

## ACM 157 Set 2

$$1. \mu = \frac{1}{N} \sum_{i=1}^N x_i = \frac{\#1}{N} \quad x_i \in \{0, 1\}$$

$$a) \sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$$

$$= \frac{1}{N} \sum_{i=1}^N (x_i^2 - 2x_i\mu + \mu^2)$$

$$= \frac{1}{N} \sum_{i=1}^N x_i - \frac{1}{N} \sum_{i=1}^N 2x_i\mu + \frac{1}{N} \sum_{i=1}^N \mu^2 \quad x_i^2 = x_i$$

$$= \mu - 2\frac{1}{N} N\mu \sum_{i=1}^N x_i + \frac{1}{N} \mu^2 \sum_{i=1}^N 1$$

$$= \mu - 2\mu^2 + \mu^2$$

$$= \mu - \mu^2$$

$$= \mu(1 - \mu)$$

$$b) s^2 = \sigma_n^2 \frac{N_n - n}{N_n - N} = (1 - \frac{1}{N}) \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x}_n)^2$$

$$= (1 - \frac{1}{N}) \frac{1}{n-1} \sum_{i=1}^n (x_i^2 - 2x_i\bar{x}_n + \bar{x}_n^2)$$

$$= (1 - \frac{1}{N}) \frac{1}{n-1} \left( \sum_{i=1}^n x_i - 2\bar{x}_n \sum_{i=1}^n x_i + \bar{x}_n^2 \sum_{i=1}^n 1 \right)$$

$$= (1 - \frac{1}{N}) \frac{n}{n-1} \left( \frac{1}{n} \sum_{i=1}^n x_i - 2\bar{x}_n \frac{1}{n} \sum_{i=1}^n x_i + \bar{x}_n^2 \frac{1}{n} n \right)$$

$$= (1 - \frac{1}{N}) \frac{n}{n-1} (\bar{x}_n - 2\bar{x}_n^2 + \bar{x}_n^2)$$

$$= (1 - \frac{1}{N}) \frac{n}{n-1} (\bar{x}_n - \bar{x}_n^2)$$

$$= (1 - \frac{1}{N}) \frac{n}{n-1} \bar{x}_n (1 - \bar{x}_n)$$

$$c) N=300 \quad n=90 \quad \#1=70 \quad \bar{x}_n = 70/90 \quad 95\% = 100(1-\alpha)\% \quad \alpha=0.05$$

$$I = \bar{x}_n \pm z_{1-\frac{\alpha}{2}} \text{se}[\bar{x}_n]$$

$$= 70/90 \pm z_{1-\frac{0.05}{2}} \frac{s}{\sqrt{n}} \sqrt{(1 - \frac{n-1}{N-1})}$$

$$= 70/90 \pm z_{0.975} \sqrt{(1 - \frac{1}{N}) \frac{n}{n-1} \bar{x}_n (1 - \bar{x}_n) \frac{1}{n}} \sqrt{(1 - \frac{n-1}{N-1})}$$

$$= 70/90 \pm 1.96 \sqrt{(1 - \frac{1}{300}) \frac{90}{89} \frac{70}{90} (1 - \frac{70}{90}) \frac{1}{90}} \sqrt{(1 - \frac{89}{299})}$$

$$= 70/90 \pm 0.0723$$

$$[0.706, 0.850]$$

$$2. 95\% \text{ interval length} \leq 0.04$$

$$2. z_{1-\frac{\alpha}{2}} \text{se}[\bar{x}_n] \leq 0.04$$

$$2. 1.96 \cdot \frac{s}{\sqrt{n}} \leq 0.04 \quad \text{ignore population correction}$$

$$2. 1.96 \cdot \frac{\sigma}{\sqrt{n}} \leq 0.04 \quad s \approx \sigma$$

$$2. 1.96 \cdot \frac{\sqrt{0.5(1-0.5)}}{\sqrt{n}} \leq 0.04 \quad \sigma = \sqrt{n(1-\mu)} = \sqrt{0.5(1-0.5)} \quad \text{worst-case variance}$$

$$n \geq 2401$$

$$4. X_1, \dots, X_n \text{ uniform on } [0, \theta] \quad \mu = \frac{\theta}{2} \quad \sigma^2 = \frac{\theta^2}{12} \quad V[\bar{X}_n] = \frac{\sigma^2}{n}$$

$$a) \hat{\theta} = 2\bar{X}_n$$

$$\text{bias}[\hat{\theta}] = E[\hat{\theta}] - \theta = E[2\bar{X}_n] - \theta = 2E[\bar{X}_n] - \theta$$

$$= 2 \frac{\theta}{2} - \theta = \theta - \theta = 0$$

$$\text{se}[\hat{\theta}] = \sqrt{V[\hat{\theta}]} = \sqrt{V[2\bar{X}_n]} = \sqrt{4V[\bar{X}_n]} = \sqrt{4 \frac{\sigma^2}{n}}$$

$$= \sqrt{4 \frac{\theta^2}{12n}} = \theta \sqrt{\frac{1}{3n}}$$

$$\text{MSE}[\hat{\theta}] = \text{bias}[\hat{\theta}]^2 + \text{se}[\hat{\theta}]^2 = \frac{\theta^2}{3n}$$

$$b) \hat{\theta} = X_{(n)} = \max\{X_1, \dots, X_n\}$$

$$\text{CDF: } F(x) = P(X_{(n)} \leq x) = P(X_1 \leq x)P(X_2 \leq x) \dots P(X_n \leq x)$$

$$= \frac{x}{\theta} \cdot \frac{x}{\theta} \dots \frac{x}{\theta} = \left(\frac{x}{\theta}\right)^n$$

$$f(x) = \frac{n x^{n-1}}{\theta^n}$$

$$E[\hat{\theta}] = \int_0^\theta x dF(x) = \int_0^\theta x f(x) dx = \int_0^\theta n \frac{x^n}{\theta^n} dx = \frac{n}{n+1} \frac{x^{n+1}}{\theta^n} \Big|_0^\theta$$

$$= \frac{n}{n+1} \frac{\theta^{n+1}}{\theta^n} = \frac{n}{n+1} \theta$$

$$\text{bias}[\hat{\theta}] = E[\hat{\theta}] - \theta = \frac{n}{n+1} \theta - \theta = -\frac{1}{n+1} \theta$$

$$E[\hat{\theta}^2] = \int_0^\theta x^2 dF(x) = \int_0^\theta x^2 f(x) dx = \int_0^\theta n \frac{x^{n+1}}{\theta^n} dx = \frac{n}{n+2} \frac{x^{n+2}}{\theta^n} \Big|_0^\theta$$

$$= \frac{n}{n+2} \frac{\theta^{n+2}}{\theta^n} = \frac{n}{n+2} \theta^2$$

$$V[\hat{\theta}] = E[\hat{\theta}^2] - E[\hat{\theta}]^2 = \frac{n}{n+2} \theta^2 - \left(\frac{n}{n+1} \theta\right)^2 = \frac{n(n+1)^2}{(n+2)(n+1)^2} \theta^2 - \frac{n^2(n+2)}{(n+1)(n+1)^2} \theta^2$$

$$= \frac{1}{(n+1)(n+1)^2} \theta^2 (n^3 + 2n^2 + n - n^3 - 2n^2) = \frac{n}{(n+2)(n+1)^2} \theta^2$$

$$\text{se}[\hat{\theta}] = \sqrt{V[\hat{\theta}]} = \theta \sqrt{\frac{n}{(n+2)(n+1)^2}}$$

$$\text{MSE}[\hat{\theta}] = \text{bias}[\hat{\theta}]^2 + \text{se}[\hat{\theta}]^2 = \left(-\frac{1}{n+1} \theta\right)^2 + \frac{n}{(n+2)(n+1)^2} \theta^2$$

$$= \frac{1}{(n+1)^2} \theta^2 + \frac{n}{(n+2)(n+1)^2} \theta^2 = \frac{2n+2}{(n+2)(n+1)^2} \theta^2 = \frac{2}{(n+2)(n+1)} \theta^2$$

c)

We can see that  $X_{(n)}$  is more efficient at  $n > 2$  because:

$$X_{(n)} \text{MSE} < 2\bar{X}_n \text{MSE}$$

$$\frac{2}{(n+2)(n+1)} \theta^2 \leq \frac{\theta^2}{3n}$$

$$6n \leq (n+2)(n+1)$$

$$n^2 + 3n + 2 - 6n \geq 0$$

$$n^2 - 3n + 2 \geq 0$$

$$(n-2)(n-1) \geq 0$$

equal at  $n=1, 2$

greater at  $n > 2$