

$X_1, X_2, \dots, X_n$  iid from distribution  
pdf  $f(x; \theta)$

Life of a lightbulb follow exp. dist.  
 $f(x) = \lambda e^{-\lambda}$

$\hat{\theta} = g(X_1, \dots, X_n)$  "estimator" for  $\theta$

what do we want from our estimators?

- Unbiased:  $E[\hat{\theta}] = \theta$
  - Efficient:  $\text{Var}(\hat{\theta})$  small
  - Consistent: as  $n \rightarrow \infty$   $\hat{\theta} \rightarrow \theta$
  - Sufficient: no additional info from  $X_1, \dots, X_n$  about  $\theta$  not already encoded by  $\hat{\theta}$ .
- } good.

Def Given a sequence of random variables  $X_n$ , and a random variable  $X$ , we say that  $X_n$  converges in probability to  $X$  if  $\forall \epsilon > 0$

$$\lim_{n \rightarrow \infty} P(|X_n - X| < \epsilon) = 1$$

Def  $\hat{\theta}$  is a consistent estimator for  $\theta$  if  $\hat{\theta}$  converges in probability to  $\theta$ .

Fact: if  $\hat{\theta}$  is unbiased and if  $\text{Var}(\hat{\theta}) \rightarrow 0$  as  $n \rightarrow \infty$   
 then  $\hat{\theta}$  is a consistent estimator.

$$P(|X - \mu| < k\sigma) \geq 1 - \frac{1}{k^2}$$

$$P(|\hat{\theta}_n - E[\hat{\theta}_n]| < k\sqrt{\text{Var}(\hat{\theta}_n)}) \geq 1 - \frac{1}{k^2}$$

$\theta$ 
 $s_n = \sqrt{\text{Var}(\hat{\theta}_n)}$

$$P(|\hat{\theta}_n - \theta| < \varepsilon) \geq 1 - \frac{1}{(k\varepsilon/s_n)^2} = 1 - \frac{s_n^2}{\varepsilon^2}$$

$s_n \rightarrow 0$

$$k\varepsilon = \frac{\varepsilon}{s_n}$$

$$\lim_{n \rightarrow \infty} P(|\hat{\theta}_n - \theta| < \varepsilon) \geq \lim_{n \rightarrow \infty} \left(1 - \frac{s_n^2}{\varepsilon^2}\right) = 1 \quad \square$$

Fact: If  $\hat{\theta}$  is asymptotically unbiased &  $\text{Var} \hat{\theta} \rightarrow 0$  as  $n \rightarrow \infty$   
 $\Rightarrow \hat{\theta}$  is consistent.

beginning  $P(|\hat{\theta}_n - \theta| < \varepsilon)$

$$E[\hat{\theta}_n] = \theta + b_n \text{ as } n \rightarrow \infty, b_n \rightarrow 0$$

$$|\hat{\theta}_n - \theta| = |\hat{\theta}_n - E[\hat{\theta}_n] + E[\hat{\theta}_n] - \theta| < \underbrace{|\hat{\theta}_n - E[\hat{\theta}_n]|}_{\text{same argument}} + |b_n|$$

$$\begin{aligned} P(|\hat{\theta}_n - \theta| < \varepsilon) &\geq P(|\hat{\theta}_n - \theta - b_n| + |b_n| < \varepsilon) \\ &\geq P(|\hat{\theta}_n - \theta - b_n| < \frac{\varepsilon}{2}, |b_n| < \frac{\varepsilon}{2}) \end{aligned}$$

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