

Note 2

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Possion

Continue from last class:

$$\lambda = 2$$

the expected number of calls is 2 per minute, and the # of calls in X minutes with

$$N_x \sim Possion(\lambda x)$$

P(wait at least X minutes for the first call)

$$P(X > x) = \int_x^\infty f_{exp}(t) dt$$

$$P(X > x) \rightarrow P(N_x = 0) = \frac{(\lambda x)^0 e^{-\lambda x}}{0!} = e^{-\lambda x}$$

Question:

Compute that an event does occur during x wait of time

$$P(X > x) \rightarrow P(N_x = 1) = \frac{(\lambda x)^1 e^{-\lambda x}}{1!}$$

Beta

Bayesian statistics

$$f(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$
$$0 \leq x \leq 1$$

$$\Gamma(r) = (r-1)!$$

$$\Gamma(1) = 1$$

$$\Gamma\left(\frac{1}{2}\right) = \pi^{\frac{1}{2}}$$

$$\alpha = \beta = 1$$

$$Beta(1, 1) = 1 \sim Uniform(0, 1)$$

Law of Large Number

Weak Law of Large Number

$$\lim_{n \rightarrow \infty} P(|\bar{x} - \mu| > \epsilon) = 0$$

$$\bar{X}_n = \frac{\sum_{i=1}^n X_i}{n}$$

\bar{X} converges in probability to μ

Strong Law of Large Number

$$P(\lim_{n \rightarrow \infty} |\bar{x} - \mu| > \epsilon) = 0$$

$$\hat{\theta} : estimator \rightarrow \theta$$

$$\bar{x} : estimator \rightarrow \mu$$

$$s^2 : estimator \rightarrow \sigma^2$$

$$\hat{\sigma}^2 = \frac{\sum (x_i - \bar{x})^2}{n}$$

$$s^2 = \frac{\sum (x_i - \bar{x})^2}{n - 1}$$

$$MSE(\hat{\theta}) = E[(\hat{\theta} - \theta)^2] = Var(\hat{\theta}) + [E(\hat{\theta}) - \theta]^2$$

$$bias : [E(\hat{\theta}) - \theta]^2$$

This is the bias-Variance trade off

$$\hat{p} = \frac{x}{n}$$

$$E(\hat{p}) = p$$

$$Var(\hat{p}) = \frac{p(1-p)}{n}$$

$$\lim_{n \rightarrow \infty} Var(\hat{p}) = 0$$

\hat{p} is a consistent estimator of p when $\lim_{n \rightarrow \infty} MSE(\hat{\theta}) = 0$

$$x_i \sim F_x(x|\theta) \rightarrow (x_1, \dots, x_n) \rightarrow \hat{\theta} = S(x_1, x_n) \rightarrow \theta$$

$$\hat{F}_x \rightarrow F_x(x)$$

Check with the MSE to see if this is a consistent estimator.

$$F_x(x_0) = P(X \leq x_0)$$

EDF:

$$\hat{F}_x(x_0) = \frac{\text{count}(x_1 \leq x_0)}{n} = \frac{\sum_{i=1}^n I(x_i \leq x_0)}{n}$$

Ex: Given 1, 2, 2, 3, 5; $\hat{F}(1) = \frac{1}{5}$; $\hat{F}(2) = \frac{3}{5}$; $\hat{F}(3) = \frac{4}{5}$; $\hat{F}(5) = 1$

0 if $x < 1$ $x < x_{(1)}$

$\frac{1}{5}$ if $1 \leq x < 2$ $x_{(1)} \leq x < x_{(3)}$

$\frac{3}{5}$ if $2 \leq x < 3$ $x_{(3)} \leq x < x_{(4)}$

$\frac{4}{5}$ if $3 \leq x < 5$ $x_{(4)} \leq x < x_{(5)}$

1 if $5 \leq x$ $x_{(5)} \leq x$

or

0, $x < x_{(1)}$

$\frac{i}{n}$, $x_{(i)} < x < x_{(i+1)}$

1, $x \geq x_{(n)}$

Population CDF

$$F_x(x_0) = \frac{\sum_{i=1}^N I(x_i \leq x_0)}{N}$$

$$\text{Set} : Y_i = I(X_i \leq x_0)$$

$$Y_i \sim \text{Ber}(F_x(x_0))$$

$$\hat{F}_n(x_0) = \frac{\sum_{i=1}^n I(x_i \leq x_0)}{n}$$

$$E(\hat{F}_n(x_0)) = \frac{n}{n} E(I(x_1 \leq x_0)) = F_x(x_0)$$

$$\text{Var}(\hat{F}_n(x_0)) = \frac{1}{n^2} \sum_{i=1}^n \text{Var}(I(x_1 \leq x_0)) = \frac{F_x(x_0)(1 - F_x(x_0))}{n} \leq \frac{1}{4n}$$

$$\lim_{n \rightarrow \infty} \text{MSE}[\hat{F}_n(x_0)] \rightarrow 0$$

Therefore $\hat{F}_n(x_0)$ is a consistent estimator

$$x_1, \dots, x_{50} \sim \text{Uniform}(0, 2)$$

$$E_n(\hat{F}(1))?$$

$$Var(\hat{F}(1))?$$

$$n^{1/2}(\hat{F}_n(x_0) - F(x_0)) \sim N(0, F(x_0)[1 - F(x_0)])$$

$$\hat{p} \sim N(p, \frac{p(1-p)}{n})$$

$$\text{CI estimator } \pm Z_{1-\alpha}SD$$