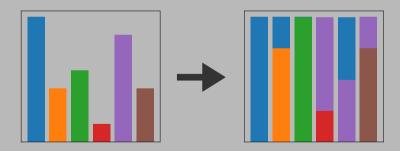
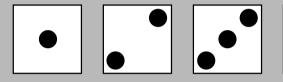
Alias Table sampling

Basile Fraboni

GDL Origami

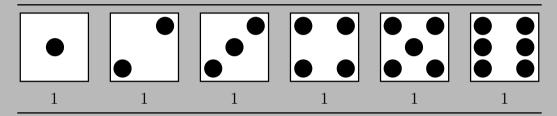




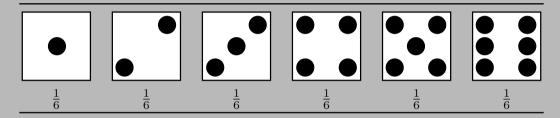








A structure to sample discrete distributions



7



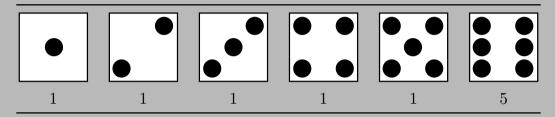


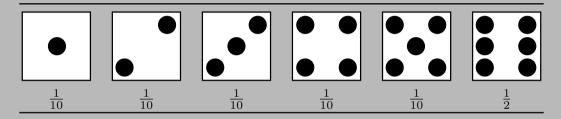


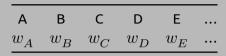




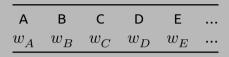




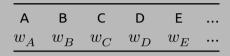




A structure to sample discrete distributions



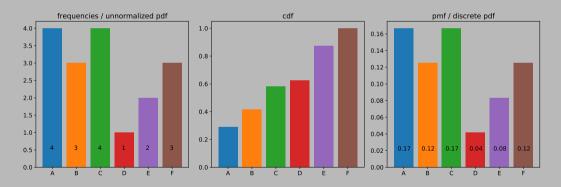
importance sampling



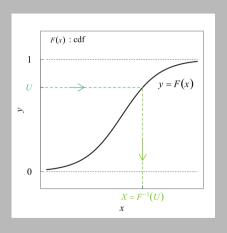
- importance sampling
- simulations, data analysis, machine learning, statistics, etc

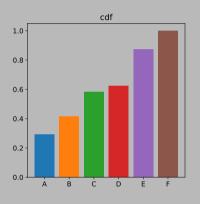
Discrete CDF



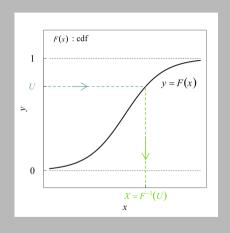


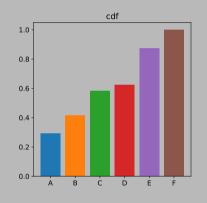
Discrete CDF inversion





Discrete CDF inversion





 $\bullet \ \ {\rm binary\ search}\ O(\log n)$

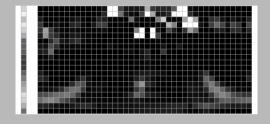
Discrete CDF implementation

```
int n:
float total:
std::vector<float> cdf;
DiscreteCDF(const std::vector<float>& values) : n(values.size()), total(0), cdf(n)
    for(int i = 0; i < n; ++i)
        total += values[i]:
       cdf[i] = total;
int sample(const float u)
   const float value = u * total:
    int p = 0, q = n - 1;
    while(p < q)
        int m = (p+q) / 2;
       if(cdf[m] < value)</pre>
           p = m + 1;
           q = m;
```

Discrete CDF

• rendering: sampling area (lights), images (environment maps), volumes, etc

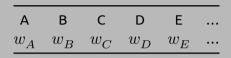




Discrete CDF

	Discrete CDF
construction	O(n)
sampling	$O(\log n)$

A structure to sample discrete distributions



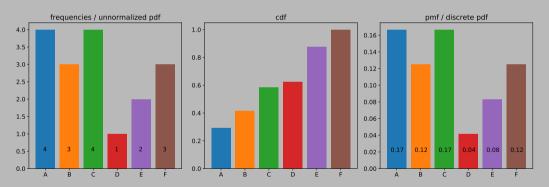
• in constant time!

Regained attention recently:

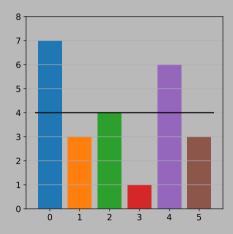
- New fast method for generating discrete random numbers with arbitrary frequency distributions, A. J. Walker, 1974
- A linear algorithm for generating random numbers with a given distribution,
 M. D. Vose, Software Engineering, 1991
- Parallel Weighted Random Sampling, L. Hübschle-Schneider and P. Sanders, 2019
 2021
- Weighted Random Sampling on GPUs, H.-P. Lehmann, L. Hübschle-Schneider and P. Sanders, 2021
- The alias method for sampling discrete distributions, C. Wyman, Ray tracing Gems 2, Chapter 21, 2021
- used in the implementation of the ReSTIR algorithm [Bitterli et al. 2020]

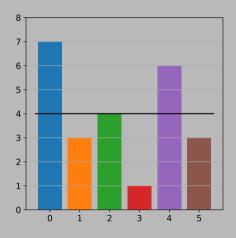
	Discrete CDF	Alias [Walker]	Alias [Vose]	Alias [Hübschle]
construction		$O(n^2)$	O(n) $O(1)$	O(n)
sampling	$O(\log n)$	O(1)	O(1)	O(1)



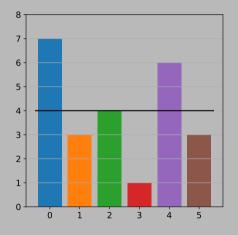


Integers are used for the example, but the construction applies for real weights – frequencies.

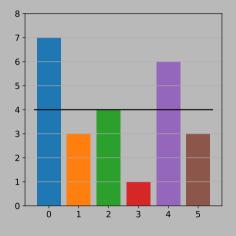




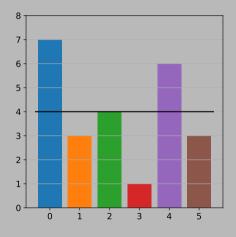
compute mean



- compute mean
- large items > mean
- small items <= mean



- compute mean
- large items > mean
- small items <= mean
- split large items and share residuals



- compute mean
- large items > mean
- small items <= mean</p>
- split large items and share residuals

```
Algorithm 2 A sweeping algorithm for building alias tables.
  Input: \langle w_1, \dots, w_n \rangle \in \mathbb{R}^n the weights of the n input items
    assume sentinel items w_{n+1} = \infty and w_{n+2} = 0 to avoid some special case treatments
  Output: b, an alias table consisting of n pairs (w, a) of (partial) weight w and alias a
1 Function sweeping Alias Table (\langle w_1, \dots, w_n \rangle)
      W := \sum_{i=1}^{n} w_i
                                                                                   - total weight
      i := \min \{k > 0 : w_k \le W/n\}
                                                                                - first light item
      i := \min \{k > 0 : w_k > W/n\}
                                                                               — first beavy item
                                                                           - current heavy item
      if i = n + 1 then \forall k = 1..n : b[k], p = w_k : b[k], a = k:

    All weights are equal

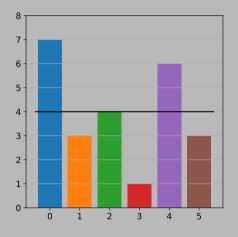
      while i \le n do
         if w > W/n then
                                                                          - Pack a light bucket.
              b[i].w := w
                                                                  - Item i completely fits here.
              b[i], a := i
                                                      - Item i fills the remainder of bucket i.
              w := (w + w_i) - W/n

    Update residual weight of item i.

11
             i := \min \{k > i : w_k \le W/n\}
                                                                                - next light item
12
13
                                                                         - Pack a heavy bucket.

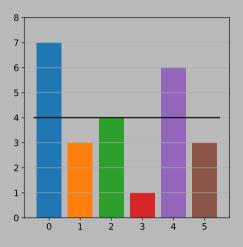
    Now item i completely fits here.

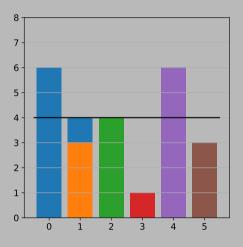
14
              b[i].w := w
              j' := \min \{k > j : w_k > W/n\}
                                                                               — next beavu item
              b[i], a := i'; i := i'
                                                                         - Proceed with item i'
              w := (w + w_{\omega}) - W/n — Compute residual weight avoiding cancellation issues
```

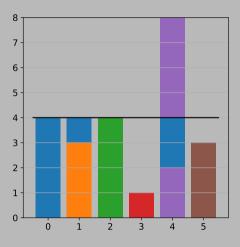


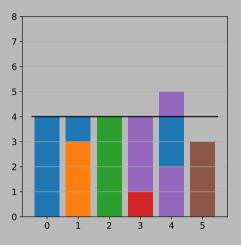
Required operations:

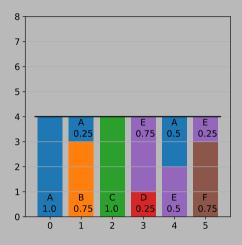
- find next small item
- find next large item
- pack small item
- pack large item

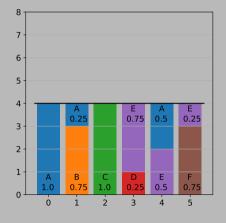




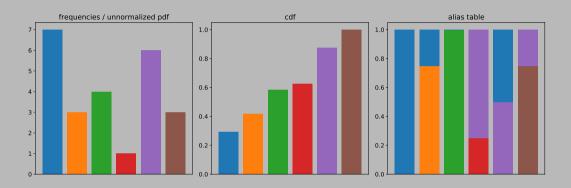








label	Α	В	С	D	E	F
split alias	1	0.75 A	1	0.25 E	0.5 A	0.75 E



Alias Table Interface

label	Α	В	С	D	Е	F
split	1	0.75	1	0.25	0.5	0.75
alias		Α		Е	Α	Е

```
int i;
int n:
std::vector<Alias> table;
AliasTable(const std::vector<float>& values);
int sample(const float u);
```

Alias Table Interface

```
AliasTable::AliasTable(const std::vector<float>& values) : n(values.size()), table(n)
   const float sum = std::accumulate( values.begin(), values.end(), 0.f ):
   const float avg = sum / n;
   std::vector<int> partition(n):
       table[i].t = values[i]/avg;
       if( table[i].t <= 1 )
           partition[lid++] = i;
           partition[hid--] = i;
   int tlid = partition[lid], thid = partition[hid], nthid:
       if( table[thid].t > 1 )
           table[tlid].i = thid:
           table[thid].t -= (1 - table[tlid].t):
           tlid = partition[++lid]:
           table[thid].i = nthid;
           table[nthid].t -= (1 - table[thid].t):
           thid = nthid;
   table[thid] = {1, std::numeric limits<int>::max()}:
```

- less than 40 loc construction
- 1 temporary array of int
- ullet 3 loops in O(n)

Alias Table Interface

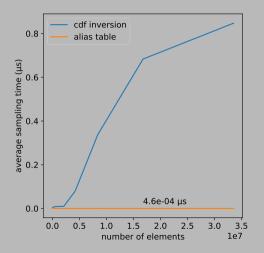
```
AliasTable::AliasTable(const std::vector<float>& values) : n(values.size()), table(n)
   const float sum = std::accumulate( values.begin(), values.end(), 0.f ):
    const float avg = sum / n;
    std::vector<int> partition(n):
    int lid = 0, hid = n-1;
       table[i].t = values[i]/avg;
       if( table[i].t <= 1 )
           partition[lid++] = i;
           partition[hid--] = i:
   int tlid = partition[lid], thid = partition[hid], nthid:
        if( table[thid].t > 1 )
            table[tlid].i = thid:
           table[thid].t -= (1 - table[tlid].t):
           tlid = partition[++lid]:
           table[thid].i = nthid:
           table[nthid].t -= (1 - table[thid].t):
            thid = nthid;
    table[thid] = {1, std::numeric limits<int>::max()}:
```

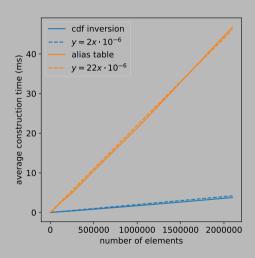
- less than 40 loc construction
- 1 temporary array of int
- ullet 3 loops in O(n)

```
int AliasTable::sample(const float u)
{
    const int id = n * u;
    const float u2 = n * u - id;
    return u2 < table[id].t ? id : table[id].i;
}</pre>
```

ullet sampling in O(1) - 3 loc

Performance

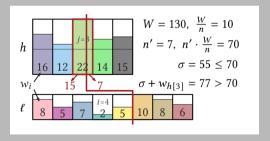




2-steps Parallel Construction

- divide and conquer approach
- ullet split the array into p subproblems
- lacktriangle a subproblem is determined by finding triplets (i,j,s) such that:

$$\sigma = \sum_{x \leq i} w_{l[x]} + \sum_{x \leq j} w_{h[x]} \leq p \cdot \frac{W}{n} \qquad \text{and} \qquad w_{h[j+1]} > p \cdot \frac{W}{n} - \sigma = s$$



[Hübschle et al. 2019]

Alias Table Sampling: takeaway

- A structure to sample discrete distributions in constant time
- Favorably compares to CDF inversion for large amounts of samples (e.g. rendering)
- Parallel construction available CPU / GPU