Optimal Dynamic Parameterized Subset Sampling

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Problem Overview

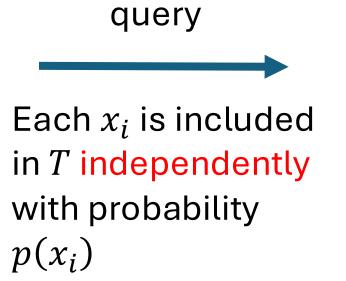
Subset Sampling

Given a set of n distinct items $S = \{x_1, ..., x_n\}$, in which each item x_i has an associated probability $p(x_i)$, a query for the subset sampling problem returns a subset $T \subseteq S$, such that every x_i is independently included in T w.p. $p(x_i)$.

 $\sum_{x_1, \dots, x_n} x_1 \dots x_n$ $p(x_1) \quad p(x_2) \quad p(x_3) \quad p(x_4) \qquad p(x_n)$ $\sum_{x_1, \dots, x_n} p(x_i) = \mu$ Each in T in with $p(x_i)$



 x_1



Best-known Result:

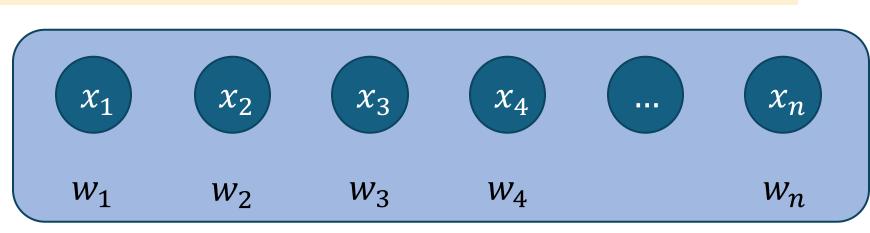
- query time: $O(1 + \mu)$
- update time: O(1) (insert/deletion events)

static setting [Bringmann & Panagiotou, 2017]

[Bhattacharya, Kiss, Sidford, Wajc, 2024] [Yi, Wang, Wei, 2023]

space complexity: O(n)

Parameterized Subset Sampling



$$p(x_i) = \min\left\{1, \frac{w_i}{W_S(\alpha, \beta)}\right\},\$$

$$W_S(\alpha, \beta) = \alpha \sum_{i=1}^n w_i + \beta$$

 $\succ w_i$: non-negative integer weights

 α , β are input parameters, and can be given on-the-fly.

Our contributions in the Word RAM model:

- query time: $0(1 + \mu)$ in expectation
- update time (insert/delete): 0(1) worst case
- space complexity: O(n) words for worst-case at all times
- preprocessing time: O(n) worst-case

Other Results

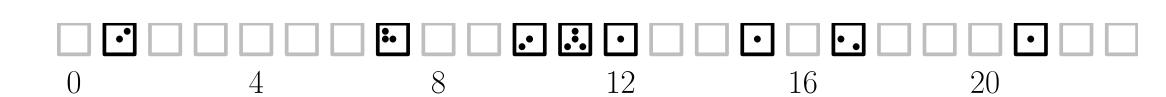
- Senerating geometric variates: Let $p \in (0,1)$ be a rational number which can be represented by a O(1)-word integer nominator and O(1)-word integer denominator. We can generate the following random variates in O(1) expected time with O(n) worst-case space.
- Bernoulli(p) [Bringmann & Friedrich, 2013]
- Bernoulli $\left(\frac{1-(1-p)^n}{pn}\right)$ (new by us)
- Bound Geometric BG(p, n): $\Pr\{BG(p, n) = i\} = \begin{cases} p(1-p)^{i-1}, i \in [1, n) \\ (1-p)^{n-1}, i = n \end{cases}$ [Bringmann & Friedrich, 2013]
- Truncated Geometric TG(p,n) (new by us): $\Pr\{TG(p,n)=i\}=\frac{p(1-p)^{i-1}}{1-(1-p)^n}$

> Hardness result:

- Integer Sorting can be reduced to the deletion-only Dynamic Parameterized Subset Sampling (DPSS) with float weights.
- Optimal deletion-only DPSS with float weights implies a O(N)-expected-time algorithm for sorting N integers in the word RAM model with $\Omega(\log N)$ bit length. The latter remains an open problem.

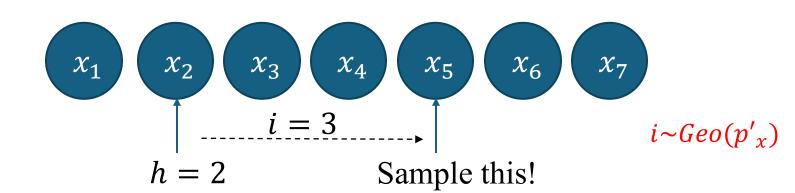
Techniques

Bucketing-Based Algorithm: A Warm-Up



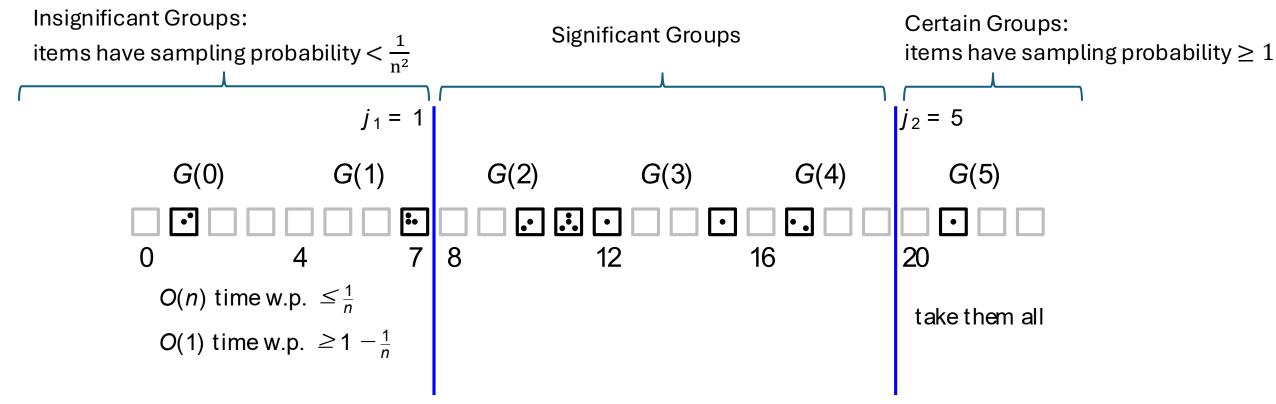
➤ Organize items into power-of-two buckets:

- lacktriangle Bucket B(i) contains items with weights in $\left[2^{i}, 2^{i+1}\right]$
- For each non-empty bucket B(i):
 - Sample potential items using upper-bound probability $p'_{\chi} = \min \left\{ 1, \frac{2^{i+1}}{W_S(\alpha, \beta)} \right\}$
- lacktriangle Accept each potential item x with probability $\frac{p_x}{p_{Ix}}$
- \triangleright Query time: $O(b + \mu)$, where b is the number of non-empty buckets



Key idea: Avoid touching all buckets!

ightharpoonupOur solution: Partition the buckets into **groups!** $\log n$ buckets.

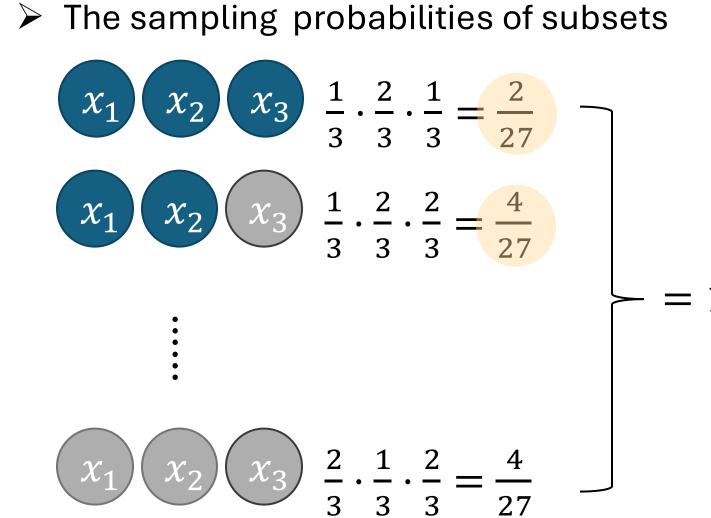


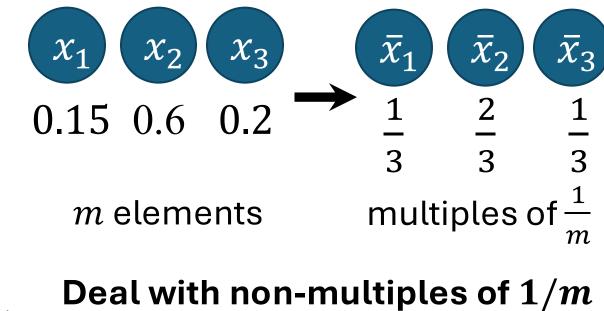
Handling Significant Groups

- ◆ Step 1: Find the potential buckets, those containing at least one potential item:
 - This is another subset sampling problem! Recursion!
 - After three times recursion, the problem size is $O(\log \log \log n)$.
 - Small enough to be solved in a pre-computed look-up table.
- ◆ Step 2: Sample from the potential buckets:
 - How to find first potential item index k from a potential bucket?
 - This is a conditional probability problem!
 - It follows the Truncated Geometric Distribution.

Look-up Table Trick Overview







$[mn(x_i)] \quad (1 \qquad m)$

$$\bar{p}(x_i) = \frac{[mp(x_i)]}{m} \in \left\{ \frac{1}{m}, \dots, \frac{m}{m} \right\}$$

$$\text{Accept the event with } \bar{p}(x_i) / m$$

- \triangleright Accept the event with $\bar{p}(x_i)/p(x_i)$
- Obtain a row for sampling!
- Just uniformly select an entry and return the subset as the sample result





111 | 111 | 110 | 110 | 110 | 110 | ... |





- Bringmann & Panagiotou, 2012] Bringmann K, Panagiotou K. Efficient sampling methods for discrete distributions // ICALP 2012
- [Bringmann & Friedrich, 2013] Bringmann K, Friedrich T. Exact and efficient generation of geometric random variates and random graphs //ICALP 2013
- [Bhattacharya, Kiss, Sidford, Wajc] Bhattacharya S, Kiss P, Sidford A, Wajc D. Near-optimal dynamic rounding of fractional matchings in bipartite graphs //STOC 2024
- [Yi, Wang, Wei, 2023] Yi L, Wang H, Wei Z. Optimal dynamic subset sampling: Theory and applications //KDD 2023