

HW2 notes.

2.1

$$H(p) = -p \log p - (1-p) \log (1-p)$$

a)

$H(p)$: Von-Neumann entropy

$H(X)$: Shannon entropy.

Information gain: (Classical mutual info.)

$$I(X;Y) = H(X) - \underbrace{H(X|Y)}_{H(XY) - H(Y)}$$

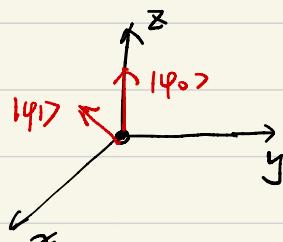
Y is measurement outcomes.

$$p(x,y) = p(y|x) p(x)$$

$$p(x) \text{ known, } p(y|x) = \text{tr}(E_y p(x)) .$$

$$p(y) = \sum_x p(x,y) .$$

$$|\psi_0\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix} , \quad |\psi_1\rangle = \begin{pmatrix} \cos(\theta/2) \\ \sin(\theta/2) \end{pmatrix} .$$



$$\rho = \frac{1}{2} |\varphi_0\rangle\langle\varphi_0| + \frac{1}{2} |\varphi_1\rangle\langle\varphi_1|$$

$$p(x_0) = p(x_1) = \frac{1}{2} .$$

$$p(0|x_0) = 1, \quad p(1|x_0) = 0$$

$$p(0|x_1) = \cos^2(\theta/2), \quad p(1|x_1) = \sin^2(\theta/2).$$

$$p(0, x_0) = \frac{1}{2}, \quad p(1, x_0) = 0$$

$$p(0, x_1) = \frac{1}{2} \cos^2(\theta/2), \quad p(1, x_1) = \frac{1}{2} \sin^2(\theta/2)$$

Information gain is $I(x; Y) = H(x) - H(x|Y)$.

$$H(x) = -\log \frac{1}{2} = \log 2 = 1.$$

$$H(x|Y=0) = -p_0 \log p_0 - p_1 \log p_1$$

$$= \frac{1}{1+\cos^2\theta/2} \log \frac{1+\cos^2\theta/2}{1+\cos^2\theta/2} + \frac{\cos^2\theta/2}{1+\cos^2\theta/2} \log \frac{\cos^2\theta/2}{1+\cos^2\theta/2}$$

$$H(x|Y=1) = 0.$$

$$p_Y(Y=0) = \frac{1}{2} (1+\cos^2\theta/2). \quad p_Y(Y=1) = \frac{1}{2} \sin^2\theta/2$$

$$\begin{aligned}
 \text{Therefore, } H(X|Y) &= \frac{1}{2} \log(1 + w^2 \theta/2) + \frac{1}{2} w^2 \theta/2 \log \frac{1 + w^2 \theta/2}{w^2 \theta/2} \\
 &= \frac{1}{2} (1 + w^2 \theta/2) \log (1 + w^2 \theta/2) \\
 &\quad - \frac{1}{2} w^2 \theta/2 \log w^2 \theta/2.
 \end{aligned}$$

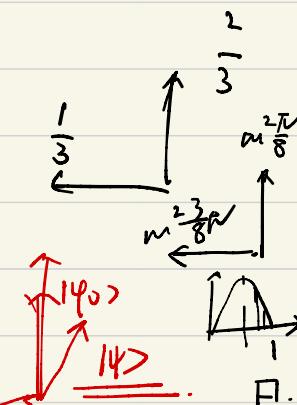
$$I(X;Y) = H(X) - H(X|Y)$$

$$= 1 - \frac{1}{2} (1 + w^2 \theta/2) \log (1 + w^2 \theta/2)$$

$$+ \frac{1}{2} w^2 \theta/2 \log (w^2 \theta/2)$$

$$\theta = 0, \quad I = 1 - 1 + 0 = 0.$$

$$\theta = \pi, \quad I = 1 - 0 + 0 = 1.$$



$$b) \quad p(x_0) = p(x_1) = \frac{1}{2}$$

$$\langle 1\psi_1 \rangle$$

$$p(y=0 | x = x_0) = \text{tr}(|1\psi\rangle\langle 4| \psi_0 \rangle\langle \varphi_0 |) = \cos^2\left(\frac{\theta}{4} - \frac{\pi}{4}\right)$$

$$p(y=1 | x = x_0) = \text{tr}(|1\psi\rangle\langle \psi^\perp| \psi_0 \rangle\langle \varphi_0 |) = \cos^2\left(\frac{\theta}{4} + \frac{\pi}{4}\right).$$

$$\begin{aligned}
 p(y=0 | x = x_1) &= \left[\cos(\theta/4 - \pi/4) \cos(\theta/2) + \sin(\theta/4 - \pi/4) \sin(\theta/2) \right]^2 \\
 &= \cos^2\left(\frac{\theta}{4} + \frac{\pi}{4}\right)
 \end{aligned}$$

$$\begin{aligned}
 p(y=1 | x = x_1) &= \left[\cos(\theta/4 + \pi/4) \cos(\theta/2) + \sin(\theta/4 + \pi/4) \sin(\theta/2) \right]^2 \\
 &= \cos^2\left(\frac{\theta}{4} - \frac{\pi}{4}\right).
 \end{aligned}$$

$$\begin{aligned}
 p_{x_1|y=0}(x_0 | y=0) &= w^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) \quad p_{x_1|y=0}(x_1 | y=0) = w^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right) \\
 \dots \quad y=1 \dots \quad w^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right) \quad \dots \quad y=0 \dots \quad w^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right)
 \end{aligned}$$

$$\text{Thus } P_Y(Y=0) = \frac{1}{2} \left(m^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right) + m^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) \right) = \frac{1}{2}$$

$$P_Y(Y=1) = \frac{1}{2} \left(m^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right) + m^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) \right) = \frac{1}{2}$$

$$\left. \begin{aligned} & \frac{1 + m \left(\frac{\theta}{2} + \frac{\pi}{2} \right)}{2} + \frac{1 + m \left(\frac{\theta}{2} - \frac{\pi}{2} \right)}{2} \\ &= 1 + \frac{1}{2} \left(\underbrace{m \left(\frac{\theta}{2} + \frac{\pi}{2} \right)}_{-\sin \frac{\theta}{2}} + \underbrace{m \left(\frac{\theta}{2} - \frac{\pi}{2} \right)}_{\sin \frac{\theta}{2}} \right) \\ &= 1. \end{aligned} \right\}$$

$$\Rightarrow H(X|Y=0) = -m^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) \log m^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) - m^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right) \log m^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right).$$

$$H(X|Y=1) = H(X|Y=0).$$

$$\text{Then } H(X|Y) = H(X|Y=0) \text{ since } P(Y=0) = P(Y=1) = \frac{1}{2}$$

$$I(X;Y) = 1 + m^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) \log m^2 \left(\frac{\theta}{4} - \frac{\pi}{4} \right) + m^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right) \log m^2 \left(\frac{\theta}{4} + \frac{\pi}{4} \right)$$

$$\theta = \pi \Rightarrow I(X;Y) = 1$$

$$\theta = 0 \Rightarrow I(X;Y) = 0.$$

□.

$$c) \quad \theta = \frac{\pi}{2} :$$

$$I_1(X;Y) \simeq 0.3113 \text{ bits.}$$

$$I_2(X;Y) \simeq 0.3991 \text{ bits}$$

$$\theta = \frac{2\pi}{3}$$

$$I_1(X;Y) \simeq 0.5488 \text{ bits}$$

$$I_2(X;Y) \simeq 0.6454 \text{ bits}$$

a	b	c
$\frac{1}{3}$	$\left(\begin{matrix} 0 \\ \frac{1}{2} \end{matrix} \right)$	$\frac{1}{2}$
$\frac{1}{3}$	$\left(\begin{matrix} \frac{1}{2} \\ 0 \end{matrix} \right)$	$\frac{1}{2}$
$\frac{1}{3}$	$\left(\begin{matrix} \frac{1}{2} \\ \frac{1}{2} \end{matrix} \right)$	0

$\rightarrow P(Y=a) = \frac{1}{3}$.

d)

$$I(X;YZ) = H(X) - H(X|YZ)$$

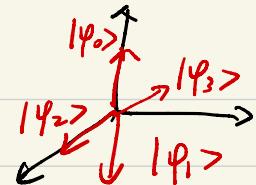
$$= H(X) - H(X|Y) \\ + H(X|Y) - H(X|YZ).$$

$$= \log_2 3 - 1 + 0.6454 \quad (\theta = \frac{2}{3}\pi)$$

$$\simeq 1.23036 \text{ bits}$$

2.2

$$\theta_1 = \pi/2, \theta_2 = -\pi/2$$



a) $P_x(x) = \frac{1}{4}$ for $x = 0, 1, 2, 3$

$$P_{Y|x=0}(y=0) = 1, P_{Y|x=0}(y=1) = 0.$$

$$P_{Y|x=1}(y=0) = 0, P_{Y|x=1}(y=1) = 1$$

$$P_{Y|x=2}(y=0) = P_{Y|x=2}(y=1) = \frac{1}{2}$$

$$P_{Y|x=3}(y=0) = P_{Y|x=3}(y=1) = \frac{1}{2}.$$

Then $H(x) = \log_2 4 = 2$ bits.

$$H(x|Y) = \langle H(x|Y=y) \rangle_y$$

$$= \frac{1}{2} \log_2 2 + \frac{1}{4} \times 2 \log_2 4 = \frac{1}{2} + 1 = \frac{3}{2}.$$

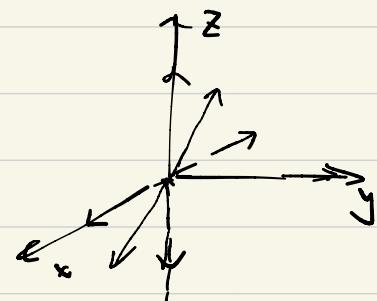
$$\Rightarrow I(x;Y) = H(x) - H(x|Y) = \frac{1}{2} \text{ bits.}$$

$$b) Z' = |\psi\rangle\langle\psi| - |\psi^\perp\rangle\langle\psi^\perp|$$

$$|\psi\rangle = \begin{pmatrix} \cos\left(\frac{\theta}{2} \cdot \frac{1}{2} - \frac{\pi}{4}\right) \\ \sin\left(\frac{\theta}{2} \cdot \frac{1}{2} - \frac{\pi}{4}\right) \end{pmatrix}$$

$$|\psi^\perp\rangle = \begin{pmatrix} \cos\left(\frac{\theta}{2} \cdot \frac{1}{2} + \frac{\pi}{4}\right) \\ \sin\left(\frac{\theta}{2} \cdot \frac{1}{2} + \frac{\pi}{4}\right) \end{pmatrix}$$

$$\theta = \pi/2.$$



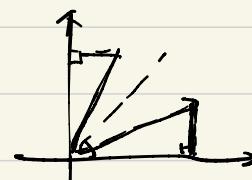
$$|\psi\rangle = \begin{pmatrix} \cos\frac{\pi}{8} \\ \sin\frac{\pi}{8} \end{pmatrix} \quad |\psi^\perp\rangle = \begin{pmatrix} \cos\frac{3\pi}{8} \\ \sin\frac{3\pi}{8} \end{pmatrix}$$

$$\frac{1}{2}\theta = \sqrt{\left(\frac{\theta}{2} - \frac{\pi}{2}\right)} \frac{1}{2}$$

$$\theta = \frac{\pi}{2} \Rightarrow \phi = -\frac{\pi}{4}$$

$$P_{Y|x=0}(y=0) = \cos^2 \frac{\pi}{8}$$

$$P_{Y|x=0}(y=1) = \cos^2 \frac{3\pi}{8}$$



$$P_{Y|x=1}(y=0) = \sin^2 \frac{\pi}{8} = \cos^2 \frac{3\pi}{8}$$

$$P_{Y|x=1}(y=1) = \sin^2 \frac{3\pi}{8} = \cos^2 \frac{\pi}{8}$$

$$P_{Y|x=2}(y=0) = \cos^2 \frac{3\pi}{8}$$

$$P_{Y|x=2}(y=1) = \sin^2 \frac{3\pi}{8}$$

$$P_{Y|x=3}(y=0) = \sin^2 \frac{\pi}{8} = \cos^2 \frac{3\pi}{8}$$

$$P_{Y|x=3}(y=1) = \sin^2 \frac{3\pi}{8} = \cos^2 \frac{\pi}{8}$$

$$\Rightarrow H(X|Y) = -w_3 \frac{2\pi}{8} \log_2 \frac{w_3 \frac{2\pi}{8}}{2} - w_3 \frac{2\frac{3}{8}\pi}{8} \log_2 \frac{w_3 \frac{2\frac{3}{8}\pi}{8}}{2}$$

$$= 1 - w_3 \frac{2\pi}{8} \log_2 w_3 \frac{2\pi}{8} - w_3 \frac{2\frac{3}{8}\pi}{8} \log_2 w_3 \frac{2\frac{3}{8}\pi}{8}.$$

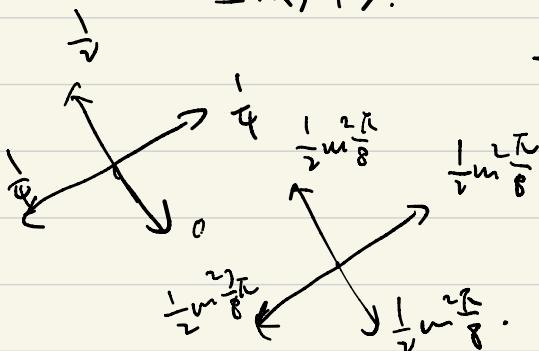
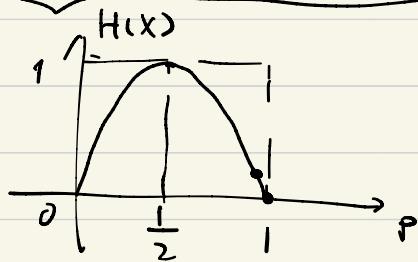
Then:

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

$$= 1 - \underbrace{\left(-w_3 \frac{2\pi}{8} \log_2 w_3 \frac{2\pi}{8} - w_3 \frac{2\frac{3}{8}\pi}{8} \log_2 w_3 \frac{2\frac{3}{8}\pi}{8} \right)}_{H(X)}.$$

$$= 0.39912.$$

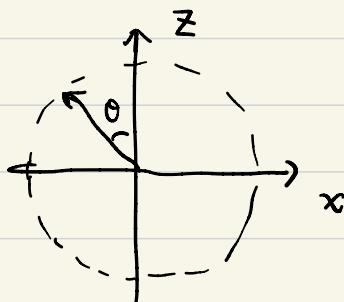
$$< I(X;Y).$$



uncertainty is higher.

2.3

a)



$$H(Y) = 1 \text{ bit.} \\ = \log_2 2. \\ \text{by symmetry.} \\ (\text{need to be checked}).$$

Measure in Z , then:

$$P_{Y|\theta=0} (y=0) = \cos^2(\theta/2)$$

$$P_{Y|\theta=0} (y=1) = \sin^2(\theta/2).$$

$$\text{Then } H(Y|\theta=0) = H_2(\cos^2(\theta/2))$$

$$H(Y|\theta) = \int \frac{d\theta}{2\pi} \cdot (H_2(\cos^2(\theta/2))).$$

$$\Rightarrow I(\theta; Y) = H(Y) - H(Y|\theta)$$

$$H_2(\cos^2(\theta/2)) = -\cos^2(\theta/2) \log_2 \cos^2(\theta/2) \\ - \sin^2(\theta/2) \log_2 \sin^2(\theta/2)$$

$$H(Y|\theta) = -2 \int_0^\pi \cos^2(\theta/2) \log_2 \cos^2(\theta/2).$$

$$\Rightarrow I(\theta; Y) = 1 + 2 \int_0^\pi \cos^2(\theta/2) \log_2 \cos^2(\theta/2).$$

$$b) E_k = \frac{2}{n} |\psi_k\rangle \langle \psi_k| \quad , \quad \phi_k = 2\pi \frac{k}{n}$$

$$\text{Then } |\psi_k\rangle = \begin{pmatrix} \cos\left(2\pi \frac{k}{n}\right) \\ \sin\left(2\pi \frac{k}{n}\right) \end{pmatrix}.$$

$$|\psi_k\rangle \langle \psi_k| = \begin{pmatrix} \cos^2\left(2\pi \frac{k}{n}\right) & \sin\left(2\pi \frac{k}{n}\right) \cos\left(2\pi \frac{k}{n}\right) \\ \sin\left(2\pi \frac{k}{n}\right) \cos\left(2\pi \frac{k}{n}\right) & \sin^2\left(2\pi \frac{k}{n}\right) \end{pmatrix}$$

$$= \begin{pmatrix} \frac{1}{2}(1 + \cos(4\pi k)) & \frac{1}{2} \sin(4\pi k) \\ \frac{1}{2} \sin(4\pi k) & \frac{1}{2}(1 - \cos(4\pi k)) \end{pmatrix}$$

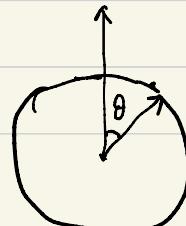
$$\frac{2}{n} \sum_k |\psi_k\rangle \langle \psi_k| \Rightarrow \frac{2}{n} \begin{pmatrix} \frac{n}{2} & 0 \\ 0 & \frac{n}{2} \end{pmatrix} = I.$$

$$\text{Thus } \sum_k E_k = I.$$

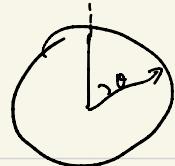
$$I(\theta; Y) = H(Y) - H(Y|\theta).$$

$$P_{Y|Y=a} = \text{tr}(\rho E_a) = \text{tr}(\rho E_b) = \frac{1}{n}.$$

$$\Rightarrow H(Y) = \log_2 n.$$



$$P_{Y|\Theta=\theta} (y=a) = \text{tr}(\rho_\theta |a\rangle \langle a|)$$



$$= \frac{2}{n} \text{tr} (| \psi(\theta) \rangle \langle \psi(\theta) | \phi_k \rangle \langle \phi_k |)$$

$$= \frac{2}{n} | \langle \psi(\theta) | \phi_k \rangle |^2$$

$$\Rightarrow \frac{2}{n} \left| \cos\left(\frac{\theta}{2}\right) \cos\left(\frac{\phi_k}{2}\right) + \sin\left(\frac{\theta}{2}\right) \sin\left(\frac{\phi_k}{2}\right) \right|^2$$

$$\Rightarrow \underbrace{\frac{2}{n} \cdot \cos^2\left(\frac{\theta - \phi_k}{2}\right)}_{= P_{Y|\Theta}(y=k)} = P_{Y|\Theta}(y=k).$$

$$\text{Then } H(Y|\Theta=\theta) = - \sum_k P_{Y|\Theta}(y=k) \log_2 P_{Y|\Theta}(y=k)$$

$$= - \frac{2}{n} \sum_k \cos^2\left(\frac{\theta - \phi_k}{2}\right) \log_2 \cos^2\left(\frac{\theta - \phi_k}{2}\right)$$

$$- \sum_k P_{Y|\Theta}(y=k) \log_2 \frac{2}{n}$$

$$= - \frac{2}{n} \sum_k \cos^2\left(\frac{\theta - \phi_k}{2}\right) \log_2 \cos^2\left(\frac{\theta - \phi_k}{2}\right)$$

$$- \log_2 \frac{2}{n}.$$

Integration :

$$H(Y|\Theta) = - \sum_k \frac{2}{n} \int_0^{\frac{\pi}{2}} \cos^2\left(\frac{\theta - \phi_k}{2}\right) \log_2 \cos^2\left(\frac{\theta - \phi_k}{2}\right)$$

$$- \log_2 \frac{2}{n}$$

$$= - 2 \int_0^{\frac{\pi}{2}} \cos^2\left(\frac{\theta}{2}\right) \log_2 \cos^2\left(\frac{\theta}{2}\right)$$

$$- 1 + \log_2 n.$$

$$\text{Thus } I(\theta; Y) = H(Y) - H(Y|\theta)$$

$$\begin{aligned}
 &= \log_2 n + 2 \int_0^1 m^2 \left(\frac{\theta}{2}\right) \log_2 m^2 \frac{\theta}{2} \\
 &\quad + 1 - \log_2 n \\
 &= 1 + 2 \int_0^1 \cos^2 \left(\frac{\theta}{2}\right) \log_2 \cos^2 \frac{\theta}{2}
 \end{aligned}$$

Same as a).

2.4

a) $|\psi\rangle$ and $|\psi^\perp\rangle$ which is in $(d-1)$ -dimensional space.

$$\text{Prob.} (|\psi^\perp\rangle) = \varepsilon. \quad \text{Prob.} (|\psi\rangle) = 1 - \varepsilon.$$

$$\begin{aligned}
 &\downarrow \\
 \text{Prob.} (|i\rangle) = \varepsilon \cdot p_i = \varepsilon \cdot P(|i\rangle | \psi^\perp \rangle), \quad i = 0, 1, \dots, d-2
 \end{aligned}$$

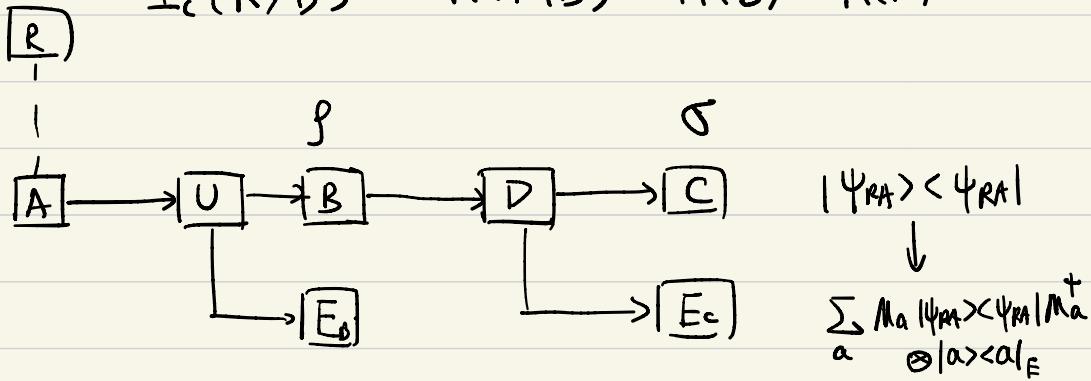
$$\begin{aligned}
 \Rightarrow H(\rho) &\leq H(X) = -(1-\varepsilon) \log_2 (1-\varepsilon) \\
 &\quad - \sum_i \varepsilon p_i \log \varepsilon p_i \\
 &= H_2(\varepsilon) + \varepsilon H(X^\perp)
 \end{aligned}$$

$$\leq H_2(\varepsilon) + \varepsilon \log_2(d-1)$$

uniform distribution
in the $(d-1)$ -dimensional
space.

b)

$$I_c(R>B) = -H(R|B) = H(B) - H(E)$$



$$I_c(R>C)_\sigma \leq I_c(R>B)_p$$

$$\Rightarrow |\Psi_{RB}\rangle = \sum_a M_a |\Psi_{RA}\rangle \otimes |a\rangle_E$$

$$\boxed{H(R)_p - 2H(RC)_\sigma \leq I_c(R>B)_p.}$$

$$I_c(R>C)_\sigma = -H(R|C)_\sigma = H(C)_p - H(RC)_\sigma$$

$$\boxed{I_c(R>B)_p - H(R)_p \geq -2H(RC)_\sigma}$$

$$I_c(R>B)_p = H(B)_p - H(RB)_p$$

$RC E_B E_C$.

$$I_c(R \triangleright B)_p + 2H(RC)_\sigma - H(R)_p \geq 0$$

$$LHS \geq I_c(R \triangleright C)_p + 2H(RC)_\sigma - H(R)_p$$

$$= H(C)_\sigma - H(RC)_\sigma + 2H(RC)_\sigma - H(R)_p$$

$$= H(C)_\sigma - H(RC)_\sigma - H(R)_p$$

$$= \underbrace{H(RE_B E_C)}_\sigma - H(E_B E_C)_\sigma - H(R)_p$$

$$\geq H(R)_\sigma + H(E_B E_C)_\sigma - H(E_B E_C)_\sigma - H(R)_p$$

$$= H(R)_p - H(R)_p$$

$$= 0.$$

□.

c) $\dim(\mathcal{R}L) = d^2$

From a) and b) we complete the proof.

