leq f0-space-usage (n, \varepsilon, \delta)
  \langle proof \rangle
theorem f0-asymptotic-space-complexity:
  f0-space-usage \in O[at-top \times_F at-right 0 \times_F at-right 0](\lambda(n, \varepsilon, \delta). ln (1 / of-rat
\varepsilon) *
  (ln (real n) + 1 / (of-rat \delta)^2 * (ln (ln (real n)) + ln (1 / of-rat \delta))))
  (\mathbf{is} - \in O[?F](?rhs))
\langle proof \rangle
end
       Frequency Moment 2
theory Frequency-Moment-2
  imports
    Universal-Hash-Families. Carter-Wegman-Hash-Family
    Equivalence-Relation-Enumeration. Equivalence-Relation-Enumeration
    Landau-Ext
    Median	ext{-}Method.Median
    Probability-Ext
    Universal\hbox{-} Hash\hbox{-} Families. \ Universal\hbox{-} Hash\hbox{-} Families\hbox{-} More\hbox{-} Product\hbox{-} PMF
```

Frequency-Moments

begin

```
hide-const (open) Discrete-Topology.discrete
hide-const (open) Isolated.discrete
```

This section contains a formalization of the algorithm for the second frequency moment. It is based on the algorithm described in [1, §2.2]. The only difference is that the algorithm is adapted to work with prime field of odd order, which greatly reduces the implementation complexity.

```
fun f2-hash where
     f2-hash p h k = (if even (ring.hash (ring-of (mod-ring <math>p))) k h) then int p-1
else - int p - 1)
type-synonym f2-state = nat \times nat \times nat \times (nat \times nat \Rightarrow nat \ list) \times (nat \times nat \Rightarrow nat \ list)
nat \Rightarrow int)
fun f2-init :: rat \Rightarrow rat \Rightarrow nat \Rightarrow f2-state pmf where
    f2-init \delta \varepsilon n =
         do \{
              let s_1 = nat \lceil 6 / \delta^2 \rceil;
              let s_2 = nat \left[ -(18 * ln (real-of-rat \varepsilon)) \right];
              let p = prime-above (max n 3);
             h \leftarrow prod\text{-}pmf \ (\{...< s_1\} \times \{...< s_2\}) \ (\lambda\text{-. }pmf\text{-}of\text{-}set \ (bounded\text{-}degree\text{-}polynomials
(ring-of (mod-ring p)) \not\downarrow));
             return-pmf (s_1, s_2, p, h, (\lambda \in \{... < s_1\} \times \{... < s_2\}. (0 :: int)))
         }
fun f2-update :: nat \Rightarrow f2-state \Rightarrow f2-state pmf where
    f2-update x (s_1, s_2, p, h, sketch) =
         return-pmf (s_1, s_2, p, h, \lambda i \in \{... < s_1\} \times \{... < s_2\}. f2-hash p (h \ i) \ x + sketch \ i)
fun f2-result :: f2-state \Rightarrow rat pmf where
    f2-result (s_1, s_2, p, h, sketch) =
         return-pmf (median s_2 (\lambda i_2 \in \{... < s_2\}).
                     (\sum i_1 \in \{... < s_1\}) . (rat\text{-}of\text{-}int\ (sketch\ (i_1,\ i_2)))^2) / (((rat\text{-}of\text{-}nat\ p)^2 - 1))
rat-of-nat s_1)))
fun f2-space-usage :: (nat \times nat \times rat \times rat) \Rightarrow real where
    f2-space-usage (n, m, \varepsilon, \delta) = (
         let s_1 = nat \lceil 6 / \delta^2 \rceil in
         let s_2 = nat \left[ -(18 * ln (real-of-rat \varepsilon)) \right] in
         3 +
         2 * log 2 (s_1 + 1) +
         2 * log 2 (s_2 + 1) +
         2 * log 2 (9 + 2 * real n) +
         s_1 * s_2 * (5 + 4*log 2 (8 + 2*real n) + 2*log 2 (real m*(18 + 4*real n
(n) + (1)
definition encode-f2-state :: <math>f2-state \Rightarrow bool \ list \ option \ \mathbf{where}
```

encode-f2-state =

```
N_e \bowtie_e (\lambda s_1.
    N_e \bowtie_e (\lambda s_2.
    N_e \bowtie_e (\lambda p.
    (List.product [0..< s_1] [0..< s_2] \rightarrow_e P_e p \not\downarrow) \times_e
    (List.product [\theta..< s_1] [\theta..< s_2] \rightarrow_e I_e))))
lemma inj-on encode-f2-state (dom encode-f2-state)
\langle proof \rangle
context
  fixes \varepsilon \delta :: rat
  fixes n :: nat
  fixes as :: nat \ list
  \mathbf{fixes}\ \mathit{result}
  assumes \varepsilon-range: \varepsilon \in \{0 < ... < 1\}
  assumes \delta-range: \delta > 0
  assumes as-range: set as \subseteq \{..< n\}
  defines result \equiv fold (\lambda a state. state \gg f2-update a) as (f2-init \delta \varepsilon n) \gg
f2-result
begin
private definition s_1 where s_1 = nat \lceil 6 / \delta^2 \rceil
lemma s1-gt-\theta: s_1 > \theta
    \langle proof \rangle definition s_2 where s_2 = nat \left[ -(18* ln (real-of-rat \varepsilon)) \right]
lemma s2-gt-\theta: s_2 > \theta
    \langle proof \rangle definition p where p = prime-above (max n 3)
lemma p-prime: Factorial-Ring.prime p
  \langle proof \rangle
lemma p-ge-\beta: p \geq \beta
    \langle proof \rangle
lemma p-gt-\theta: p > \theta \langle proof \rangle
lemma p-gt-1: p > 1 \langle proof \rangle
lemma p-ge-n: p \ge n \langle proof \rangle
interpretation carter-wegman-hash-family ring-of (mod-ring p) 4
definition sketch where sketch = fold (\lambda a state. state \gg f2-update a) as (f2-init
private definition \Omega where \Omega = prod-pmf ({..<s_1} \times {..<s_2}) (\lambda-. pmf-of-set
private definition \Omega_p where \Omega_p = measure-pmf \Omega
```

```
private definition sketch-rv where sketch-rv \omega = of-int (sum-list (map (f2-hash
p(\omega)(as))^2
private definition mean-rv where mean-rv \omega = (\lambda i_2. (\sum i_1 = 0... < s_1. sketch-rv)
(\omega (i_1, i_2))) / (((of-nat p)^2 - 1) * of-nat s_1))
private definition result-rv where result-rv \omega = median \ s_2 \ (\lambda i_2 \in \{... < s_2\}. \ mean-rv
\omega i_2
lemma mean-rv-alg-sketch:
    sketch = \Omega \gg (\lambda \omega. \ return-pmf \ (s_1, \ s_2, \ p, \ \omega, \ \lambda i \in \{...< s_1\} \times \{...< s_2\}. \ sum-list
(map (f2-hash p (\omega i)) as)))
\langle proof \rangle
lemma distr: result = map-pmf \ result-rv \ \Omega
\langle proof \rangle lemma f2-hash-pow-exp:
   assumes k < p
   shows
         expectation (\lambda \omega. real-of-int (f2-hash p \omega k) \widehat{m} =
         ((real \ p-1) \ \hat{\ } m * (real \ p+1) + (-real \ p-1) \ \hat{\ } m * (real \ p-1)) / (2 *
real p
\langle proof \rangle
lemma
     shows var-sketch-rv:variance sketch-rv \leq 2*(real-of-rat (F 2 as)^2) * ((real
(p)^2 - 1)^2 (is ?A)
   and exp-sketch-rv:expectation sketch-rv = real-of-rat (F \ 2 \ as) * ((real \ p)^2 - 1) (is
?B)
\langle proof \rangle
lemma space-omega-1 [simp]: Sigma-Algebra.space \Omega_p = UNIV
interpretation \Omega: prob-space \Omega_p
    \langle proof \rangle
lemma integrable-\Omega:
    fixes f :: ((nat \times nat) \Rightarrow (nat \ list)) \Rightarrow real
    shows integrable \Omega_p f
    \langle proof \rangle
lemma sketch-rv-exp:
    assumes i_2 < s_2
    assumes i_1 \in \{\theta ... < s_1\}
    shows \Omega.expectation (\lambda \omega. sketch-rv (\omega (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real-of-rat (F 2 as) * ((real (i_1, i_2))) = real
(p)^2 - 1)
\langle proof \rangle
lemma sketch-rv-var:
   assumes i_2 < s_2
   assumes i_1 \in \{\theta .. < s_1\}
```

```
shows \Omega.variance\ (\lambda\omega.\ sketch-rv\ (\omega\ (i_1,\ i_2))) \le 2*(real-of-rat\ (F\ 2\ as))^2*
((real \ p)^2 - 1)^2
\langle proof \rangle
lemma mean-rv-exp:
  assumes i < s_2
  shows \Omega.expectation (\lambda \omega. mean-rv \omega i) = real-of-rat (F 2 as)
\langle proof \rangle
lemma mean-rv-var:
  assumes i < s_2
  shows \Omega. variance (\lambda \omega. mean-rv \omega i) \leq (real-of-rat (\delta * F 2 as))^2 / 3
\langle proof \rangle
lemma mean-rv-bounds:
  assumes i < s_2
 shows \Omega.prob~\{\omega.~real\mbox{-}of\mbox{-}rat~\delta*~real\mbox{-}of\mbox{-}rat~(F~2~as)<|mean\mbox{-}rv~\omega~i~-~real\mbox{-}of\mbox{-}rat
(F \ 2 \ as)|\} \le 1/3
\langle proof \rangle
lemma f2-alg-correct':
   \mathcal{P}(\omega \text{ in measure-pmf result. } |\omega - F 2 \text{ as}| \leq \delta * F 2 \text{ as}) \geq 1 - \text{of-rat } \varepsilon
\langle proof \rangle
lemma f2-exact-space-usage':
   AE \omega in sketch . bit-count (encode-f2-state \omega) \leq f2-space-usage (n, length as, \varepsilon,
\langle proof \rangle
end
Main results of this section:
theorem f2-alg-correct:
  assumes \varepsilon \in \{0 < .. < 1\}
  assumes \delta > 0
  assumes set as \subseteq \{..< n\}
  defines \Omega \equiv fold (\lambda a \ state. \ state \gg f2-update a) as (f2-init \delta \in n) \gg f2-result
  shows \mathcal{P}(\omega \text{ in measure-pmf } \Omega. |\omega - F 2 \text{ as}| \leq \delta * F 2 \text{ as}) \geq 1 - \text{of-rat } \varepsilon
  \langle proof \rangle
theorem f2-exact-space-usage:
  assumes \varepsilon \in \{0 < .. < 1\}
  assumes \delta > 0
  assumes set \ as \subseteq \{..< n\}
  defines M \equiv fold (\lambda a \ state. \ state \gg f2-update a) as (f2-init \delta \in n)
  shows AE \omega in M. bit-count (encode-f2-state \omega) \leq f2-space-usage (n, length as,
\varepsilon, \delta
  \langle proof \rangle
```

```
theorem f2-asymptotic-space-complexity:

f2-space-usage \in O[at\text{-}top \times_F at\text{-}top \times_F at\text{-}right 0 \times_F at\text{-}right 0](\lambda (n, m, \varepsilon, \delta).

(ln (1 / of\text{-}rat \varepsilon)) / (of\text{-}rat \delta)^2 * (ln (real n) + ln (real m)))

(\mathbf{is} - \in O[?F](?rhs))

\langle proof \rangle
```

end

## 8 Frequency Moment k

```
theory Frequency-Moment-k
  imports
     Frequency-Moments
     Landau-Ext
     Lp.Lp
    Median	ext{-}Method.Median
     Probability	ext{-}Ext
     Universal\hbox{-} Hash\hbox{-} Families. \ Universal\hbox{-} Hash\hbox{-} Families\hbox{-} More\hbox{-} Product\hbox{-} PMF
begin
This section contains a formalization of the algorithm for the k-th frequency
moment. It is based on the algorithm described in [1, §2.1].
type-synonym \textit{fk-state} = \textit{nat} \times \textit{nat} \times \textit{nat} \times \textit{nat} \times \textit{nat} \times \textit{nat} \times \textit{nat} \Rightarrow (\textit{nat} \times \textit{nat}))
fun fk-init :: nat \Rightarrow rat \Rightarrow rat \Rightarrow nat \Rightarrow fk-state pmf where
  fk-init k \delta \varepsilon n =
    do \{
       let s_1 = nat \left[ 3 * real \ k * n \ powr \left( 1 - 1 / real \ k \right) / \left( real-of-rat \ \delta \right)^2 \right];
       let s_2 = nat \left[ -18 * ln \left( real-of-rat \varepsilon \right) \right];
       return-pmf (s_1, s_2, k, \theta, (\lambda - \in \{\theta ... < s_1\} \times \{\theta ... < s_2\}. (\theta, \theta)))
fun fk-update :: nat \Rightarrow fk-state \Rightarrow fk-state pmf where
  fk-update a(s_1, s_2, k, m, r) =
      coins \leftarrow prod\text{-}pmf (\{0...< s_1\} \times \{0...< s_2\}) (\lambda -. bernoulli-pmf (1/(real m+1)));
       return-pmf (s_1, s_2, k, m+1, \lambda i \in \{0... < s_1\} \times \{0... < s_2\}.
          if coins i then
            (a, \theta)
          else (
            let (x,l) = r i in (x, l + of\text{-bool}(x=a))
```

```
fun fk-result :: fk-state \Rightarrow rat pmf where
fk-result (s_1, s_2, k, m, r) =
return-pmf (median \ s_2 \ (\lambda i_2 \in \{0... < s_2\}).
```

}

```
(\sum i_1 \in \{0... < s_1\}. \ rat\text{-of-nat} \ (let \ t = snd \ (r \ (i_1, \ i_2)) + 1 \ in \ m * (t^k - (t - i_1)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k - (t - i_2)) + 1 \ in \ m * (t^k 
1)\hat{k}))) / (rat-of-nat s_1))
lemma bernoulli-pmf-1: bernoulli-pmf 1 = return-pmf True
     \langle proof \rangle
fun fk-space-usage :: (nat \times nat \times nat \times rat \times rat) \Rightarrow real where
    \mathit{fk\text{-}space\text{-}usage}\ (k,\ n,\ m,\ \varepsilon,\ \delta) = (
        let s_1 = nat [3*real k* (real n) powr (1-1/real k) / (real-of-rat \delta)^2] in
        let s_2 = nat \left[ -(18 * ln (real-of-rat \varepsilon)) \right] in
        4 +
        2 * log 2 (s_1 + 1) +
        2 * log 2 (s_2 + 1) +
        2 * log 2 (real k + 1) +
        2 * log 2 (real m + 1) +
        s_1 * s_2 * (2 + 2 * log 2 (real n+1) + 2 * log 2 (real m+1)))
definition encode-fk-state :: fk-state \Rightarrow bool \ list \ option \ \mathbf{where}
     encode-fk-state =
         N_e \bowtie_e (\lambda s_1.
        N_e \bowtie_e (\lambda s_2.
        N_e \times_e
        N_e \times_e
        (List.product \ [\theta...< s_1] \ [\theta...< s_2] \ \rightarrow_e \ (N_e \ \times_e \ N_e))))
lemma inj-on encode-fk-state (dom encode-fk-state)
\langle proof \rangle
This is an intermediate non-parallel form fk-update used only in the correct-
ness proof.
fun fk-update-2 :: 'a \Rightarrow (nat \times 'a \times nat) \Rightarrow (nat \times 'a \times nat) pmf where
   fk-update-2 a (m,x,l) =
             coin \leftarrow bernoulli-pmf(1/(real m+1));
             return-pmf (m+1,if\ coin\ then\ (a,0)\ else\ (x,\ l+of\ bool\ (x=a)))
definition sketch where sketch as i = (as ! i, count-list (drop (i+1) as) (as ! i))
lemma fk-update-2-distr:
    assumes as \neq []
    shows fold (\lambda x \ s. \ s \gg fk\text{-update-2} \ x) as (return\text{-pmf} \ (0,0,0)) =
    pmf-of-set {..<length as} \gg (\lambda k. return-pmf (length as, sketch as k))
    \langle proof \rangle
context
     fixes \varepsilon \delta :: rat
    fixes n k :: nat
```

```
fixes as
  assumes k-ge-1: k \ge 1
  assumes \varepsilon-range: \varepsilon \in \{0 < .. < 1\}
  assumes \delta-range: \delta > 0
  assumes as-range: set as \subseteq \{... < n\}
begin
definition s_1 where s_1 = nat [3 * real k * (real n) powr (1-1/real k) / (real-of-rat)]
\delta)^2
definition s_2 where s_2 = nat \left[ -(18 * ln (real-of-rat \varepsilon)) \right]
definition M_1 = \{(u, v). \ v < count\text{-list as } u\}
definition \Omega_1 = measure-pmf (pmf-of-set M_1)
definition M_2 = prod\text{-}pmf \ (\{0...< s_1\} \times \{0...< s_2\}) \ (\lambda\text{-. }pmf\text{-}of\text{-}set \ M_1)
definition \Omega_2 = measure-pmf M_2
interpretation prob-space \Omega_1
  \langle proof \rangle
interpretation \Omega_2:prob-space \Omega_2
  \langle proof \rangle
lemma split-space: (\sum a \in M_1. f \ (snd \ a)) = (\sum u \in set \ as. \ (\sum v \in \{0.. < count-list \})
as\ u}. f\ v))
\langle proof \rangle
lemma
  assumes as \neq []
  shows fin-space: finite M_1
    and non-empty-space: M_1 \neq \{\}
    and card-space: card M_1 = length as
\langle proof \rangle
lemma
  assumes as \neq []
  shows integrable-1: integrable \Omega_1 (f :: - \Rightarrow real) and
    integrable-2: integrable \Omega_2 (g :: - \Rightarrow real)
\langle proof \rangle
\mathbf{lemma}\ \mathit{sketch-distr}\colon
  assumes as \neq []
 shows pmf-of-set {..<length\ as} \gg (\lambda k.\ return-pmf\ (sketch\ as\ k)) = pmf-of-set
M_1
\langle proof \rangle
lemma fk-update-distr:
 fold (\lambda x \ s. \ s \gg fk-update x) as (fk-init k \ \delta \ \varepsilon \ n) =
 prod\text{-}pmf ({0..<s_1} × {0..<s_2}) (\lambda-. fold (\lambda x s. s \gg fk-update-2 x) as (return\text{-}pmf
```

```
(0,0,0))
         \gg (\lambda x. return\text{-}pmf (s_1, s_2, k, length as, <math>\lambda i \in \{0... < s_1\} \times \{0... < s_2\}. snd (x i)))
\langle proof \rangle
lemma power-diff-sum:
    fixes a \ b :: 'a :: \{comm-ring-1, power\}
    assumes k > \theta
    shows a^k - b^k = (a-b) * (\sum i = 0... < k. \ a^i * b^k = (k-1-i)) (is ?lhs =
?rhs)
\langle proof \rangle
lemma power-diff-est:
    assumes k > \theta
    assumes (a :: real) \ge b
    assumes b > 0
    shows a^k - b^k \le (a-b) * k * a^k - 1
Specialization of the Hoelder inquality for sums.
\mathbf{lemma}\ \mathit{Holder-inequality-sum}\colon
    assumes p > (0::real) \ q > 0 \ 1/p + 1/q = 1
    assumes finite A
    shows |\sum x \in A. |f | x * g | x| \le (\sum x \in A. |f | x| | powr | p) | powr | (1/p) * (\sum x \in A. |g | x|
powr \ q) \ powr \ (1/q)
\langle proof \rangle
lemma real-count-list-pos:
    assumes x \in set \ as
    shows real (count-list as x) > 0
    \langle proof \rangle
lemma fk-estimate:
    assumes as \neq []
   shows length as * of-rat (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1 \ / \ real \ k) * (of-rat \ (F(2*k-1) \ as) \le n \ powr(1-1) \ po
k \ as))^2
    (is ?lhs \leq ?rhs)
\langle proof \rangle
\mathbf{definition}\ \mathit{result}
    where result a = of-nat (length as) * of-nat (Suc (snd a) ^k - snd a ^k)
lemma result-exp-1:
    assumes as \neq []
    shows expectation result = real-of-rat (F k as)
\langle proof \rangle
\mathbf{lemma}\ \mathit{result-var-1}\colon
    assumes as \neq []
    shows variance result \leq (of\text{-rat }(F \ k \ as))^2 * k * n \ powr \ (1 - 1 \ / \ real \ k)
```

```
\langle proof \rangle
theorem fk-alg-sketch:
  assumes as \neq []
  shows fold (\lambda a state. state \gg fk-update a) as (fk-init k \delta \varepsilon n) =
     map-pmf (\lambda x. (s_1, s_2, k. length as, x)) M_2 (is ?lhs = ?rhs)
\langle proof \rangle
definition mean-rv
  where mean-rv \omega i_2 = (\sum i_1 = 0... < s_1. result (\omega (i_1, i_2))) / of-nat s_1
definition median-rv
    where median-rv \omega = median \ s_2 \ (\lambda i_2. \ mean-rv \ \omega \ i_2)
lemma fk-alg-correct':
  defines M \equiv fold \ (\lambda a \ state. \ state \gg fk\text{-update } a) \ as \ (fk\text{-init} \ k \ \delta \ \epsilon \ n) \gg fk\text{-result}
  shows \mathcal{P}(\omega \text{ in measure-pmf } M. |\omega - F k \text{ as}| \leq \delta * F k \text{ as}) \geq 1 - \text{of-rat } \varepsilon
\langle proof \rangle
lemma fk-exact-space-usage':
  defines M \equiv fold \ (\lambda a \ state. \ state \gg fk-update a) as (fk-init k \ \delta \ \varepsilon \ n)
  shows AE \omega in M. bit-count (encode-fk-state \omega) \leq fk-space-usage (k, n, length
    (is AE \omega in M. (- \leq ?rhs))
\langle proof \rangle
end
Main results of this section:
theorem fk-alg-correct:
  assumes k \geq 1
  assumes \varepsilon \in \{0 < .. < 1\}
  assumes \delta > 0
  assumes set \ as \subseteq \{.. < n\}
  defines M \equiv fold \ (\lambda a \ state. \ state \gg fk\text{-update } a) \ as \ (fk\text{-init} \ k \ \delta \ \varepsilon \ n) \gg fk\text{-result}
  shows \mathcal{P}(\omega \text{ in measure-pmf } M. |\omega - F \text{ } k \text{ } as| \leq \delta * F \text{ } k \text{ } as) \geq 1 - of\text{-rat } \varepsilon
  \langle proof \rangle
\textbf{theorem} \textit{ fk-exact-space-usage} :
  assumes k \geq 1
  assumes \varepsilon \in \{0 < .. < 1\}
  assumes \delta > \theta
  assumes set as \subseteq \{..< n\}
  defines M \equiv fold (\lambda a \ state. \ state \gg fk-update a) as (fk-init k \ \delta \ \varepsilon \ n)
  shows AE \omega in M. bit-count (encode-fk-state \omega) \leq fk-space-usage (k, n, length
as, \varepsilon, \delta)
  \langle proof \rangle
```

**theorem** fk-asymptotic-space-complexity:

```
 \begin{array}{l} \mathit{fk\text{-}space\text{-}usage} \in \\ O[\mathit{at\text{-}top} \times_F \mathit{at\text{-}top} \times_F \mathit{at\text{-}top} \times_F \mathit{at\text{-}right} (0 :: \mathit{rat}) \times_F \mathit{at\text{-}right} (0 :: \mathit{rat})](\lambda \ (k, \ n, \ m, \ \varepsilon, \ \delta). \\ \mathit{real} \ k * \mathit{real} \ \mathit{n} \ \mathit{powr} \ (1 - 1 / \mathit{real} \ k) \ / \ (\mathit{of\text{-}rat} \ \delta)^2 * (\mathit{ln} \ (1 \ / \mathit{of\text{-}rat} \ \varepsilon)) * (\mathit{ln} \ (\mathit{real} \ n) + \mathit{ln} \ (\mathit{real} \ m))) \\ (\mathit{is} \ - \in O[?F](?\mathit{rhs})) \\ \langle \mathit{proof} \rangle \\ \end{array}
```

end

## 9 Tutorial on the use of Pseudorandom-Objects

```
theory Tutorial-Pseudorandom-Objects
imports
Universal-Hash-Families.Pseudorandom-Objects-Hash-Families
Expander-Graphs.Pseudorandom-Objects-Expander-Walks
Equivalence-Relation-Enumeration.Equivalence-Relation-Enumeration
Median-Method.Median
Concentration-Inequalities.Bienaymes-Identity
Frequency-Moments.Frequency-Moments
```

begin

This section is a tutorial for the use of pseudorandom objects. Starting from the approximation algorithm for the second frequency moment by Alon et al. [1], we will improve the solution until we achieve a space complexity of  $\mathcal{O}(\ln n + \varepsilon^{-2} \ln(\delta^{-1}) \ln m)$ , where n denotes the range of the stream elements, m denotes the length of the stream,  $\varepsilon$  denotes the desired accuracy and  $\delta$  denotes the desired failure probability.

The construction relies on a combination of pseudorandom object, in particular an expander walk and two chained hash families.

```
hide-const (open) topological-space-class.discrete
hide-const (open) Abstract-Rewriting.restrict
hide-fact (open) Abstract-Rewriting.restrict-def
hide-fact (open) Henstock-Kurzweil-Integration.integral-cong
hide-fact (open) Henstock-Kurzweil-Integration.integral-mult-right
hide-fact (open) Henstock-Kurzweil-Integration.integral-diff
```

The following lemmas show a one-side and two-sided Chernoff-bound for  $\{0,1\}$ -valued independent identically distributed random variables. This to show the similarity with expander walks, for which similar bounds can be established: expander-chernoff-bound-one-sided and expander-chernoff-bound.

```
\mathbf{lemma}\ classic\text{-}chernoff\text{-}bound\text{-}one\text{-}sided:
```

```
fixes l:: nat assumes AE \ x in measure-pmf p.\ f\ x \in \{0,1::real\} assumes (\int x.\ f\ x\ \partial p) \le \mu\ l > 0\ \gamma \ge 0 shows measure (prod\text{-}pmf\ \{0...< l\}\ (\lambda\text{--}.\ p))\ \{w.\ (\sum i< l.\ f\ (w\ i))/l-\mu \ge \gamma\} \le exp\ (-2*real\ l*\gamma^2)
```

```
(is ?L \leq ?R)
\langle proof \rangle
lemma classic-chernoff-bound:
  assumes AE x in measure-pmf p. f x \in \{0,1::real\} l > (0::nat) \gamma \geq 0
  defines \mu \equiv (\int x. f x \partial p)
  shows measure (prod-pmf \{0..< l\} (\lambda-. p)) \{w. | (\sum i < l. f(w i))/l - \mu| \ge \gamma\} \le
2*exp (-2*real l*\gamma^2)
    (is ?L \leq ?R)
\langle proof \rangle
Definition of the second frequency moment of a stream.
definition F2 :: 'a \ list \Rightarrow real \ \mathbf{where}
  F2 \ xs = (\sum x \in set \ xs. \ (of-nat \ (count-list \ xs \ x)^2))
lemma prime-power-ls: is-prime-power (pro-size (\mathcal{L} [-1, 1]))
\langle proof \rangle
lemma prime-power-h2: is-prime-power (pro-size (\mathcal{H} \not 4 n (\mathcal{L} [-1, 1::real])))
  \langle proof \rangle
abbreviation \Psi where \Psi \equiv pmf-of-set \{-1,1::real\}
lemma f2-exp:
  assumes finite (set-pmf p)
  assumes \bigwedge I. I \subseteq \{0... < n\} \Longrightarrow card \ I \le 4 \Longrightarrow map-pmf \ (\lambda x. \ (\lambda i \in I. \ x \ i)) \ p =
prod-pmf\ I\ (\lambda-. \Psi)
  assumes set xs \subseteq \{0..< n:: nat\}
  shows (\int h. (\sum x \leftarrow xs. \ h \ x)^2 \ \partial p) = F2 \ xs \ (is \ ?L = ?R)
\langle proof \rangle
lemma f2-exp-sq:
  assumes finite (set-pmf p)
  assumes \bigwedge I. \ I \subseteq \{0... < n\} \Longrightarrow card \ I \le 4 \Longrightarrow map-pmf \ (\lambda x. \ (\lambda i \in I. \ x \ i)) \ p = 0... < n\}
prod-pmf\ I\ (\lambda-. \Psi)
  assumes set xs \subseteq \{0..< n:: nat\}
  shows (\int h. ((\sum x \leftarrow xs. \ h \ x)^2)^2 \ \partial p) \le 3 * F2 \ xs^2 \ (is \ ?L \le ?R)
\langle proof \rangle
lemma f2-var:
  assumes finite\ (set\text{-}pmf\ p)
  assumes \bigwedge I. I \subseteq \{0... < n\} \Longrightarrow card \ I \le 4 \Longrightarrow map-pmf \ (\lambda x. \ (\lambda i \in I. \ x \ i)) \ p =
prod-pmf I (\lambda-. \Psi)
  assumes set \ xs \subseteq \{0..< n:: nat\}
  shows measure-pmf.variance p(\lambda h. (\sum x \leftarrow xs. h. x)^2) \le 2 * F2 xs^2
    (is ?L \leq ?R)
\langle proof \rangle
```

lemma

```
assumes s \in set\text{-pmf} (\mathcal{H}_P \not \downarrow n (\mathcal{L} [-1,1]))
  assumes set xs \subseteq \{\theta ... < n\}
  shows f2-exp-hp: (\int h. (\sum x \leftarrow xs. \ h. x)^2 \partial sample-pro \ s) = F2 \ xs \ (is ?T1)
    and f2-exp-sq-hp: (\int h. ((\sum x \leftarrow xs. h x)^2)^2 \partial sample-pro s) \leq 3* F2 xs^2
   and f2-var-hp: measure-pmf.variance s (\lambda h. (\sum x \leftarrow xs. h. x)^2) \leq 2* F2 xs^2
(is ?T3)
\langle proof \rangle
lemmas f2-exp-h = f2-exp-hp[OF\ hash-pro-in-hash-pro-pmf[OF\ prime-power-ls]]
lemmas f2-var-h = f2-var-hp[OF hash-pro-in-hash-pro-pmf[OF prime-power-ls]]
lemma F2-definite:
  assumes xs \neq []
 shows F2 xs > 0
\langle proof \rangle
The following algorithm uses a completely random function, accordingly it
requires a lot of space: \mathcal{O}(n + \ln m).
fun example-1 :: nat \Rightarrow nat \ list \Rightarrow real \ pmf
  where example-1 n xs =
   do \{
     h \leftarrow prod\text{-}pmf \{0..< n\} (\lambda\text{-. }pmf\text{-}of\text{-}set \{-1,1::real\});
     return-pmf ((\sum x \leftarrow xs. \ h \ x)^2)
   }
lemma example-1-correct:
  assumes set xs \subseteq \{\theta ... < n\}
 shows
    measure-pmf.expectation (example-1 n xs) id = F2 xs (is ?L1 = ?R1)
    measure-pmf.variance (example-1 n xs) id \le 2 * F2 xs^2 (is ?L2 \le ?R2)
\langle proof \rangle
This version replaces a the use of completely random function with a pseu-
dorandom object, it requires a lot less space: \mathcal{O}(\ln n + \ln m).
fun example-2 :: nat \Rightarrow nat \ list \Rightarrow real \ pmf
  where example-2 n xs =
    do {
      h \leftarrow sample-pro (\mathcal{H} \not i n (\mathcal{L} [-1,1]));
     return-pmf ((\sum x \leftarrow xs. \ h \ x)^2)
lemma example-2-correct:
  assumes set xs \subseteq \{\theta ... < n\}
 shows
    measure-pmf.expectation (example-2 n xs) id = F2 xs (is ?L1 = ?R1)
    measure-pmf.variance (example-2 n xs) id < 2 * F2 xs^2 (is ?L2 < ?R2)
\langle proof \rangle
```

The following version replaces the deterministic construction of the pseudorandom object with a randomized one. This algorithm is much faster, but the correctness proof is more difficult.

```
fun example-3 :: nat \Rightarrow nat \ list \Rightarrow real \ pmf
  where example-3 n xs =
       h \leftarrow sample-pro = << \mathcal{H}_P \not \downarrow n \ (\mathcal{L} \ [-1,1]);
      return-pmf ((\sum x \leftarrow xs. \ h \ x)^2)
lemma
  assumes set xs \subseteq \{\theta ... < n\}
  shows
    measure-pmf.expectation (example-3 n xs) id = F2 xs (is ?L1 = ?R1)
     measure-pmf.variance (example-3 n xs) id \le 2 * F2 xs^2 (is ?L2 \le ?R2)
\langle proof \rangle
context
  fixes \varepsilon \delta :: real
  assumes \varepsilon-qt-\theta: \varepsilon > \theta
  assumes \delta-range: \delta \in \{0 < ... < 1\}
By using the mean of many independent parallel estimates the following
algorithm achieves a relative accuracy of \varepsilon, with probability \frac{3}{4}. It requires
\mathcal{O}(\varepsilon^{-2}(\ln n + \ln m)) bits of space.
fun example-4 :: nat \Rightarrow nat \ list \Rightarrow real \ pmf
  where example-4 n xs =
       let s = nat [8 / \varepsilon^2];
      \begin{array}{l} h \leftarrow \textit{prod-pmf} \ \{0... < s\} \ (\lambda\text{-. sample-pro} \ (\mathcal{H} \ \textit{4} \ n \ (\mathcal{L} \ [-1,1]))); \\ \textit{return-pmf} \ ((\sum j < s. \ (\sum x \leftarrow xs. \ h \ j \ x)^2)/s) \end{array}
lemma example-4-correct-aux:
  assumes set xs \subseteq \{0..< n\}
  defines s \equiv nat [8 / \varepsilon^2]
  defines R \equiv (\lambda h :: nat \Rightarrow nat \Rightarrow real. (\sum j < s. (\sum x \leftarrow xs. \ h \ j \ x)^2)/real \ s)
  assumes fin: finite (set-pmf p)
   assumes indep: prob-space.k-wise-indep-vars (measure-pmf p) 2 (\lambda-. discrete)
(\lambda i \ x. \ x \ i) \ \{... < s\}
  assumes comp: \Lambda i. \ i < s \Longrightarrow map-pmf \ (\lambda x. \ x \ i) \ p = sample-pro \ (\mathcal{H} \ 4 \ n \ (\mathcal{L}
  shows measure p\{h. |Rh - F2xs| > \varepsilon * F2xs\} \le 1/4 (is ?L \le ?R)
\langle proof \rangle
lemma example-4-correct:
  assumes set xs \subseteq \{0...< n\}
```

```
shows \mathcal{P}(\omega \text{ in example-4 } n \text{ xs. } |\omega - F2 \text{ xs}| > \varepsilon * F2 \text{ xs}) \leq 1/4 \text{ (is } ?L \leq ?R) \langle proof \rangle
```

Instead of independent samples, we can choose the seeds using a second pair-wise independent pseudorandom object. This algorithm requires only  $\mathcal{O}(\ln n + \varepsilon^{-2} \ln m)$  bits of space.

```
fun example-5 :: nat \Rightarrow nat \ list \Rightarrow real \ pmf
  where example-5 n xs =
     do {
        let s = nat [8 / \varepsilon^2];
        h \leftarrow sample-pro (\mathcal{H} \ 2 \ s \ (\mathcal{H} \ 4 \ n \ (\mathcal{L} \ [-1,1]))); return-pmf ((\sum j < s. \ (\sum x \leftarrow xs. \ h \ j \ x)^2)/s)
lemma example-5-correct-aux:
  assumes set xs \subseteq \{0..< n\}
  defines s \equiv nat \lceil 8 / \varepsilon^2 \rceil
  defines R \equiv (\lambda h :: nat \Rightarrow nat \Rightarrow real. (\sum j < s. (\sum x \leftarrow xs. \ h \ j \ x)^2)/real \ s)
  shows measure (sample-pro (\mathcal{H}\ 2\ s\ (\mathcal{H}\ 4\ n\ (\mathcal{L}\ [-1,1])))) {h. |R\ h\ -\ F2\ xs|>\varepsilon
* F2 xs} \leq 1/4
\langle proof \rangle
lemma example-5-correct:
  assumes set xs \subseteq \{\theta ... < n\}
  shows \mathcal{P}(\omega \text{ in example-5 n xs. } |\omega - F2 \text{ xs}| > \varepsilon * F2 \text{ xs}) \leq 1/4 \text{ (is } ?L \leq ?R)
\langle proof \rangle
```

The following algorithm improves on the previous one, by achieving a success probability of  $\delta$ . This works by taking the median of  $\mathcal{O}(\ln(\delta^{-1}))$  parallel independent samples. It requires  $\mathcal{O}(\ln(\delta^{-1})(\ln n + \varepsilon^{-2} \ln m))$  bits of space.

```
fun example-6:: nat \Rightarrow nat \ list \Rightarrow real \ pmf where example-6 \ n \ xs = do \ \{ let \ s = nat \ \lceil 8 \ / \ \varepsilon \widehat{\ }^2 \rceil; \ let \ t = nat \ \lceil 8 \ * \ ln \ (1/\delta) \rceil; \\ h \leftarrow prod-pmf \ \{0..< t\} \ (\lambda -. \ sample-pro \ (\mathcal{H} \ 2 \ s \ (\mathcal{H} \ 4 \ n \ (\mathcal{L} \ [-1,1])))); \\ return-pmf \ (median \ t \ (\lambda i. \ ((\sum j < s. \ (\sum x \leftarrow xs. \ h \ i \ j \ x) \widehat{\ }^2)/\ s)))) \ \}
\mathbf{lemma} \ example-6-correct: \\ \mathbf{assumes} \ set \ xs \subseteq \{0..< n\} \\ \mathbf{shows} \ \mathcal{P}(\omega \ in \ example-6 \ n \ xs. \ |\omega - F2 \ xs| > \varepsilon \ * F2 \ xs) \le \delta \ (\mathbf{is} \ ?L \le ?R) \\ /nroof \ \rangle
```

The following algorithm uses an expander random walk, instead of independent samples. It requires only  $\mathcal{O}(\ln n + \ln(\delta^{-1})\varepsilon^{-2}\ln m)$  bits of space.

```
fun example-7 :: nat \Rightarrow nat \ list \Rightarrow real \ pmf

where example-7 n \ xs =

do \ \{
```

```
let s = nat \lceil 8 / \varepsilon \widehat{\ }^2 \rceil; let t = nat \lceil 32 * ln (1/\delta) \rceil; h \leftarrow sample-pro (\mathcal{E} \ t (1/8) \ (\mathcal{H} \ 2 \ s \ (\mathcal{H} \ 4 \ n \ (\mathcal{L} \ [-1,1])))); return-pmf (median t \ (\lambda i. \ ((\sum j < s. \ (\sum x \leftarrow xs. \ h \ i \ j \ x) \widehat{\ }^2)/\ s))))} lemma example-7-correct: assumes set xs \subseteq \{0...< n\} shows \mathcal{P}(\omega \ in \ example-7 \ n \ xs. \ |\omega - F2 \ xs| > \varepsilon * F2 \ xs) \le \delta \ (\mathbf{is} \ ?L \le ?R) \langle proof \rangle end
```

# A Informal proof of correctness for the $F_0$ algorithm

This appendix contains a detailed informal proof for the new Rounding-KMV algorithm that approximates  $F_0$  introduced in Section 6 for reference. It follows the same reasoning as the formalized proof.

Because of the amplification result about medians (see for example [1, §2.1]) it is enough to show that each of the estimates the median is taken from is within the desired interval with success probability  $\frac{2}{3}$ . To verify the latter, let  $a_1, \ldots, a_m$  be the stream elements, where we assume that the elements are a subset of  $\{0, \ldots, n-1\}$  and  $0 < \delta < 1$  be the desired relative accuracy. Let p be the smallest prime such that  $p \ge \max(n, 19)$  and let p be a random polynomial over F(p) with degree strictly less than 2. The algorithm also introduces the internal parameters p defined by:

$$t := \lceil 80\delta^{-2} \rceil \qquad \qquad r := 4\log_2 \lceil \delta^{-1} \rceil + 23$$

The estimate the algorithm obtains is R, defined using:

$$H := \{ \lfloor h(a) \rfloor_r | a \in A \} \qquad R := \begin{cases} tp \left( \min_t(H) \right)^{-1} & \text{if } |H| \ge t \\ |H| & \text{othewise,} \end{cases}$$

where  $A := \{a_1, \ldots, a_m\}$ ,  $\min_t(H)$  denotes the *t*-th smallest element of H and  $\lfloor x \rfloor_r$  denotes the largest binary floating point number smaller or equal to x with a mantissa that requires at most r bits to represent. With these definitions, it is possible to state the main theorem as:

$$P(|R - F_0| \le \delta |F_0|) \ge \frac{2}{3}.$$

which is shown separately in the following two subsections for the cases  $F_0 \ge t$  and  $F_0 < t$ .

<sup>&</sup>lt;sup>1</sup>This rounding operation is called *truncate-down* in Isabelle, it is defined in HOL-Library.Float.

#### **A.1** Case $F_0 \geq t$

Let us introduce:

$$H^* := \{h(a)|a \in A\}^\#$$
  $R^* := tp\left(\min_t^\#(H^*)\right)^{-1}$ 

These definitions are modified versions of the definitions for H and R: The set  $H^*$  is a multiset, this means that each element also has a multiplicity, counting the number of distinct elements of A being mapped by h to the same value. Note that by definition:  $|H^*| = |A|$ . Similarly the operation  $\min_t^\#$  obtains the t-th element of the multiset H (taking multiplicities into account). Note also that there is no rounding operation  $\lfloor \cdot \rfloor_r$  in the definition of  $H^*$ . The key reason for the introduction of these alternative versions of H, R is that it is easier to show probabilistic bounds on the distances  $|R^* - F_0|$  and  $|R^* - R|$  as opposed to  $|R - F_0|$  directly. In particular the plan is to show:

$$P(|R^* - F_0| > \delta' F_0) \le \frac{2}{9}, \text{ and}$$
 (1)

$$P\left(|R^* - F_0| \le \delta' F_0 \wedge |R - R^*| > \frac{\delta}{4} F_0\right) \le \frac{1}{9}$$
 (2)

where  $\delta' := \frac{3}{4}\delta$ . I.e. the probability that  $R^*$  has not the relative accuracy of  $\frac{3}{4}\delta$  is less that  $\frac{2}{9}$  and the probability that assuming  $R^*$  has the relative accuracy of  $\frac{3}{4}\delta$  but that R deviates by more that  $\frac{1}{4}\delta F_0$  is at most  $\frac{1}{9}$ . Hence, the probability that neither of these events happen is at least  $\frac{2}{3}$  but in that case:

$$|R - F_0| \le |R - R^*| + |R^* - F_0| \le \frac{\delta}{4} F_0 + \frac{3\delta}{4} F_0 = \delta F_0.$$
 (3)

Thus we only need to show Equation 1 and 2. For the verification of Equation 1 let

$$Q(u) = |\{h(a) < u \mid a \in A\}|$$

and observe that  $\min_t^\#(H^*) < u$  if  $Q(u) \ge t$  and  $\min_t^\#(H^*) \ge v$  if  $Q(v) \le t-1$ . To see why this is true note that, if at least t elements of A are mapped by h below a certain value, then the t-smallest element must also be within them, and thus also be below that value. And that the opposite direction of this conclusion is also true. Note that this relies on the fact that  $H^*$  is a multiset and that multiplicities are being taken into account, when computing the t-th smallest element. Alternatively, it is also possible to write  $Q(u) = \sum_{a \in A} 1_{\{h(a) < u\}}^2$ , i.e., Q is a sum of pairwise independent  $\{0,1\}$ -valued random variables, with expectation  $\frac{u}{p}$  and variance  $\frac{u}{p} - \frac{u^2}{p^2}$ .

<sup>&</sup>lt;sup>2</sup>The notation  $1_A$  is shorthand for the indicator function of A, i.e.,  $1_A(x) = 1$  if  $x \in A$  and 0 otherwise.

<sup>3</sup> Using linearity of expectation and Bienaymé's identity, it follows that  $\operatorname{Var} Q(u) \leq \operatorname{E} Q(u) = |A|up^{-1} = F_0up^{-1}$  for  $u \in \{0, \dots, p\}$ . For  $v = \left| \frac{tp}{(1-\delta')F_0} \right|$  it is possible to conclude:

$$t-1 \leq \frac{4}{(1-\delta')} - 3\sqrt{\frac{t}{(1-\delta')}} - 1 \leq \frac{F_0v}{p} - 3\sqrt{\frac{F_0v}{p}} \leq \mathrm{E}Q(v) - 3\sqrt{\mathrm{Var}Q(v)}$$

and thus using Tchebyshev's inequality:

$$P\left(R^* < (1 - \delta') F_0\right) = P\left(\operatorname{rank}_t^{\#}(H^*) > \frac{tp}{(1 - \delta') F_0}\right)$$

$$\leq P(\operatorname{rank}_t^{\#}(H^*) \geq v) = P(Q(v) \leq t - 1) \qquad (4)$$

$$\leq P\left(Q(v) \leq \operatorname{E}Q(v) - 3\sqrt{\operatorname{Var}Q(v)}\right) \leq \frac{1}{9}.$$

Similarly for  $u = \left\lceil \frac{tp}{(1+\delta')F_0} \right\rceil$  it is possible to conclude:

$$t \geq \frac{t}{(1+\delta')} + 3\sqrt{\frac{t}{(1+\delta')} + 1} + 1 \geq \frac{F_0u}{p} + 3\sqrt{\frac{F_0u}{p}} \geq \mathrm{E}Q(u) + 3\sqrt{\mathrm{Var}Q(v)}$$

and thus using Tchebyshev's inequality:

$$P\left(R^* > \left(1 + \delta'\right) F_0\right) = P\left(\operatorname{rank}_t^{\#}(H^*) < \frac{tp}{(1 + \delta') F_0}\right)$$

$$\leq P(\operatorname{rank}_t^{\#}(H^*) < u) = P(Q(u) \geq t)$$

$$\leq P\left(Q(u) \geq \operatorname{E}Q(u) + 3\sqrt{\operatorname{Var}Q(u)}\right) \leq \frac{1}{9}.$$
(5)

Note that Equation 4 and 5 confirm Equation 1. To verfiy Equation 2, note that

$$\min_{t}(H) = \lfloor \min_{t}^{\#}(H^*) \rfloor_{r} \tag{6}$$

if there are no collisions, induced by the application of  $\lfloor h(\cdot) \rfloor_r$  on the elements of A. Even more carefully, note that the equation would remain true, as long as there are no collision within the smallest t elements of  $H^*$ . Because Equation 2 needs to be shown only in the case where  $R^* \geq (1 - \delta') F_0$ , i.e., when  $\min_t^\#(H^*) \leq v$ , it is enough to bound the probability of a collision in the range [0; v]. Moreover Equation 6 implies  $|\min_t(H) - \min_t^\#(H^*)| \leq \max(\min_t^\#(H^*), \min_t(H))2^{-r}$  from which it is possible to derive  $|R^* - R| \leq \frac{\delta}{4}F_0$ . Another important fact is that h is injective with probability  $1 - \frac{1}{p}$ ,

 $<sup>^{3}</sup>$ A consequence of h being chosen uniformly from a 2-independent hash family.

<sup>&</sup>lt;sup>4</sup>The verification of this inequality is a lengthy but straightforward calculcation using the definition of  $\delta'$  and t.

this is because h is chosen uniformly from the polynomials of degree less than 2. If it is a degree 1 polynomial it is a linear function on GF(p) and thus injective. Because  $p \ge 18$  the probability that h is not injective can be bounded by 1/18. With these in mind, we can conclude:

$$P\left(|R^* - F_0| \le \delta' F_0 \wedge |R - R^*| > \frac{\delta}{4} F_0\right)$$

$$\le P\left(R^* \ge (1 - \delta') F_0 \wedge \min_t^\# (H^*) \ne \min_t(H) \wedge h \text{ inj.}\right) + P(\neg h \text{ inj.})$$

$$\le P\left(\exists a \ne b \in A. \lfloor h(a) \rfloor_r = \lfloor h(b) \rfloor_r \le v \wedge h(a) \ne h(b)\right) + \frac{1}{18}$$

$$\le \frac{1}{18} + \sum_{a \ne b \in A} P\left(\lfloor h(a) \rfloor_r = \lfloor h(b) \rfloor_r \le v \wedge h(a) \ne h(b)\right)$$

$$\le \frac{1}{18} + \sum_{a \ne b \in A} P\left(|h(a) - h(b)| \le v2^{-r} \wedge h(a) \le v(1 + 2^{-r}) \wedge h(a) \ne h(b)\right)$$

$$\le \frac{1}{18} + \sum_{a \ne b \in A} \sum_{\substack{a',b' \in \{0,\dots,p-1\} \wedge a' \ne b' \\ |a'-b'| \le v2^{-r} \wedge a' \le v(1+2^{-r})}} P(h(a) = a') P(h(b) = b')$$

$$\le \frac{1}{18} + \frac{5F_0^2 v^2}{2p^2} 2^{-r} \le \frac{1}{9}.$$

which shows that Equation 2 is true.

#### **A.2** Case $F_0 < t$

Note that in this case  $|H| \leq F_0 < t$  and thus R = |H|, hence the goal is to show that:  $P(|H| \neq F_0) \leq \frac{1}{3}$ . The latter can only happen, if there is a collision induced by the application of  $\lfloor h(\cdot) \rfloor_r$ . As before h is not injective

with probability at most  $\frac{1}{18}$ , hence:

$$P(|R - F_0| > \delta F_0) \leq P(R \neq F_0)$$

$$\leq \frac{1}{18} + P(R \neq F_0 \land h \text{ inj.})$$

$$\leq \frac{1}{18} + P(\exists a \neq b \in A. \lfloor h(a) \rfloor_r = \lfloor h(b) \rfloor_r \land h \text{ inj.})$$

$$\leq \frac{1}{18} + \sum_{a \neq b \in A} P(\lfloor h(a) \rfloor_r = \lfloor h(b) \rfloor_r \land h(a) \neq h(b))$$

$$\leq \frac{1}{18} + \sum_{a \neq b \in A} P(|h(a) - h(b)| \leq p2^{-r} \land h(a) \neq h(b))$$

$$\leq \frac{1}{18} + \sum_{a \neq b \in A} \sum_{\substack{a',b' \in \{0,\dots,p-1\}\\ a' \neq b' \land |a'-b'| \leq p2^{-r}}} P(h(a) = a')P(h(b) = b')$$

$$\leq \frac{1}{18} + F_0^2 2^{-r+1} \leq \frac{1}{18} + t^2 2^{-r+1} \leq \frac{1}{9}.$$

Which concludes the proof.

#### References

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