

4.

$$a. \Pr(x_1, x_2, \dots, x_n; \theta) = \prod_{i=1}^n \Pr(x_i; \theta) \quad \text{assuming independence}$$

$$= \prod_{i=1}^n (\theta \mathbf{1}_{x_i=1} + (1-\theta) \mathbf{1}_{x_i \neq 1})$$

$$\text{Note that } x_i = \mathbf{1}_{x_i=1}, \quad 1-x_i = \mathbf{1}_{x_i \neq 1}$$

$$= \prod_{i=1}^n (x_i \theta + (1-\theta)(1-x_i))$$

~~$$= \prod_{i=1}^n (x_i \theta + 1 - x_i - \theta + \theta x_i)$$~~

~~$$= \prod_{i=1}^n (2x_i \theta - x_i - \theta + 1)$$~~

$$= \theta^y (1-\theta)^{n-y}$$

where $y = \text{number of } x_i \text{ such that } x_i = 1$

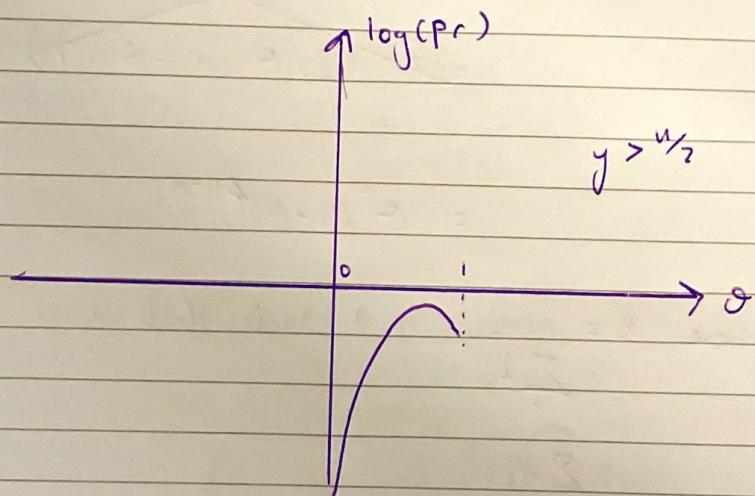
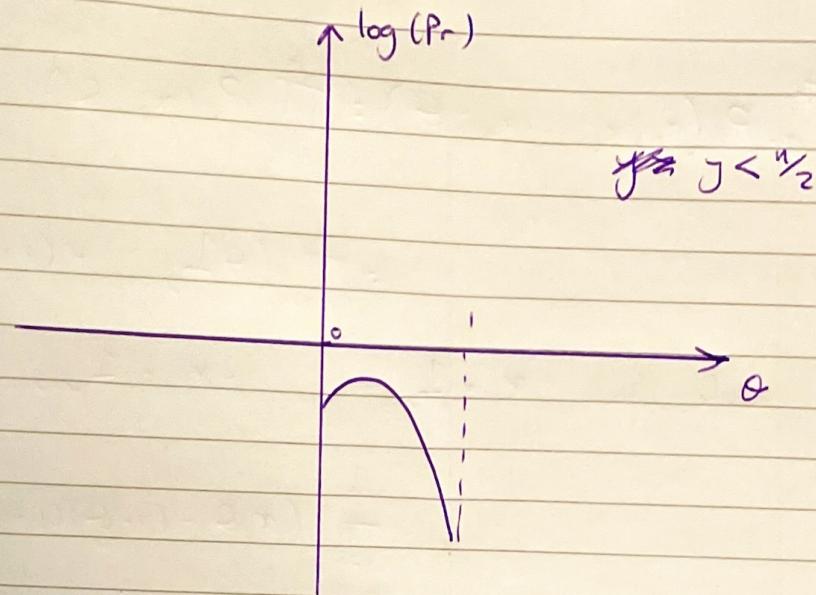
$$= \sum_{i=1}^n \mathbf{1}_{x_i=1}$$

$$= \sum_{i=1}^n x_i$$

$$= y$$

$$\therefore \Pr(x_1, \dots, x_n; \theta) = \theta^y (1-\theta)^{n-y}$$

$$b. \log(\Pr \dots) = \log(\theta^y (1-\theta)^{n-y}) \\ = y \log \theta + (n-y) \log(1-\theta)$$



$$c. \frac{d}{d\theta} \log(P_r \dots) = \frac{\gamma}{\theta} + \frac{y-n}{1-\theta} \stackrel{!}{=} 0$$

$$\therefore (1-\theta)\gamma + \theta(y-n) = 0$$

$$\therefore y - \theta y + \theta y - \theta n = 0$$

$$\therefore \theta = \gamma_n$$

$$\hat{\theta} = \max(\frac{1}{2}, \gamma_n)$$

5. Let X be the length of an unbroken chain of failures

$$X \sim \text{Geo}(1-\theta)$$

$$\Pr(X=k) = (1-\theta)^k \theta$$

$$\begin{aligned}\therefore \Pr(X \geq n) &= \sum_{k=n}^{\infty} \Pr(X=k) \\ &= \sum_{k=n}^{\infty} (1-\theta)^k \theta \\ &= 1 - \sum_{k=0}^{n-1} (1-\theta)^k \theta \\ &= 1 - \theta^n \frac{1 - (1-\theta)^n}{1 - (1-\theta)} \\ &= 1 - \theta^{n-1} (1 - (1-\theta)^n)\end{aligned}$$

Looking for some n

such that $\Pr(X \geq n) \leq 0.05$

$$\therefore 1 - \theta^{n-1} (1 - (1-\theta)^n) \leq 0.05$$

$$\therefore \theta^{n-1} (1 - (1-\theta)^n) \geq 0.95$$

~~∴~~ Not sure how to solve

$$1. P(X_3=r | X_0=g) = \sum_{x_2} P(X_3=r | X_2=x_2) P(X_2=x_2 | X_0=g)$$

by the law of total probability

$$= \sum_{x_2} P_{x_2|r} P(X_2=x_2 | X_0=g)$$

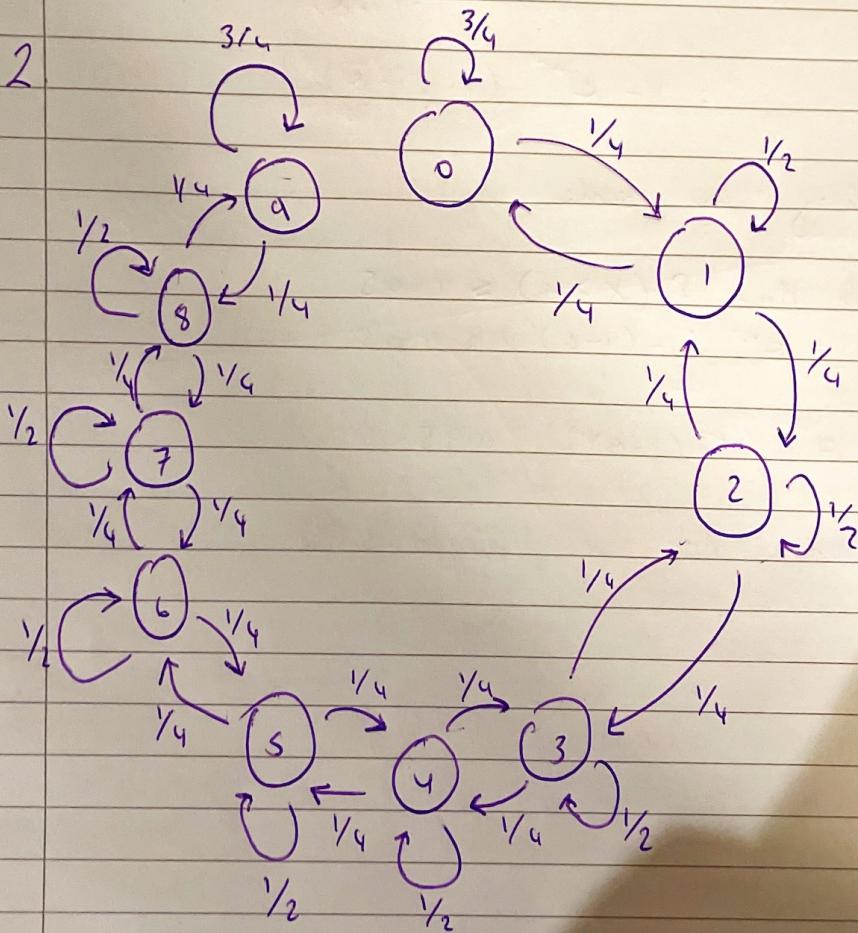
$$= \sum_{x_2} P_{x_2|r} \sum_{x_1} P(X_2=x_2 | X_1=x_1) P(X_1=x_1 | X_0=g)$$

by the law of total probability

$$= \sum_{x_2} P_{x_2|r} \sum_{x_1} P_{x_1|x_2} P_{g|x_1}$$

$$= \sum_{x_1, x_2} P_{g|x_1} P_{x_1|x_2} P_{x_2|r}$$

2



3 Let π be a stationary distribution of the Markov chain.

$$\therefore \mathbb{P}(X_0 = a) = \pi_a$$

$$\mathbb{P}(X_0 = b) = \pi_b = 1 - \pi_a$$

$$\mathbb{P}(X_1 = a) = \sum_{x_0} \mathbb{P}(X_1 = a | X_0 = x_0) \mathbb{P}(X_0 = x_0)$$

$$= \mathbb{P}(X_1 = a | X_0 = a) \pi_a + \mathbb{P}(X_1 = a | X_0 = b) \pi_b$$

$$= (1-\alpha)\pi_a + \beta(1-\pi_a)$$

$$= \pi_a - \alpha\pi_a + \beta - \beta\pi_a$$

$$\therefore \pi_a$$

$$\therefore \beta - \pi_a(\alpha + \beta) = 0$$

$$\therefore \pi_a = \frac{\beta}{\alpha + \beta}, \quad \pi_b = \frac{\alpha}{\alpha + \beta}$$

This is a unique solution.

5a.

7. ~~Hot Counter~~

Let π = a stationary distribution for X such that

$$\Pr_{X_0}(k) = \pi_k$$

$$\Pr_{X_1}(k) = \mu + \lambda(\Pr_{X_0}(k) - \mu) \quad \text{by the shifting properties of a normal distribution}$$

$$= \mu + \lambda(\pi_k - \mu)$$

$$= \pi_k$$

$$\therefore \mu + \lambda\pi_k - \pi_k - \lambda\mu = 0$$

$$\therefore \pi_k(\lambda - 1) = \mu(\lambda - 1)$$

$$\therefore \pi_k = \mu$$

$$\therefore \pi \sim N(\mu, 0)$$

?? doesn't seem right