Tsinghua-Berkeley Shenzhen Institute Information Theory and Statistical Learning Fall 2020

Homework 3

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- Acknowledgments: This template takes some materials from course CSE 547/Stat 548 of Washington University: https://courses.cs.washington.edu/courses/cse547/17sp/index.html.
- I certify that all solutions are entirely in my words and that I have not looked at another student's solutions. I have credited all external sources in this write up.

3.1.

$$I(X_{1},...,X_{n};Y_{1},...,Y_{m})$$

$$= \sum_{i=1}^{n} I(X_{i};Y_{1},...,Y_{m}|X_{i-1},...,X_{1})$$

$$= \sum_{i=1}^{n} I(Y_{1},...,Y_{m};X_{i}|X_{i-1},...,X_{1})$$

$$= \sum_{i=1}^{n} (H(Y_{1},...,Y_{m}|X_{i-1},...,X_{1}) - H(Y_{1},...,Y_{m}|X_{i},...,X_{1}))$$

$$= \sum_{i=1}^{n} (\sum_{j=1}^{m} H(Y_{j}|Y_{j-1},...,Y_{1}|X_{i-1},...,X_{1}) - \sum_{j=1}^{m} H(Y_{j}|Y_{j-1},...,Y_{1}|X_{i},...,X_{1}))$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{m} (H(Y_{j}|Y_{j-1},...,Y_{1}|X_{i-1},...,X_{1}) - \sum_{j=1}^{m} H(Y_{j}|Y_{j-1},...,Y_{1}|X_{i},...,X_{1}))$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{m} I(X_{i};Y_{j}|X_{1},...,X_{i-1};Y_{1},...,Y_{j-1})$$

$$(1)$$

3.2. (a) Define
$$X = \begin{cases} 0 & w.p.\frac{1}{2} \\ 1 & w.p.\frac{1}{2} \end{cases}$$
, $Y = 2X$ and $Z = 2Y$. Then
$$I(X;Y) = H(X) + H(Y) - H(X,Y) = 1.$$
$$I(X;Y|Z) = H(X|Z) - H(X|Y,Z)$$
$$= H(X,Z) - H(Z) - (H(X,Y,Z) + H(Y,Z))$$
$$= 0$$
$$I(X;Y|Z) < I(X;Y).$$
 (2)

(b) Define
$$X=Y=\left\{\begin{array}{ll} 0 & w.p.\frac{1}{2}\\ 1 & w.p.\frac{1}{2} \end{array}\right.,\quad Z=X+Y\ X$$
 and Y are independent. Then $I(X;Y)=0.$

$$I(X;Y|Z) = H(X|Z) - H(X|Y,Z)$$

$$= H(X,Z) - H(Z) - (H(X,Y,Z) + H(Y,Z))$$

$$= 2 - \frac{2}{3} - 2 + 2$$

$$= \frac{1}{2}$$
(3)

I(X;Y|Z) > I(X;Y).

3.3. (a) Since Z_1, \ldots, Z_n are i.i.d and $Bern(\frac{1}{2})$ distributed, we can get

$$H(Z_i) = -\sum_{i=1}^{2} \frac{1}{2} \log_2 \left(\frac{1}{2}\right) = 1 \quad \forall i \in 1, 2, \dots, n$$

$$\to H(X_A) = H\left((Z_i)_{i \in A}\right)$$

$$= \sum_{i \in A} H(Z_i)$$

$$= |A| \tag{4}$$

(b)

$$H(X_{A_{1}}, X_{A_{2}}) = H((Z_{i})_{i \in A_{1}}, (Z_{i})_{i \in A_{2}})$$

$$= H((Z_{i})_{i \in A_{1} \cup A_{2}})$$

$$= |A_{1} \cup A_{2}|$$

$$H(X_{A_{1}} | X_{A_{2}}) = H(X_{A_{1}}, X_{A_{2}}) - H(X_{A_{2}})$$

$$= |A_{1} \cup A_{2}| - |A_{2}|$$

$$= |A_{1} \setminus A_{2}|$$

$$I(X_{A_{1}}; X_{A_{2}}) = H(X_{A_{1}}) - H(X_{A_{1}} | X_{A_{2}})$$

$$= |A_{1}| - |A_{1} \setminus A_{2}|$$

$$= |A_{1} \cap A_{2}|$$
(5)

3.4. (a) When p = 1,

$$P(x,y) = \begin{cases} \frac{1}{2} & |x| + |y| \le 1 \\ 0 & else \end{cases} . \tag{6}$$

Let x = 0, y = 1, we have $P(x, y) = \frac{1}{2}, P_X(0) = 1, P_Y(1) = 0$. Then $P(1, 0) \neq P_X(0) * P_Y(1)$ and X and Y are not independent.

(b) When $p = \frac{1}{2}$,

$$H(X,Y) = \iint_{S} \frac{3}{2} \log_{2} \frac{2}{3} dx dy = \log_{2} \frac{2}{3} \quad S: |x|^{\frac{1}{2}} + |y|^{\frac{1}{2}} \le 1$$

$$H(Y) = H(X) = 2 * \int_{0}^{1} \frac{3 * 2 * (1 - x^{\frac{1}{2}})^{2}}{2} \log_{2} \frac{2}{3 * 2 * (1 - x^{\frac{1}{2}})^{2}} dx =$$
(7)

Thus,
$$I(X,Y) = H(X) + H(Y) - H(X,Y) =$$
.
When $p = 1$, as the same procedure, $I(X,Y) = H(X) + H(Y) - H(X,Y) = 1 + 1 - 1 = 1$.
When $p = \infty$, as the same procedure, $I(X,Y) = H(X) + H(Y) - H(X,Y) = 1 + 1 - 2 = 1$.

(c) I(X;Y) converge to ∞ .

3.5.

$$D\left(\mathcal{N}\left(\boldsymbol{m}_{1}, \boldsymbol{\Sigma}_{1}\right) \| \mathcal{N}\left(\boldsymbol{m}_{0}, \boldsymbol{\Sigma}_{0}\right)\right)$$

$$= \int \left[\frac{1}{2} \log \frac{|\boldsymbol{\Sigma}_{1}|}{|\boldsymbol{\Sigma}_{0}|} - \frac{1}{2} \left(x - \boldsymbol{m}_{0}\right)^{T} \boldsymbol{\Sigma}_{0}^{-1} \left(x - \boldsymbol{m}_{0}\right) + \frac{1}{2} \left(x - \boldsymbol{m}_{1}\right)^{T} \boldsymbol{\Sigma}_{1}^{-1} \left(x - \boldsymbol{m}_{1}\right)\right] \times p(x) dx$$

$$= \frac{1}{2} \log \frac{|\boldsymbol{\Sigma}_{1}|}{|\boldsymbol{\Sigma}_{0}|} - \frac{1}{2} \operatorname{tr} \left\{ E\left[\left(x - \boldsymbol{m}_{0}\right) \left(x - \boldsymbol{m}_{0}\right)^{T}\right] \boldsymbol{\Sigma}_{0}^{-1} \right\} + \frac{1}{2} E\left[\left(x - \boldsymbol{m}_{1}\right)^{T} \boldsymbol{\Sigma}_{1}^{-1} \left(x - \boldsymbol{m}_{1}\right)\right]$$

$$= \frac{1}{2} \log \frac{|\boldsymbol{\Sigma}_{1}|}{|\boldsymbol{\Sigma}_{0}|} - \frac{1}{2} \operatorname{tr} \left\{ I_{n} \right\} + \frac{1}{2} \left(\boldsymbol{m}_{0} - \boldsymbol{m}_{1}\right)^{T} \boldsymbol{\Sigma}_{1}^{-1} \left(\boldsymbol{m}_{0} - \boldsymbol{m}_{1}\right) + \frac{1}{2} \operatorname{tr} \left\{ \boldsymbol{\Sigma}_{1}^{-1} \boldsymbol{\Sigma}_{0} \right\}$$

$$= \frac{1}{2} \left[\log \frac{|\boldsymbol{\Sigma}_{1}|}{|\boldsymbol{\Sigma}_{0}|} - n + \operatorname{tr} \left\{ \boldsymbol{\Sigma}_{1}^{-1} \boldsymbol{\Sigma}_{0} \right\} + \left(\boldsymbol{m}_{1} - \boldsymbol{m}_{0}\right)^{T} \boldsymbol{\Sigma}_{1}^{-1} \left(\boldsymbol{m}_{1} - \boldsymbol{m}_{0}\right) \right]$$
(8)

- (a) When Σ_0 is non-singular.
- (b) Based on 8, $D(\mathcal{N}(\boldsymbol{m}, \boldsymbol{\Sigma}) || \mathcal{N}(0, \boldsymbol{I}_n)) = \frac{1}{2} \left[\log |\boldsymbol{\Sigma}| - n + \operatorname{tr} \left\{ \boldsymbol{\Sigma}^{-1} \right\} + \boldsymbol{m}^T \boldsymbol{\Sigma}^{-1} \boldsymbol{m} \right].$
- (c) Based on 8, $D\left(\mathcal{N}\left(\boldsymbol{m}_{1}, \boldsymbol{\Sigma}_{1}\right) \| \mathcal{N}\left(\boldsymbol{m}_{0}, \boldsymbol{\Sigma}_{0}\right)\right) = \frac{1}{2} \left[\log \frac{|\boldsymbol{\Sigma}_{1}|}{|\boldsymbol{\Sigma}_{0}|} n + \operatorname{tr}\left\{\boldsymbol{\Sigma}_{1}^{-1} \boldsymbol{\Sigma}_{0}\right\} + \left(\boldsymbol{m}_{1} \boldsymbol{m}_{0}\right)^{T} \boldsymbol{\Sigma}_{1}^{-1} \left(\boldsymbol{m}_{1} \boldsymbol{m}_{0}\right) \right].$

3.6. (a)

$$\sum_{i=1}^{k} f(P_i) + \sum_{i=1}^{k} f(Q_i)$$

$$\geq 2 \sum_{i=1}^{k} f\left(\frac{P_i + Q_i}{2}\right)$$
(9)

$$\Rightarrow \sum_{i=1}^{k} f\left(P_{i}\right) - \sum_{i=1}^{k} f\left(\frac{P_{i} + Q_{i}}{2}\right) \ge \sum_{i=1}^{k} f\left(\frac{P_{i} + Q_{i}}{2}\right) - \sum_{i=1}^{k} f\left(Q_{i}\right)$$

Thus, $\sum_{i=1}^{k} f(P_i) \le \sum_{i=1}^{k} f(Q_i)$.

(b) Define $f(x) = x \log x$, which is a convex function because $f''(x) = \frac{1}{x} \le 0$. Based on (a), it has

$$\sum_{i=1}^{k} f(P_i) \le \sum_{i=1}^{k} f(Q_i)$$

$$\Rightarrow -\sum_{i=1}^{k} P_i \log P_i \ge -\sum_{i=1}^{k} Q_i \log Q_i$$

$$\Rightarrow H(P) \ge H(Q)$$
(10)

3.7. (a)
$$C(X_{1}, X_{2}, ..., X_{n})$$

$$=D\left(P_{X^{n}} \| \prod_{i=1}^{n} P_{X_{i}}\right)$$

$$=\sum_{x_{1} \in X} \cdots \sum_{x_{n} \in X} P_{X^{n}}(x_{1}, ..., x_{n}) \log \frac{P_{X^{n}}(x_{1}, ..., x_{n})}{\prod_{i=1}^{n} P_{X_{i}}(x_{i})}$$

$$=-\sum_{x_{1} \in X} \cdots \sum_{x_{n} \in X} P_{X^{n}}(x_{1}, ..., x_{n}) \log \prod_{i=1}^{n} P_{X_{i}}(x_{i})$$

$$+\sum_{x_{1} \in X} \cdots \sum_{x_{n} \in X} P_{X^{n}}(x_{1}, ..., x_{n}) \log P_{X^{n}}(x_{1}, ..., x_{n})$$

$$=-\sum_{i=1}^{n} \sum_{x_{1} \in X} \cdots \sum_{x_{n} \in X} P_{X^{n}}(x_{1}, ..., x_{n}) \log P_{X^{n}}(x_{1}, ..., x_{n})$$

$$=-\sum_{i=1}^{n} \sum_{x_{i} \in X} P_{X_{i}}(x_{i}) \log P_{X_{i}}(x_{i})$$

$$+\sum_{x_{1} \in X} \cdots \sum_{x_{n} \in X} P_{X^{n}}(x_{1}, ..., x_{n}) \log P_{X^{n}}(x_{1}, ..., x_{n})$$

$$=\sum_{i=1}^{n} \sum_{x_{i} \in X} P_{X_{i}}(x_{i}) \log P_{X_{i}}(x_{i})$$

$$+\sum_{x_{1} \in X} \cdots \sum_{x_{n} \in X} P_{X^{n}}(x_{1}, ..., x_{n}) \log P_{X^{n}}(x_{1}, ..., x_{n})$$

$$=\sum_{i=1}^{n} H(X_{i}) - H(X_{1}, ..., X_{n})$$

$$=\sum_{i=1}^{n} H(X_{i}) - \sum_{i=1}^{n} H(X_{i}|X_{1}, ..., X_{i-1})$$

$$=\sum_{i=1}^{n} (H(X_{i}) - H(X_{i}|X_{1}, ..., X_{i-1}))$$

$$=\sum_{i=1}^{n-1} (H(X_{i+1}) - H(X_{i+1}|X_{1}, ..., X_{i}))$$

$$=\sum_{i=1}^{n-1} I(X^{i}; X_{i+1})$$
(11)

- (b) When n random variables are independent.
- 3.8. (a)
 - (b)