

HW5.

1.

| x_n | | | y_n | g | f_1 | f_2 | f_3 | f_4 | f_5 | f_6 | f_7 | f_8 |
|-------|---|---|-------|-----|-------|-------|-------|-------|-------|-------|-------|-------|
| 0 | 0 | 0 | ○ | ○ | ○ | ○ | ○ | ○ | ○ | ○ | ○ | ○ |
| 0 | 0 | 1 | ● | ● | ● | ● | ● | ● | ● | ● | ● | ● |
| 0 | 1 | 0 | ● | ● | ● | ● | ● | ● | ● | ● | ● | ● |
| 0 | 1 | 1 | ○ | ○ | ○ | ○ | ○ | ○ | ○ | ○ | ○ | ○ |
| 1 | 0 | 0 | ● | ● | ● | ● | ● | ● | ● | ● | ● | ● |
| 1 | 0 | 1 | | ? | ○ | ○ | ○ | ○ | ● | ● | ● | ● |
| 1 | 1 | 0 | | ? | ○ | ○ | ● | ● | ○ | ○ | ● | ● |
| 1 | 1 | 1 | | ? | ○ | ● | ○ | ● | ○ | ● | ○ | ● |

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⋮

a) i) $g = [\bullet \bullet \bullet \bullet \bullet \bullet \bullet \bullet] = h_1.$

ii) Match 3 : $\{f_8\}$

2 : $\{f_4, f_6, f_7\}$

1 : $\{f_2, f_3, f_5\}$

0 : $\{f_1\}$

b) i) $g = [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0] = h_2.$

ii) Match 3 : $\{f_1\}$

2 : $\{f_2, f_3, f_5\}$

1 : $\{f_4, f_6, f_7\}$

0 : $\{f_8\}$

1. c) i).

$$g = [0 \cdot \cdot \cdot 0 \cdot | 0 \cdot \cdot]$$

ii) Match 3: (f_2)

$$2: (f_1, f_4, f_6)$$

$$1: (f_3, f_5, f_8)$$

$$0: (f_7)$$

d) i).

$$g = [0 \cdot \cdot \cdot 0 \cdot | \cdot \cdot \cdot 0]$$

ii) Match 3: (f_7)

$$2: (f_3, f_5, f_8)$$

$$1: (f_1, f_4, f_6)$$

$$0: (f_2)$$

$$3. \quad X_n \sim \text{Bernoulli}(0.5), \quad \bar{X}_N = \left(\frac{1}{N}\right) \sum_{n=1}^N X_n$$

$$v) \text{ Show } \mathbb{P}(\bar{X}_N - \mu \geq \varepsilon) \leq 2^{-\beta N}.$$

$$\beta = 1 + \left(\frac{1}{2} + \varepsilon\right) \log_2 \left(\frac{1}{2} + \varepsilon\right) + \left(\frac{1}{2} - \varepsilon\right) \log_2 \left(\frac{1}{2} - \varepsilon\right), \quad \mu = 0.5.$$

Chernoff bound states that:

$$\begin{aligned} \mathbb{P}(\bar{X}_N - \mu \geq \varepsilon) &= \mathbb{P}\left(\frac{1}{N} \sum_{n=1}^N X_n - \mu \geq \varepsilon\right) \\ &= \mathbb{P}\left(\sum_{n=1}^N X_n \geq (\varepsilon + \mu)N\right) \\ &= \mathbb{P}\left(e^{s \sum_{n=1}^N X_n} \geq e^{s(\varepsilon + \mu)N}\right) \\ &\leq \frac{\mathbb{E}\left[e^{s \sum_{n=1}^N X_n}\right]}{e^{s(\varepsilon + \mu)N}} \\ &= \frac{\mathbb{E}[e^{sX_n}]^N}{e^{s(\varepsilon + \mu)N}} \\ &= \left[\frac{\mathbb{E}[e^{sX_n}]}{e^{s(\varepsilon + \mu)}}\right]^N. \end{aligned}$$

$$\mathbb{E}[e^{sX_n}] = e^s \mu + 1 - \mu = M_{X_n}(s)$$

$$\mathbb{P}(\bar{X}_N - \mu \geq \varepsilon) \leq e^{(-s(\varepsilon + \mu) + \log(e^s \mu + 1 - \mu))N}.$$

$$\frac{d}{ds} \left\{ e^{-s(\varepsilon + \mu) + \log(e^s \mu + 1 - \mu)} \right\} = 0.$$

$$\frac{d}{ds} \left\{ -s(\varepsilon + \mu) + \log(e^s \mu + 1 - \mu) \right\} = 0.$$

$$\Sigma + \mu - \frac{\mu e^{\Sigma}}{\mu e^{\Sigma} + 1 - \mu} = 0.$$

$$\frac{\mu e^{\Sigma}}{\Sigma + \mu} = \mu e^{\Sigma} + 1 - \mu.$$

$$\mu e^{\Sigma} \left(\frac{1}{\Sigma + \mu} - 1 \right) = 1 - \mu.$$

$$1 - \mu = \mu.$$

$$\mu e^{\Sigma} \left(\frac{1 - \Sigma - \mu}{\Sigma + \mu} \right) = 1 - \mu$$

$$e^{S_{\min}} = \frac{(\Sigma + \mu)}{(\mu - \Sigma)}, \quad S_{\min} = \log \left[\frac{\Sigma + \mu}{\mu - \Sigma} \right].$$

sub S_{\min} back

$$P(\bar{X}_N - \mu \geq \Sigma) \leq \exp \left\{ -\log \left[\frac{(\Sigma + \mu)}{(\mu - \Sigma)} \right] (\Sigma + \mu) + \log \left[\frac{(\Sigma + \mu)\mu}{(\mu - \Sigma)} + \mu \right] \right\}^N$$

$$= \exp \left\{ -\log \left[\frac{\Sigma + \mu}{\mu - \Sigma} \right] (\Sigma + \mu) + \log \left[\frac{\mu \cdot 2\mu}{\mu - \Sigma} \right] \right\}^N.$$

$$= \exp \left\{ -\log \left[\frac{\Sigma + \mu}{\mu - \Sigma} \right] (\Sigma + \mu) + \log \left[\frac{\mu}{\mu - \Sigma} \right] \right\}^N.$$

= I don't know where you pulled the 2 out of, but ok.

Project: still trying to figure out
the code for N2N.