

8.1 a  $\theta$  is m

$$L(\theta) = \prod_{i=0}^N P(d_i | \theta) = \prod_{i=1}^N \frac{1}{\sqrt{2\pi}\sigma} \exp^{-\frac{(x_i - \theta)^2}{2\sigma^2}}$$

$$\operatorname{argmax}_{\theta} L(\theta) = \operatorname{argmax}_{\theta} \log(L(\theta))$$

$$\frac{\partial}{\partial \theta} \log(L(\theta)) = 0$$

$$\log(L(\theta)) = \sum_{i=0}^N \log \frac{1}{\sqrt{2\pi}\sigma} + \frac{(x_i - \theta)^2}{2\sigma^2}$$

$$\frac{\partial}{\partial \theta} \log(L(\theta)) = \sum_{i=0}^N \frac{x_i - \theta}{\cancel{\sigma}} = 0$$

$$\sum_{i=0}^N x_i = n\theta$$

$$\theta = \frac{\sum_{i=0}^N x_i}{n} = \text{mean}(\{x_i\}) \quad \checkmark$$

b.  $\theta$  is  $\theta$ , m constant.

$$L(\theta) = \prod_{i=0}^N P(d_i | \theta) = \prod_{i=0}^N \frac{1}{\sqrt{2\pi}\theta} \exp^{-\frac{(x_i - m)^2}{2\theta^2}}$$

$$\operatorname{argmax}_{\theta} L(\theta) = \operatorname{argmax}_{\theta} \log(L(\theta))$$

$$\begin{aligned} \log(L(\theta)) &= \sum_{i=0}^N \log \frac{1}{\sqrt{2\pi}\theta} + \sum_{i=0}^N \frac{-(x_i - m)^2}{2\theta^2} \\ &= \sum_{i=0}^N -\log \sqrt{2\pi} - \log \theta + \frac{1}{2\theta^2} \cdot \sum_{i=0}^N (x_i - m)^2 \end{aligned}$$

$$\frac{\partial}{\partial \theta} \log(L(\theta)) = -\frac{n}{\theta^2} + \frac{1}{\theta^3} \sum_{i=0}^N (x_i - m)^2 = 0$$

$$n = \frac{1}{\theta^2} \sum_{i=0}^N (x_i - m)^2, \quad \theta^2 = \frac{\sum_{i=0}^N (x_i - m)^2}{n}$$

$$\theta = \sqrt{\frac{\sum_{i=0}^N (x_i - m)^2}{n}} = \text{std}(\{x_i\}) \quad \checkmark$$

C. If all numbers equal some value  $x$ , then  $M=x$

$$L(\theta) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\theta}} \cdot \exp -\frac{(x_i - \mu)^2}{2\theta^2}$$

$$= \prod_{i=1}^n \frac{1}{\sqrt{2\pi\theta}} \cdot \exp \theta$$

$$= \prod_{i=1}^n \frac{1}{\sqrt{2\pi\theta}} = \frac{1}{\sqrt{2\pi\theta}}, \text{ which is maximized with } \boxed{\theta = 0}$$

- 8.2a  $p(x) = \frac{\lambda^x e^{-\lambda}}{x!}, L(\theta) = \prod_{i=1}^n p(d_i | \theta)$

$$\arg \max L(\theta) = \arg \max \log(L(\theta))$$

$$\log(L(\theta)) = \sum_{i=0}^n (\log \lambda^{x_i} + \log e^{-\lambda} - \log x_i!)$$

$\theta$  is  $\lambda$

$$\log(L(\theta)) = \sum_{i=0}^n (x_i \log \theta + \theta - \log x_i!)$$

$$\frac{\partial}{\partial \theta} \log(L(\theta)) = \sum_{i=0}^n \left( \frac{x_i}{\theta} + -1 \right) = 0$$

$$\cancel{\sum_{i=0}^n} \frac{1}{\theta} \sum_{i=0}^n x_i = n.$$

$$\theta = \frac{\sum_{i=0}^n x_i}{n} = \text{mean}(\{x\}) \quad \checkmark$$

day 1:  $\text{mean}(\{3, 1, 4, 2\}) = \frac{2.5}{4} = \boxed{2.5}$

day 2:  $\text{mean}(\{2, 1, 2\}) = \frac{5}{3} = \boxed{1.67}$

day 3:  $\text{mean}(\{3, 2, 2, 4\}) = \frac{12}{6} = \boxed{2.4}$

b. day 4:  $\text{mean}(\{x\}) = \frac{13}{6} = \boxed{2.17}$

$$8.2 \text{ C. } \text{mean}(\{x_i\}) = \frac{\sum_{i=1}^n x_i}{n} = \frac{5+17+13+10}{4+3+5+6} = \boxed{2.22}$$

$$8.3 \text{ a. } P(D|\theta) = \prod_{i=1}^r (d_i | \theta) = \left(1 - \frac{\theta-36}{\theta}\right)^r \cdot \left(1 - \frac{36}{\theta}\right)$$

$$= \left(\frac{36}{\theta}\right)^r \cdot \left(\frac{\theta-36}{\theta}\right)$$

$$\frac{\partial}{\partial \theta} L(\theta) = 0 = \cancel{\frac{\partial}{\partial \theta} \ln L(\theta)}$$

$$-\frac{r \cdot 36^r}{\theta^{r+1}} \cdot \left(1 - \frac{36}{\theta}\right) + \frac{36}{\theta^2} \cdot \frac{36^r}{\theta^r} = 0$$

$$\frac{36^{r+1}}{\theta^{r+2}} = \frac{r \cdot 36^r}{\theta^{r+1}} \cdot \left(1 - \frac{36}{\theta}\right)$$

$$\frac{36}{\theta} = r - \frac{36r}{\theta}$$

$$(36 + 36r)\frac{1}{\theta} = r$$

$$\theta = \frac{36}{r} + 36$$

- b. This isn't very reliable because only 1 series was observed, indicating high variance. You should watch the wheel for several series of zeroes before making an estimate to improve accuracy (as is done in part C)

$$8.3c \quad L(\theta) = \prod_{i=1}^k P(d_i | \theta) = \prod_{i=1}^k \left(\frac{36}{\theta}\right)^{r_i} \left(1 - \frac{36}{\theta}\right)^{k-r_i}$$

Let  $R = \sum_{i=1}^k r_i$

$$L(\theta) = \left(\frac{36}{\theta}\right)^R \left(1 - \frac{36}{\theta}\right)^{k-R}$$

$$\arg \max L(\theta) = \arg \max \log(L(\theta))$$

$$\begin{aligned} \log L(\theta) &= \log \left(\frac{36}{\theta}\right)^R + \log \left(1 - \frac{36}{\theta}\right)^{k-R} \\ &= R(\log 36 - \log \theta) + k(\log(1 - \frac{36}{\theta})) \end{aligned}$$

$$\frac{\partial}{\partial \theta} \log L(\theta) = \frac{-R}{\theta} + k \left( \frac{36}{\theta^2} \cdot \frac{1}{1 - \frac{36}{\theta}} \right) = 0$$

$$\frac{36}{\theta^2} \cdot \frac{1}{1 - \frac{36}{\theta}} = \frac{R}{k\theta}$$

$$\frac{1}{1 - \frac{36}{\theta}} = \frac{\theta R}{36k}$$

$$1 - \frac{36}{\theta} = \frac{36k}{\theta R}$$

$$\theta - 36 = \frac{36k}{R}$$

$$\theta = 36 + \frac{36k}{\sum_{i=1}^{k-r} r_i}$$

$$= 36 + \frac{36}{\text{mean}(\{r_i\})}$$

$$8.4 \text{ a. } P(D|\theta) = \theta^b \cdot (1-\theta)^y \binom{b+y}{b}$$

$$= 1 \cdot 1 \cdot (1-\theta)$$

$$= 1 - \theta$$

$$\underset{\theta}{\operatorname{argmax}} \quad P(D|\theta) = 0$$

$$\boxed{\theta = 0}$$

$$\text{b. } P(D|\theta) = \binom{10}{3} \theta^3 (1-\theta)^7$$

$$\frac{\partial}{\partial \theta} L(\theta) = \cancel{(10)(3)(2)(1)(0)} = 0$$

$$= \binom{10}{3} (3\theta^2(1-\theta)^7 + 7(1-\theta)^6\theta^3) = 0$$

$$3\theta^2(1-\theta)^7 = 7(1-\theta)^6\theta^3$$

$$3 - 3\theta = 7\theta$$

$$\boxed{\theta = \frac{3}{10}}$$

$$8.5 \text{ a. } P(n|z) = \frac{z}{36+z}$$

$$P(n|z=0) = 0$$

$$P(n|z=1) = \frac{1}{37}$$

$$P(n|z=2) = \frac{1}{19}$$

$$P(n|z=3) = \frac{1}{13}$$

b. When you have observed no zeroes,  $P(z=0 \text{ obs})$  is non-zero

$$\text{c. } P(z|n) = \frac{P(n|z) \cdot p(z)}{P(n)}$$

$$P(z=0 \text{ obs}) = 0$$

$$P(z=1 \text{ obs}) = \frac{36(2)(\frac{1}{37})^2 (\frac{36}{37})^{34} \cdot 1}{P(n)} = \frac{\cancel{36}(2)(\frac{1}{37})^2 (\frac{36}{37})^{34}}{\cancel{36} P(n)} = \frac{.0363}{P(n)}$$

$$P(z=2 \text{ obs}) = \frac{36(2)(\frac{2}{38})^2 (\frac{36}{38})^{34} \cdot 2}{P(n)} = \frac{.111}{P(n)}$$

$$P(z=3 \text{ obs}) = \frac{36(2)(\frac{3}{39})^2 (\frac{36}{39})^{34} \cdot 3}{P(n)} = \frac{.073}{P(n)}$$

Probabilities need to sum to 1 b/c definition of distribution

$$\text{so } P(n) = .0363 + .111 + .073 = .2203$$

$$P(z=0 \text{ obs}) = 0$$

$$P(z=1 \text{ obs}) = .165$$

$$P(z=2 \text{ obs}) = .504$$

$$P(z=3 \text{ obs}) = .331$$

$$8.7 \text{ a. estimated popmean } \{\bar{x}\} = E(\{X^{(k)}\}) = \frac{\sum_{i=1}^K x_i}{K}$$

$$= \frac{199}{10} = \boxed{19.9}$$

$$\text{b. standard error} = \frac{\text{popstd}\{\bar{x}\}}{\sqrt{K}}$$

$$\text{popstd}\{\bar{x}\} \approx \text{std}(x)$$

$$= \sqrt{\frac{\sum_{i=1}^K (x_i - \bar{x})^2}{K-1}} = 3.51$$

$$\text{std}(\{X^{(k)}\}) = \frac{3.51}{\sqrt{10}} = \boxed{1.11}$$

$$\text{c. std}(X^{(k)}) = \frac{3.51}{\sqrt{K}} \leq .1$$

$$K \geq 35.1^2 \geq \boxed{1232} \text{ mice}$$

$$8.8 \text{ a. popstd}\{\bar{x}\} = \text{std}(\{\bar{x}\}) = \sqrt{\frac{\sum_{i=1}^K (\bar{x}_i - .3)^2}{K-1}} = .483$$

$$\text{std}(X^{(k)}) = \frac{.483}{\sqrt{10}} = \boxed{.153}$$

$$\text{b. } \frac{.483}{\sqrt{n}} \leq .05$$

$$\sqrt{n} \geq 9.66$$

$$n \geq 93.3$$

or 94 repeats