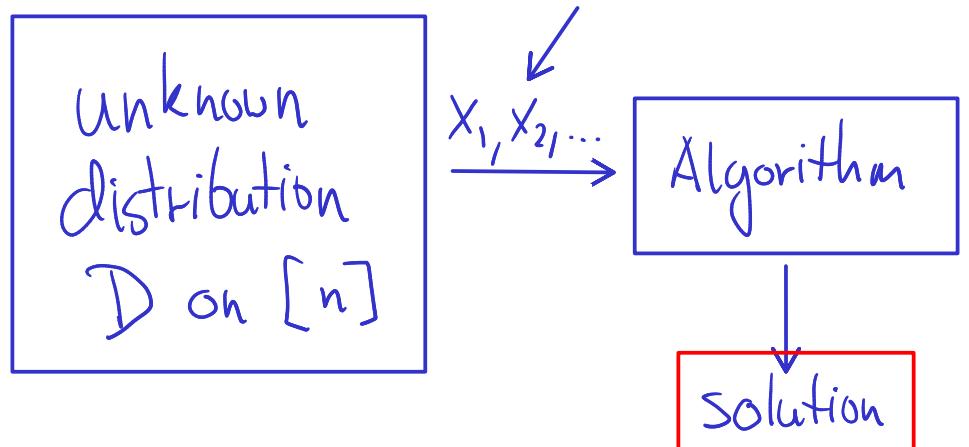


Today: Continue uniformity testing

## Review

Model:



Task: Uniformity testing

① If  $D = U_{[n]}$ , output YES w.p. 99/100

$\uparrow$   
uniform distribution on  $[n]$

② If  $d_{TV}(D, U_{[n]}) \geq \varepsilon$ , output NO w.p. 99/100

Want to use as few samples as possible

## Algorithm proposed last time

sufficiently large constant

- collect  $s = C \cdot \sqrt{n}/\varepsilon^4$  independent

samples  $X_1, X_2, \dots, X_s$  from  $D$

- count collisions:

$$Y = \sum_{i < j} Y_{ij}$$

$= \begin{cases} 1 & \text{if } X_i = X_j \\ 0 & \text{otherwise} \end{cases}$

- if  $\frac{Y}{\binom{s}{2}} \geq \frac{1}{n} + \frac{2\varepsilon^2}{n}$

output NO

else output YES

## Analysis last time:

$$\mathbb{E}\left[\frac{Y}{\binom{s}{2}}\right] = \frac{1}{\binom{s}{2}}$$

$$\textcircled{1}: D = U_{[n]} \Rightarrow \frac{1}{\binom{s}{2}} = \frac{1}{n}$$

distribution  
treated as  
n-dimensional  
vector of probabilities

$$\textcircled{2}: d_{TV}(D, U_{[n]}) \geq \varepsilon \Rightarrow \frac{1}{\binom{s}{2}} \geq \frac{1+2\varepsilon^2}{n}$$

Our approach: show that  $\frac{Y}{\binom{s}{2}}$  is a good estimator for  $\|D\|_2^2$

$$\text{Lemma: } \text{Var}[Y] \leq 7 \left( \binom{s}{2} \|D\|_2^2 \right)^{3/2}$$

Definition:  $\bar{Y}_{ij} = Y_{ij} - \mathbb{E}[Y_{ij}]$   
 (of course,  $\mathbb{E}[\bar{Y}_{ij}] = 0$ )

Useful facts:

$$\boxed{1} \quad \mathbb{E}[\bar{Y}_{ij} \bar{Y}_{kl}] \leq \mathbb{E}[Y_{ij} Y_{kl}]$$

$\|D\|_2^2$

Why?  $\mathbb{E}[\bar{Y}_{ij} \bar{Y}_{kl}] = \mathbb{E}[(Y_{ij} - \mathbb{E}[Y_{ij}]) (Y_{kl} - \mathbb{E}[Y_{kl}])]$

$$= \mathbb{E}[Y_{ij} Y_{kl} - Y_{ij} \mathbb{E}[Y_{kl}] - Y_{kl} \mathbb{E}[Y_{ij}] + \mathbb{E}[Y_{ij}] \mathbb{E}[Y_{kl}]]$$

$$= \mathbb{E}[Y_{ij} Y_{kl}] - 2 \left( \mathbb{E}[Y_{ij}] \mathbb{E}[Y_{kl}] \right) + \left( \mathbb{E}[Y_{ij}] \mathbb{E}[Y_{kl}] \right)$$

$$= \mathbb{E}[Y_{ij} Y_{kl}] - 2 \left( \mathbb{E}[Y_{ij}] \mathbb{E}[Y_{kl}] \right) \leq \mathbb{E}[Y_{ij} Y_{kl}]$$

$$\boxed{2} \quad \underbrace{\|D\|_3 \leq \|D\|_2}_{\text{This is an inequality on norms that follows from Hölder's inequality}}$$

This is an inequality on norms that follows from Hölder's inequality

$$\boxed{3} \quad \frac{s^2 \leq 3 \binom{s}{2}}{}$$

$$s^2 \stackrel{\Leftrightarrow}{\leq} \frac{3}{2}s(s-1)$$

$$s \stackrel{\Leftrightarrow}{\leq} \frac{3}{2}s - \frac{3}{2}$$

$$3 \stackrel{\Leftrightarrow}{\leq} s \leftarrow$$

$$\binom{s}{3} \leq \frac{s^3}{6}$$

true if the constant C defining  $s$  is large enough

$\boxed{4}$

$$\frac{\binom{s}{3} \leq \frac{s^3}{6}}{\binom{s}{3} = \frac{s(s-1)(s-2)}{6} \leq \frac{s^3}{6}}$$

Proof of the variance lemma:

$$\begin{aligned} \text{Var}[Y] &= \text{Var}\left[\sum_{i < j} Y_{ij}\right] = \text{Var}\left[\sum_{i < j} \bar{Y}_{ij}\right] \\ &= \mathbb{E}\left[\left(\sum_{i < j} \bar{Y}_{ij}\right)^2\right] - \underbrace{\left(\mathbb{E}\left[\sum_{i < j} \bar{Y}_{ij}\right]\right)^2}_{=0} \end{aligned}$$

$$= \mathbb{E} \left[ \sum_{i < j} \bar{Y}_{ij} + \sum_{\substack{i < j \\ k < l}} \bar{Y}_{ij} \bar{Y}_{kl} \right]$$

1

$i, j, k, l$  distinct

$$+ \sum_{\substack{i < j \\ i < l \\ i, j, l \text{ distinct}}} \bar{Y}_{ij} \bar{Y}_{il}$$

2

$$+ \sum_{\substack{i < j \\ k < j \\ i, j, k \text{ distinct}}} \bar{Y}_{ij} \bar{Y}_{kj}$$

3

$$+ \sum_{\substack{i < j \\ j < l}} \bar{Y}_{ij} \bar{Y}_{jl}$$

4

$$+ \sum_{\substack{i < j \\ k < i}} \bar{Y}_{ij} \bar{Y}_{ki}$$

5

We analyze the expectation term by term:

$$\text{① } \Delta : E\left[\sum_{i < j} \bar{Y}_{ij}^2\right] \leq E\left[\sum_{i < j} Y_{ij}^2\right]$$

$$= E\left[\sum_{i < j} Y_{ij}\right]^2 = \binom{s}{2} \|D\|_2^2$$

$$\text{② } \Delta : E\left[\sum_{\substack{i < j \\ k < l}} \bar{Y}_{ij} \bar{Y}_{kl}\right]$$

$i, j, k, l$  independent

$$\text{independence} = \sum \underbrace{E[\bar{Y}_{ij}]}_{=0} \underbrace{E[\bar{Y}_{kl}]}_{=0} = 0$$

$$\text{③ } \Delta : E\left[\sum_{\substack{i < j \\ i < l}} \bar{Y}_{ij} \bar{Y}_{il}\right]$$

$i, j, l$  distinct

$$\leq E\left[\sum Y_{ij} Y_{il}\right]$$

$$= \sum_{\substack{i < j \\ i < l \\ i, j, l \text{ distinct}}} \Pr[X_i = X_j = X_l]$$

$i, j, l$  distinct

$$\leq 2 \left( \frac{S}{3} \right) \sum_{i=1}^n p_i^3$$

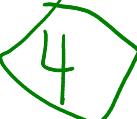
$p_i$  = probability of  
drawing  $i$  from  $D$

Because for  
any selection  
of three distinct  
elements we  
have either

$$i < j < l$$

$$\text{or } i < l < j$$

$$\leq 2 \cdot \frac{S^3}{6}$$

due to 

$$\cdot \binom{\|D\|_2}{3}$$

due to 

$$\leq \sqrt{3} \left( \binom{s}{2} \|D\|_2^2 \right)^{3/2}$$



due to  $\boxed{3}$ ,  $s^3 = (s^2)^{3/2} \leq \left(3 \binom{s}{2}\right)^{3/2}$



: same bound as  $\triangle 3$

$\triangle 5$  &  $\triangle 6$  :  $\leq \frac{\sqrt{3}}{2} \left( \binom{s}{2} \|D\|_2^2 \right)^{3/2}$  each

Almost the same as  $\triangle 3$  &  $\triangle 4$ ,  
except for each triple selected from  
s options, there is a unique assignment  
to indices in the summation. So no  
factor of 2 needed.

Overall:

$$\text{Var}[Y] \leq \binom{s}{2} \|D\|_2^2 + 3 \sqrt{3} \left( \binom{s}{2} \|D\|_2^2 \right)^{3/2}$$

If  $C$  in the definition of  $s$  is large  
enough,  $\binom{s}{2} \geq n$ . Then  $\binom{s}{2} \|D\|_2^2 \geq n \cdot \frac{1}{n} \geq 1$ ,  
and therefore,  $\binom{s}{2} \|D\|_2^2 \leq \left( \binom{s}{2} \|D\|_2^2 \right)^{3/2}$ .

Hence

$$\text{Var}[Y] \leq \underbrace{\left(1 + 3\sqrt{3}\right)}_{\leq 7} \left(\binom{s}{2} \frac{\|D\|^2}{2}\right)^{3/2}$$



(to be continued in the next lecture)