

Lecture 12

(1)

Review

For testing if X_1, \dots, X_n are generated by

$P = \text{unif}[k]$ or some $Q \in \mathcal{P}_k$ s.t. $d(P, Q) \geq \varepsilon$,

the least no. of samples n^* ($= n_{Y_3 - Y_3}$)
needed satisfy

$$n^* = O\left(\frac{k}{\varepsilon^2}\right).$$

Today: - Complete the proof of upper
bound

- Lower bound

[A] Analysis of Collision-based tester
(contd.)

Recall $S = \sum_{i < j} \mathbb{1}_{\{x_i = x_j\}}$

$$\mathbb{E}_P[S] = \binom{n}{2} \|P\|_2^2$$

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$$\mathbb{E}_P[S^2] = \binom{n}{2} \|P\|_2^2 + n(n-1)(n-2) \|P\|_3^3$$

$$\Rightarrow \text{Var}(S) \leq n^2 \|P\|_2^2 + n^3 (\|P\|_3^3 - \|P\|_2^4)$$

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With large prob.,

$$S \leq \binom{n}{2} \frac{1}{k} + \sqrt{3 \frac{n^2}{k}}$$

For a P s.t. $d(P, \text{unif}) \geq \varepsilon$

With large prob.,

$$S \geq \binom{n}{2} \|P\|_2^2 - \sqrt{3(n^2 \|P\|_2^2 + n^3 (\|P\|_3^3 - \|P\|_2^4))}$$

$$\text{Observations: } (i) \|P\|_2^2 = \|P - \text{unif}\|_2^2 + \frac{1}{k}$$

$$(ii) \|P - \text{unif}\|_2^2 \geq \frac{\|P - \text{unif}\|_2^2}{k}$$

$$= \frac{4d^2(P, \text{unif})}{k}$$

Thus, with large prob.,

$$S \geq \binom{n}{2} \frac{1}{k} + \binom{n}{2} \|P - \text{unif}\|_2^2 - \sqrt{3(n^2 \|P\|_2^2 + n^3 (\|P\|_3^3 - \|P\|_2^4))}$$

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Therefore, it suffices to have

$$(a) \frac{n^2}{k} d^2(P, \text{unif}) \geq \sqrt{\frac{n^2}{k}}$$

$$\Leftrightarrow n \geq \frac{\sqrt{k}}{d^2(P, \text{unif})} \Leftarrow n \geq \frac{\sqrt{k}}{\varepsilon^2},$$

$$\text{and (b)} n^2 \|P - \text{unif}\|_2^2 \geq n \sqrt{\|P\|_2^2 + n(\|P\|_3^3 - \|P\|_2^4)}.$$

Note that

$$\begin{aligned} & \|P\|_2^2 + n(\|P\|_3^3 - \|P\|_2^4) \\ & \leq \|P - \text{unif}\|_2^2 + \frac{1}{k} + n \|P - P_{\text{unif}}\|_2^3 \\ & \quad + \frac{3n}{k} \|P - P_{\text{unif}}\|_2^2 \end{aligned}$$

Thus, it suffices to have

$$n \geq \max \left\{ \frac{1}{\|P - \text{unif}\|_2}, \frac{1}{\sqrt{k} \|P - \text{unif}\|_2^2}, \sqrt{\frac{n}{\|P - \text{unif}\|_2}} \right\}$$

$$\sqrt{\frac{n}{k} \frac{1}{\|P - P_{\text{unif}}\|_2^2}}$$

$$\Leftrightarrow n \geq \max \left\{ \frac{\sqrt{k}}{\varepsilon}, \frac{\sqrt{k}}{\varepsilon^2}, \sqrt{\frac{nk}{\varepsilon}}, \sqrt{\frac{nk}{\varepsilon^2}} \right\}.$$

$$\Leftrightarrow n \geq \frac{\sqrt{k}}{\varepsilon^2}.$$

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B Lower Bound

We shall prove $n^* = \Omega\left(\frac{\sqrt{k}}{\varepsilon^2}\right)$

(1) Let's show a simpler result first,
namely, $n^* = \Omega(\sqrt{k})$

Lemma (Birthday Paradox)

For a uniform distribution on $[k]$,

for $n \ll \sqrt{k}$, with large prob., no. repetitions occur.

Proof Sketch. Poisson approximation

Let $N \sim \text{Poi}(n)$. Consider N iid samples X_1, \dots, X_N

from $P \in P_k$. Then,

$$(a) N_x := \sum_{i=1}^N \mathbb{1}_{\{X_i=x\}} \equiv \# \text{ of times } x \text{ occurs}$$

in X^N , $x \in \mathcal{X}$.

$\{N_x\}$ are independent for different x ,

with $N_x \sim \text{Poi}(n p_x)$.

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Let $M_2 = \sum_x \mathbb{1}_{\{N_x \geq 2\}}$, i.e., no. of symbols appearing 2 or more times.

Then,

$$\begin{aligned} P(M_2 > 0) &= P(M_2 \geq 1) \\ &\leq E[M_2] = \sum_x P(N_x \geq 2) \end{aligned}$$

$$= \sum_x \frac{(n p_n)^2}{2} e^{-n p_n}$$

For $P \equiv \text{unif}[k]$,

$$P(M_2 > 0) \leq k \cdot e^{-n/k} \frac{n^2}{k^2} \leq \frac{n^2}{k}$$

(b) Removing Poisson approx.

$$\begin{aligned} P(M_2 > 0) &\geq P\left(N \in \left[\frac{n}{2}, 2n\right]\right) P\left(M_2 > 0 \mid N = \frac{n}{2}\right) \\ &\geq (1 - e^{-cn}) \cdot P\left(M_2 > 0 \mid N = \frac{n}{2}\right) \end{aligned}$$

\Rightarrow For $\frac{n}{2}$ samples, no repetitions occur

$$\text{with prob. } \leq \frac{n^2}{k} (1 - e^{-cn}).$$

Therefore, no repetitions occur w.p. ≈ 1 if $n \ll \sqrt{k}$

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Now the proof of lower bound proceeds as follows:

Consider the distributions $\mathcal{Q}_A = \text{unif}\{A\}$ for subsets A with $|A| = k_2$. Each \mathcal{Q}_A is at a distance $d(\text{unif}[k], \mathcal{Q}_A) = \frac{1}{2}$.

$$\text{Let } \bar{\mathcal{Q}}^{(n)} = \frac{1}{\binom{k}{k_2}} \sum_{A: |A|=k_2} \mathcal{Q}_A^{\otimes n}.$$

By symmetry, a sufficient statistic to distinguish unif from $\bar{\mathcal{Q}}^{(n)}$ is the so called profile, i.e., the vector $\Phi = (\Phi_1, \dots, \Phi_n)$

where $\Phi_i = \# \text{ of symbols appearing } i \text{ times}$.

But for $n < \sqrt{k}$, by Birthday Paradox, the profiles under $\text{unif}[k]$ and $\bar{\mathcal{Q}}^{(n)}$ are exactly $\mathbb{I} = (n, 0, \dots, 0)$ with large prob.

The proof is completed by noting that any test for $\epsilon \leq \frac{1}{2}$ can distinguish $\bar{\mathcal{Q}}^{(n)}$ from $\text{unif}[k]$.

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(2) Paninski's lower bound

Consider now the family $\{Q_z\}_{z \in \{-1, 1\}^{k/2}}$ defined as follows: under Q_z , the elements

$2i$ and $2i+1$, $0 \leq i \leq \frac{k-1}{2}$, have masses

$\frac{1+2\varepsilon z_i}{k}$ and $\frac{1-2\varepsilon z_i}{k}$, resp.

Thus, $d(Q_z, \text{unif}[k]) = \varepsilon$ for every $z \in \{-1, 1\}^{k/2}$.

Let $\bar{Q}^{(n)} = \frac{1}{2^{k/2}} \sum_z Q_z^{\otimes n}$.

We want to bound $d(\text{unif}[k]^{\otimes n}, \bar{Q}^{(n)})$.

Aside: $d(P, Q) \leq D(P \| Q)$ ($\text{if } P \ll Q, Q \ll P$)

$$= \mathbb{E}_P \left[\log \frac{P(X)}{Q(X)} \right]$$

$$\leq \mathbb{E}_P \left[\left| \frac{P(X)}{Q(X)} - 1 \right| \right]$$

Alternatively,

$$4d^2(P, Q) = \mathbb{E}_Q \left[\left| \frac{P(X)}{Q(X)} - 1 \right| \right]$$

$$\leq \sqrt{\mathbb{E}_Q \left[\left(\frac{P(X)}{Q(X)} - 1 \right)^2 \right]}.$$

Q. Which of these bounds do you prefer?

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For our construction,

$$\frac{Q_z^{\otimes n}(\underline{x})}{P(\underline{x})} = \prod_{i=1}^n [1 + g(x_i, z)]$$

where $g(x_i, z) = \begin{cases} 2\varepsilon z_j & \text{if } x_i = 2j \\ -2\varepsilon z_j & \text{if } x_i = 2j+1. \end{cases}$

Thus,

$$\begin{aligned} & \left(\sum_z 2^{-k/2} \frac{Q_z^{\otimes n}(\underline{x})}{P(\underline{x})} - 1 \right)^2 \\ &= \left(2^{-k/2} \sum_z \left[1 + \sum_i g(x_i, z) + \sum_{i_1 < i_2} g(x_{i_1}, z) g(x_{i_2}, z) + \dots \right] - 1 \right)^2 \\ &= 2^{-k} \sum_{z, z'} \left(\sum_{i_1, i_2} g(x_{i_1}, z) g(x_{i_2}, z') + \sum_{\substack{i_1, i_2, i_3 \\ i_1 < i_2 < i_3}} g(x_{i_1}, z) g(x_{i_2}, z) g(x_{i_3}, z') + \dots \right) \end{aligned}$$

Under P , $\mathbb{E}[g(x_i, z)] = 0$ and x_1, \dots, x_n are iid. Thus, only the terms involving pairs $g(x_i, z) g(x_i, z')$ remain on taking \mathbb{E}_P .

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Therefore,

$$\mathbb{E}_{P^{\otimes n}} \left[\left(\frac{\bar{Q}^{\otimes n}}{P^{\otimes n}} (X^n) - 1 \right)^2 \right]$$

$$= 2^{-k} \sum_{z, z'} \left[\sum_j H_j(z, z') + \sum_{j > j'} H_j(z, z') H_{j'}(z, z') + \dots \right],$$

where

$$H_j(z, z') = \mathbb{E}_P [g(X_j, z) g(X_j, z')]$$

$$\begin{aligned} &= \frac{1}{k} \sum_{i=1}^{k/2} g(2i, z) g(2i, z') \\ &\quad + g(2i+1, z) g(2i+1, z') \\ &= \frac{8\varepsilon^2}{k} \sum_{i=1}^{k/2} z_i \cdot z'_i. \end{aligned}$$

Thus, the left-side above is bounded by

$$2^{-k} \sum_{z, z'} \prod_{j=1}^n (1 + H_j(z, z')) - 1$$

$$\leq \mathbb{E}_{z, z'} \left[e^{\sum_{j=1}^n H_j(z, z')} \right] - 1$$

$$\leq \mathbb{E}_{z, z'} \left[e^{\frac{8n\varepsilon^2}{k} \sum_{i=1}^{k/2} z_i \cdot z'_i} \right] - 1$$

$$\leq e^{C \cdot \frac{n^2 \varepsilon^4}{k}} - 1 \Rightarrow \boxed{\frac{n^2 \varepsilon^4}{k} \geq \text{constant}}$$