

Maximum Likelihood Estimator :-

let X_1, X_2, \dots, X_n be i.i.d X $\begin{cases} \text{p.m.f.} \\ \text{p.d.f.} \end{cases} f(x|p)$
 $p \in \mathcal{P} \subseteq \mathbb{R}^d.$

Likelihood function

$$L(p; X_1, \dots, X_n) = \prod_{i=1}^n f(x_i | p)$$

Suppose $\hat{p} \equiv \hat{p}(X_1, \dots, X_n)$ is a point at which $L(p; X_1, \dots, X_n)$ attains the maximum as a function of p

$\Rightarrow \hat{p} \equiv$ maximum likelihood estimator of p
(M.L.E.)

Applications to Edge Probability of Erdős-Rényi

Recall (Friday) Worksheet

(i) choose $x \in \{1, 2, 3, 4, 5\}$. \equiv uniformly.

(ii) $A \equiv$ designate an event in the

Experiment of rolling a die, whose
probability $p = \frac{x}{6}$

[$X \equiv$ outcome of roll of a die

$$P(X=1) = 1/6$$

$$P(X \leq 2) = 2/6 = x$$

$$P(X \leq 3) = 3/6$$

$$P(X \leq 4) = 4/6$$

$$P(X \leq 5) = 5/6 \quad (\text{E.g.}]$$

Review - Erdős - Renyi Graph $G(n, p)$

- late 40's - applications in networks
- well studied - simple to understand.

• $V = \{1, \dots, n\}$

$$E \subseteq V \times V$$

$$= \{i, j\} \mid i, j \in V\}$$

• $i \sim j$ w.p. p

(place an edge with probability p)

- each vertex $i \in V \equiv$ ^{can be} connected to $n-1$ vertices

So there are $\frac{n(n-1)}{2}$ possible edges

Expected number of edges :- $\frac{n(n-1)}{2} p$
in Erdős Rens

(iii) Went to each edge (i, j)

- Rolled the die

- if event A occurred then placed an edge; otherwise not.

[Construction]

↳ graph and Adjacency matrix

Question:- From the graph $G(n, \frac{x}{6})$ Can we estimate x ? [n=10]

Answer:- let E be number of edges in $G(n, \frac{x}{6})$

 $n-1$ edges w.p. p

- performing $\frac{n(n-1)}{2}$ trials with success probability $p \equiv \frac{x}{6}$

- $E \equiv \#$ of successes in $\frac{n(n-1)}{2}$ trials

$$E \sim \text{Binomial} \left(\frac{n(n-1)}{2}, \frac{\alpha}{6} \right)$$

Likelihood for E

$$L(\alpha; E) = \binom{\frac{n(n-1)}{2}}{E} \left(\frac{\alpha}{6} \right)^E \left(1 - \frac{\alpha}{6} \right)^{\frac{n(n-1)}{2} - E}$$

\therefore Given E :- Find α^* such that

$$L(\alpha^*, E) = \max_{0 \leq \alpha \leq 6} L(\alpha; E)$$

Take logarithm

$$\begin{aligned} T(\alpha) \equiv \log(L(\alpha; E)) &= \log \left(\binom{\frac{n(n-1)}{2}}{E} \right) + E \log \left(\frac{\alpha}{6} \right) \\ &\quad + \left(\frac{n(n-1)}{2} - E \right) \log \left(1 - \frac{\alpha}{6} \right) \end{aligned}$$

$$T'(\alpha) = E \cdot \frac{6}{\alpha} \cdot \frac{1}{6} + \left(\frac{n(n-1)}{2} - E \right) \frac{1}{1 - \frac{\alpha}{6}} \left(-\frac{1}{6} \right)$$

Set $T'(x) = 0 \Rightarrow$

$$\frac{E}{x} = \left(\frac{n(n-1)}{2} - E \right) \frac{1}{6} \cdot \frac{1}{1 - \frac{x}{6}}$$

$$\Rightarrow \frac{1 - \frac{x}{6}}{x/6} = \frac{n(n-1)}{2E} - 1$$

$$\Rightarrow \frac{1}{x/6} = \frac{n(n-1)}{2E}$$

$$\Rightarrow x = \frac{6 \times 2 \times E}{n(n-1)}$$

$\therefore x^* = \frac{12E}{n(n-1)}$ is MLE for x in $G(n, \frac{x}{6})$

if there are E - edges.

E.g.:- Suppose $E = 7$ $\Rightarrow x^* = \frac{84}{90} = \frac{14}{15} \approx 1$
 $n = 10$

Estimate for $x \approx 1$.

Question 2 : Consider $G(n, \frac{\lambda}{n})$ for $\lambda > 0$.

and n - large.

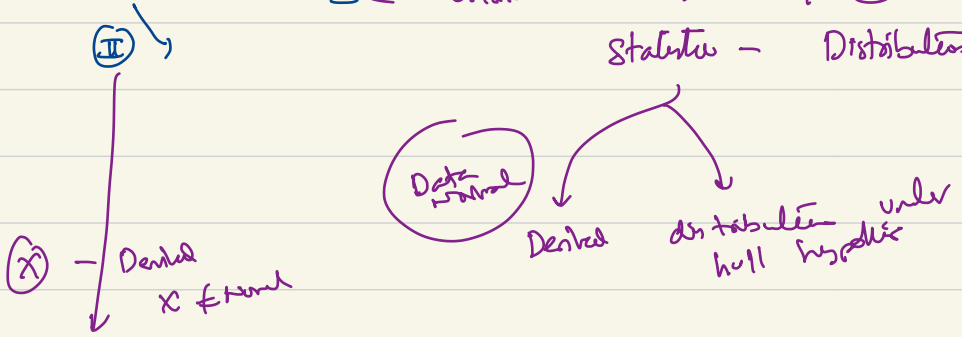
• let $d_n \equiv$ degree of a uniformly chosen vertex.

• What is $\lim_{n \rightarrow \infty} P(d_n = k) \equiv ?$

Veritasium

• σ - known $\bar{X} - \bar{Y}$ - understood

• σ - unknown $\frac{\bar{X} - \mu}{S} \sim t$ | $\frac{\bar{X} - \bar{Y}}{S} \rightarrow$
 $S \leftarrow$ unknown \hookrightarrow estimate
 Statistik - Distributionen



$\frac{\bar{X} - \bar{Y}}{S} \stackrel{!}{=} \text{das}$ mit unknown σ
 depend μ

$Z_1 \dots Z_n \dots \frac{\bar{Z} - \bar{W}}{S} \leftarrow \text{da}$

$\sigma = .1$, Für any Simulate

E.g. $U(-1,1) \times S \leftarrow \mu_1 \neq \mu_2$
 $U(-1,1) \tilde{S}$

E.C.D.F $\leftarrow \dots$ C.D.F.