

# Probabilistic Method and Random Graphs

## Lecture 6. Bins&Balls: Poisson Approximation<sup>1</sup>

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<sup>1</sup>The slides are mainly based on Chapter 5 of the textbook *Probability and Computing*.

Questions, comments, or suggestions?

# Review: bins-and-balls

General model:  $(m, n)$ -bins-and-balls

$m$  balls independently randomly placed in  $n$  bins

Distribution of the load  $X$  of a certain bin:  $\text{Bin}(m, 1/n)$

When  $m, n \gg r$ ,  $\Pr(X = r) \approx e^{-\mu} \frac{\mu^r}{r!}$  with  $\mu = \frac{m}{n}$ .

Poisson distribution

Poisson distribution:  $\Pr(X_\mu = r) = e^{-\mu} \frac{\mu^r}{r!}$ .

Law of rare events

Rooted at **Law of Small Numbers**

Law of Small Numbers can be extended to **non-independent** cases

## Low-order moments

$$\mathbb{E}[X_\mu] = \text{Var}[X_\mu] = \mu.$$

## Additive

By uniqueness of moment generation functions,  
 $X_{\mu_1} + X_{\mu_2} = X_{\mu_1 + \mu_2}$  if independent.

## Chernoff-like bounds

1. If  $x > \mu$ , then  $\Pr(X_\mu \geq x) \leq \frac{e^{-\mu}(e\mu)^x}{x^x}$ .
2. If  $x < \mu$ , then  $\Pr(X_\mu \leq x) \leq \frac{e^{-\mu}(e\mu)^x}{x^x}$ .

## Basic observation

Loads of multiple bins are not independent.

Hard to handle

## Maximum load

- $\Pr(L \geq 2) \geq 0.5$  if  $m \geq \sqrt{2n \ln 2}$ 
  - Birthday paradox
- $\Pr(L \geq 3 \frac{\ln n}{\ln \ln n}) \leq \frac{1}{n}$  if  $m = n$

Let's be ambitious

Is there a **closed form** of  $\Pr(X_1 = k_1, \dots, X_n = k_n)$ ?

Easy when  $n = 2$ .

Let's be ambitious

Is there a **closed form** of  $\Pr(X_1 = k_1, \dots, X_n = k_n)$ ?

Easy when  $n = 2$ . Hard in general?

# Joint Distribution of Bin Loads

## Theorem

$$\Pr(X_1 = k_1, \dots, X_n = k_n) = \frac{m!}{k_1!k_2!\dots k_n!n^m}$$

## Proof.

By the chain rule,

$$\begin{aligned} & \Pr(X_1 = k_1, \dots, X_n = k_n) \\ &= \prod_{i=0}^{n-1} \Pr(X_{i+1} = k_{i+1} | X_1 = k_1, \dots, X_i = k_i) \end{aligned}$$

Note that  $X_{i+1} | (X_1 = k_1, \dots, X_i = k_i)$  is a binomial r.v. of  $m - (k_1 + \dots + k_i)$  trials with success probability  $\frac{1}{n-i}$ .



## Remark

- You can also prove by counting
- Multinomial coefficient  $\frac{m!}{k_1!k_2!\dots k_n!}$ : the number of ways to allocate  $m$  distinct balls into groups of sizes  $k_1, \dots, k_n$



# Silver bullet for Bins&Balls problems?

## In principle

Yes, since it can be computed

## In practice

Usually No, since it is too hard to compute.

Example: what's the probability of having empty bins?

## In need

Approximation for computing or **insights for analysis**

# Poisson Approximation

## At the first glance

The (marginal) load  $X_i \sim \text{Bin}(m, \frac{1}{n})$  for each bin  $i$ .

$\{X_1, \dots, X_n\}$  are not independent.

But seemingly the only dependence is that their sum is  $m$ . So,

## A plausible conjecture

The joint distribution  $(X_1, \dots, X_n) \sim (Y_1, \dots, Y_n | \sum Y_i = m)$ ,  
where  $Y_i \sim \text{Bin}(m, \frac{1}{n})$  are mutually independent

If this is true, a good simplification is obtained.

## However

It is NOT the case!

$(Y_1, \dots, Y_n | \sum Y_i = m)$  doesn't have marginal distr. as  $Y_i$ 's.

# General Fact

$Y_i$ : mutual independent,  $1 \leq i \leq n$ .

$(Y_1, \dots, Y_n | g(\vec{Y}))$  doesn't have marginal distr. as  $Y_i$ 's.

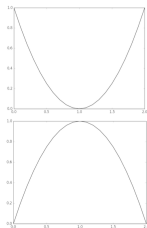


Figure: Distributions of  $X, Y$

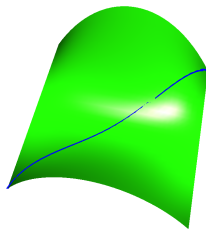


Figure: Joint distribution  $(X, Y) | X + Y = 1$  (blue curve)

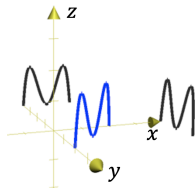


Figure: The marginals

Recall the **false** conjecture

The joint distribution  $(X_1, \dots, X_n) \sim (Y_1, \dots, Y_n | \sum Y_i = m)$ ,  
where  $Y_i \sim \text{Bin}(m, \frac{1}{n})$  are mutually independent

Is the conjecture true for any distribution other than binomial?

Yes!

Poisson distribution again. (Better than the conjecture)

# Poisson Approximation Theorem

## Notation

$X_i^{(m)}$ : the load of bin  $i$  in  $(m, n)$ -model,  $1 \leq i \leq n$ .

$Y_i^{(\mu)}$ : independent Poisson r.v.s with expectation  $\mu$ ,  $1 \leq i \leq n$ .

## Theorem

$$\left( X_1^{(m)}, X_2^{(m)}, \dots, X_n^{(m)} \right) \sim \left( Y_1^{(\mu)}, Y_2^{(\mu)}, \dots, Y_n^{(\mu)} \mid \sum Y_i^{(\mu)} = m \right).$$

## Remarks

- The equation is independent of  $\mu$ : For any  $m$ , the same Poisson distribution works.
- Since  $\Pr \left( X_1^{(m)}, X_2^{(m)}, \dots, X_n^{(m)} \right) \propto \Pr \left( Y_1^{(\mu)}, Y_2^{(\mu)}, \dots, Y_n^{(\mu)} \right)$ , the  $X_i$ 's are **decoupled**.
- The two distributions are exactly equal, not approximate.

## Proof

By straightforward calculation.

## Coupon Collector Problem

$X$ : the number of purchases until  $n$  types are collected.

For any constant  $c$ ,  $\lim_{n \rightarrow \infty} \Pr(X > n \ln n + cn) = 1 - e^{-e^{-c}}$ .

Remark:  $\Pr(n \ln n - 4n \leq X \leq n \ln n + 4n) \geq 0.98$

## Basic idea of the proof

Use bins-and-balls model and the Poisson approximation.

It holds under the Poisson approximation.

The approximation is actually accurate.

## Modeling

$X > n \ln n + cn$  is equivalent to event  $\bar{\mathcal{E}}$ , where  $\mathcal{E}$  means that there is no empty bin in the  $(n \ln n + cn, n)$ -Bins&Balls model.

## It holds under the Poisson approximation

**Approximation experiment:**  $n$  bins, each having an independent Poisson number  $Y_i$  of balls with the expectation  $\ln n + c$ .

Event  $\mathcal{E}'$ : No bin is empty.

$$\Pr(\mathcal{E}') = (1 - e^{-(\ln n + c)})^n = \left(1 - \frac{e^{-c}}{n}\right)^n \rightarrow e^{-e^{-c}}.$$

## The approximation is accurate

Obj.: **Asymptotically,  $\Pr(\mathcal{E}) = \Pr(\mathcal{E}')$ .**

By Poisson Approximation,  $\Pr(\mathcal{E}) = \Pr(\mathcal{E}' | Y = n \ln n + cn)$ , so we prove  $\Pr(\mathcal{E}') = \Pr(\mathcal{E}' | Y = n \ln n + cn)$  where  $Y = \sum_{i=1}^n Y_i$ .

### Further reduction

Since  $\Pr(\mathcal{E}') = \Pr(\mathcal{E}'|Y \in \mathbb{Z})$ , there should be  $\mathcal{N} \subset \mathbb{Z}$  s.t  $n \ln n + cn \in \mathcal{N}$  and  $\Pr(\mathcal{E}') \approx \Pr(\mathcal{E}'|Y \in \mathcal{N})$ .

If  $\mathcal{N}$  is not too small or too big, i.e.

- $\Pr(Y \in \mathcal{N}) \approx 1$ ;
- $\Pr(\mathcal{E}'|Y \in \mathcal{N}) \approx \Pr(\mathcal{E}'|Y = n \ln n + cn)$ .

We finish the proof by total probability formula.

### Does such $\mathcal{N}$ exist?

Yes! Try the  $\sqrt{2m \ln m}$ -neighborhood of  $m = n \ln n + cn$ .



Proof:  $\Pr(|Y - m| \leq \sqrt{2m \ln m}) \rightarrow 1$

$Y \sim \text{Poi}(m)$ .

By Chernoff bound  $\Pr(Y \geq y) \leq \frac{e^{-m}(em)^y}{y^y} = e^{y-m-y \ln \frac{y}{m}}$ ,

$$\begin{aligned} \Pr\left(Y > m + \sqrt{2m \ln m}\right) &\leq e^{\sqrt{2m \ln m} - (m + \sqrt{2m \ln m}) \ln\left(1 + \sqrt{\frac{2 \ln m}{m}}\right)} \\ &\quad \text{by } \ln(1+z) \geq z - z^2/2 \text{ for } z \geq 0 \\ &\leq e^{-\ln m + \frac{\ln^{3/2} m}{\sqrt{m}}} \rightarrow 0. \end{aligned}$$

Likewise,  $\Pr(Y < m - \sqrt{2m \ln m}) \rightarrow 0$ .

Altogether

$\Pr(Y \in \mathcal{N}) \approx 1$ . So,  $\mathcal{N}$  is not too small.

Proof:  $\Pr(\mathcal{E}' \mid |Y - m| \leq \sqrt{2m \ln m}) \approx \Pr(\mathcal{E}' \mid Y = m)$

$\Pr(\mathcal{E}' \mid Y = k)$  increases with  $k$ , so

$$\begin{aligned} & \Pr(\mathcal{E}' \mid Y = m - \sqrt{2m \ln m}) \\ & \leq \Pr(\mathcal{E}' \mid |Y - m| \leq \sqrt{2m \ln m}) \\ & \leq \Pr(\mathcal{E}' \mid Y = m + \sqrt{2m \ln m}). \end{aligned}$$

$$\begin{aligned} & |\Pr(\mathcal{E}' \mid |Y - m| \leq \sqrt{2m \ln m}) - \Pr(\mathcal{E}' \mid Y = m)| \\ & \leq \Pr(\mathcal{E}' \mid Y = m + \sqrt{2m \ln m}) - \Pr(\mathcal{E}' \mid Y = m - \sqrt{2m \ln m}) \\ & = \Pr(A) \text{ (By Poisson approximation)}. \end{aligned}$$

Event  $A$ : In the  $(m + \sqrt{2m \ln m}, n)$ -Bins&Balls model, the first  $m - \sqrt{2m \ln m}$  balls leave at least one bin empty, but at least one among the next  $2\sqrt{2m \ln m}$  balls goes into each of the empty bins.

By Union Bound,  $\Pr(A) \rightarrow 0$