

Sublinear Edge Sampling and Bernoulli Factories

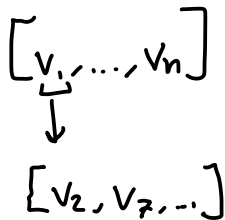
University of Toronto Theory Student Seminar
(February 2023)

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Edge Sampling

Given $G = (V, E)$, sample an edge uniformly at random using a **sublinear** number of queries

$$|V| = n, |E| = m$$



Allowed queries:

- ▶ degree of i th vertex
- ▶ j th neighbor of i th vertex
- know approx #edges m

Edge Sampling: Results

What is the expected number of queries to sample an ϵ -approximately-uniform edge?

▶ Eden and Rosenbaum (2018): $O\left(\frac{n}{\sqrt{\epsilon m}}\right)$

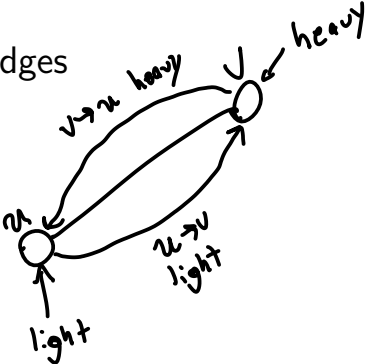
▶ Tětek and Thorup (2022): $O\left(\frac{n \log\left(\frac{1}{\epsilon}\right)}{\sqrt{m}}\right)$

▶ Lower bound: $\Omega\left(\frac{n}{\sqrt{m}}\right)$  \sqrt{m} vertices
 m edges

▶ Eden, Narayanan, and Tětek (2023): $O\left(\frac{n}{\sqrt{m}}\right)$
Exactly uniform!

Eden and Rosenbaum: Setup

Split edges into two directed edges



Degree threshold: θ

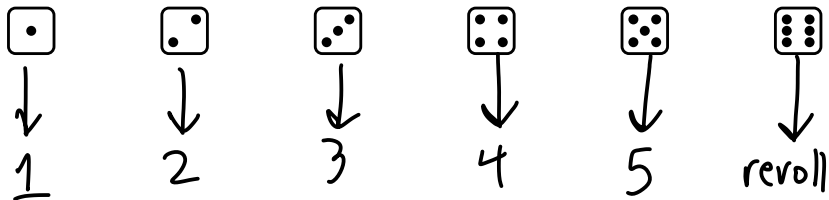
$$\sqrt{\frac{3}{\epsilon}}$$

$\text{deg}(v) \leq \theta$: v is *light*

$\text{deg}(v) > \theta$: v is *heavy*

Rejection Sampling

How do I simulate a 5-sided die if I only have a 6-sided die?



Eden and Rosenbaum: Light Edges

Sampling a light edge *exactly* uniformly

- ▶ Sample random v and reject if heavy
- ▶ Sample random edge incident to v
- ▶ Return edge with probability $\deg(v)/\theta$

For a given light edge, the probability we sampled it:

$$\frac{1}{n} \cdot \frac{1}{\deg(v)} \cdot \frac{\deg(v)}{\theta} = \frac{1}{n\theta}$$

Probability we return an edge:

$$\sum_{\text{light } v} \frac{1}{n} \frac{\deg(v)}{\theta} = \frac{\# \text{ light edges}}{n\theta}$$

Eden and Rosenbaum: Heavy Edges

Sampling a heavy edge *approximately* uniformly

- ▶ Sample a uniform random light edge $u \rightarrow v$
- ▶ If v is light, reject
- ▶ Otherwise return a random edge incident to v

For a given heavy edge, the probability we sampled it:

$$\frac{1}{n\theta} \cdot \left[\deg_x(v) \cdot \frac{1}{\deg(v)} \right]$$

light vertices Adj. to v

heavy vertices $< \frac{m}{\theta}$

$$= \frac{m}{\sqrt{\frac{m}{\epsilon}}} = \sqrt{\epsilon m}$$

Probability we return an edge:

$$\sum_{\text{heavy } e} P_r[e \text{ returned}] > \frac{(1-\epsilon) \# \text{ heavy edges}}{n\theta}$$

∈ $[(1-\epsilon), 1]$ so $\deg_x(v) > (1-\epsilon) \deg(v)$

$\deg_x(v) \leq \# \text{ heavy vertices}$
 $< \sqrt{\epsilon m} < \epsilon \cdot \deg(v)$

Eden and Rosenbaum: Final algorithm

Sampling an edge *approximately* uniformly

- ▶ Flip a fair coin
 - ▶ If heads, sample a light edge
 - ▶ If tails, sample a heavy edge
- ▶ Repeat until the above step returns an edge

Probability we return an edge on a given iteration:

$$> \frac{1}{2} \frac{\# \text{ light edges}}{n \Theta} + \frac{1}{2} \frac{(1-\epsilon) \# \text{ heavy edges}}{n \Theta} > \frac{(1-\epsilon)m}{n \Theta}$$

$\approx \frac{\sqrt{\epsilon m}}{\epsilon}$

Expected number of attempts:

$$O\left(\frac{n}{\sqrt{\epsilon m}}\right)$$

How do we get rid of ϵ ?

Sampling a light edge *exactly* uniformly

- ▶ Sample random v and reject if heavy
- ▶ Sample random edge incident to v
- ▶ Return edge with probability $\frac{\deg(v)}{\theta}$

Edge probability: $\frac{1}{n\theta}$



Set $\theta = \sqrt{cm}$ for some constant c .

Sampling a heavy edge *approximately* uniformly

- ▶ Sample a uniform random light edge $u \rightarrow v$
- ▶ If v is light, reject
- ▶ Otherwise return a random edge incident to v

Edge probability: $\frac{\deg_\ell(v)}{\deg(v)} \frac{1}{n\theta}$

$$p \frac{1}{n\theta} \frac{1}{p} = \frac{1}{n\theta}$$

How do we get rid of ϵ ?

Problem: $\frac{\deg_\ell(v)}{\deg(v)} = p$ varies.

Idea: Use rejection sampling with probability $\frac{1}{p}$.

Fix: reject w/p $\frac{1}{2} \rightarrow \frac{1}{2p}$

Bernoulli Factories

Given: Coin with unknown probability p

Goal: Simulate a coin with probability $f(p)$

Exactly and with low expected coin flips

Easy examples

$$f(p) = \frac{1}{2}$$

Flip twice

00 \rightarrow reroll

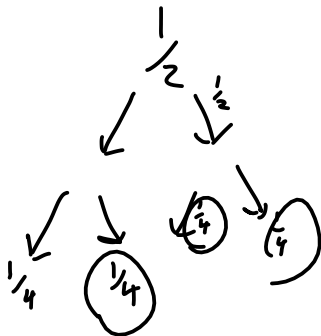
01
10 } equal prob

11 \rightarrow reroll

$$\Pr[\text{flip more than } n \text{ times}] \leq (p^2 + (1-p)^2)^{\lfloor n/2 \rfloor}$$

Easy examples

$$f(p) = \frac{3}{4}$$



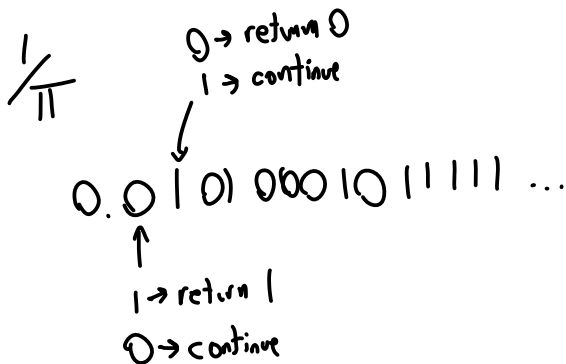
Easy examples

$$f(p) = \frac{a}{2^k}$$



Easy examples

$$f(p) = c \text{ for any } c \in [0, 1]$$



Easy examples

$$f(p) = p^2$$

Flip p twice

$$\sum_{i=0}^n \binom{n}{i} p^i (1-p)^{n-i} a[i]$$

coeff \downarrow $0 \leq a[i] \leq 1$

Our desired factory

$$f(p) = \frac{1}{2^p} \text{ (promised to have } p > \frac{1}{2} + c)$$

Existential results: This is possible and can be done efficiently

Possibility results

Keane and O'Brien (1994): The following are *necessary* and *sufficient* to simulate $f(p)$ on domain $\mathcal{P} \subseteq [0, 1]$:

1. f is continuous on \mathcal{P}
2. Either f is constant on \mathcal{P} or $\exists t \in \mathbb{N}, \forall p \in \mathcal{P}$

$$\min\{f(p), 1 - f(p)\} \geq \min\{p^t, (1 - p)^t\}.$$

Efficiency results

Nacu and Peres (2005): If f is *real analytic* on \mathcal{P} , f has an efficient simulation on \mathcal{P} .

A function f is real-analytic if f matches its Taylor series at all points

Step 1: Reduce to $2p$

2014 Hüber

$$f(p) = \frac{1}{2p} \text{ (promised to have } p > \frac{1}{2} + c \text{)}$$

$$\frac{1}{2p} = \frac{1}{\underbrace{1}_{p_1} + \underbrace{(2p-1)}_{p_2}}$$

Bernoulli Race

Flip a fair coin

Heads: output heads w/p p_1

Tails: output tails w/p $(2p-1)$

$$\Pr[\text{heads}] = \frac{1}{2} \sum_{k=0}^{\infty} \left(1 - \left(\frac{1}{2} + \frac{1}{2}(2p-1)\right)\right)^k = \frac{1}{2} \sum_{k=0}^{\infty} (1-p)^k = \frac{1}{2p}$$

$$(2p-1) \leftarrow 1 - (2p-1) = 2-2p = 2(1-p) \leftarrow 2p$$

Step 2: Factory for $2p$



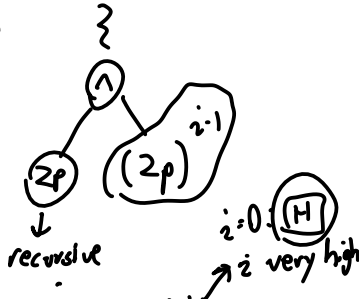
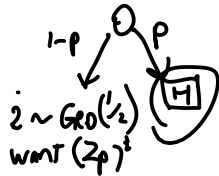
want: $\frac{p}{1-p}$ $p + (1-p)\frac{p}{1-p} = 2p$

$$\frac{p}{1-p} = p + p^2 + p^3 + \dots$$

$$= \frac{1}{2}(2p) + \frac{1}{4}(2p)^2 + \dots$$

$$= \sum_{i=1}^{\infty} \Pr[\text{Geo}(\frac{1}{2}) = i] (2p)^i$$

want $(2p)^{i+j}$



Step 2: Factory for $2p$

$$2p \leq 0.99$$
$$(2p)^{100} < \frac{1}{2}$$

$$2(2p)^{100}$$
$$= \left(2^{\frac{101}{p}} p\right)^{100}$$