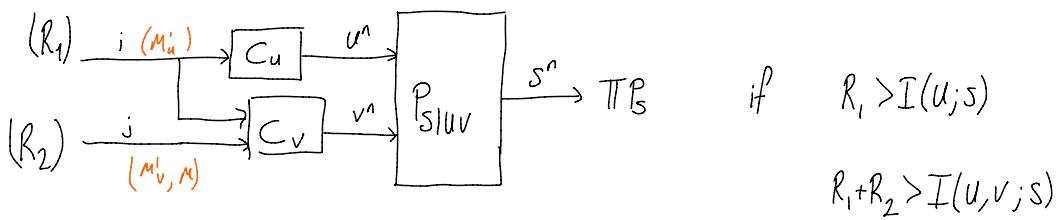


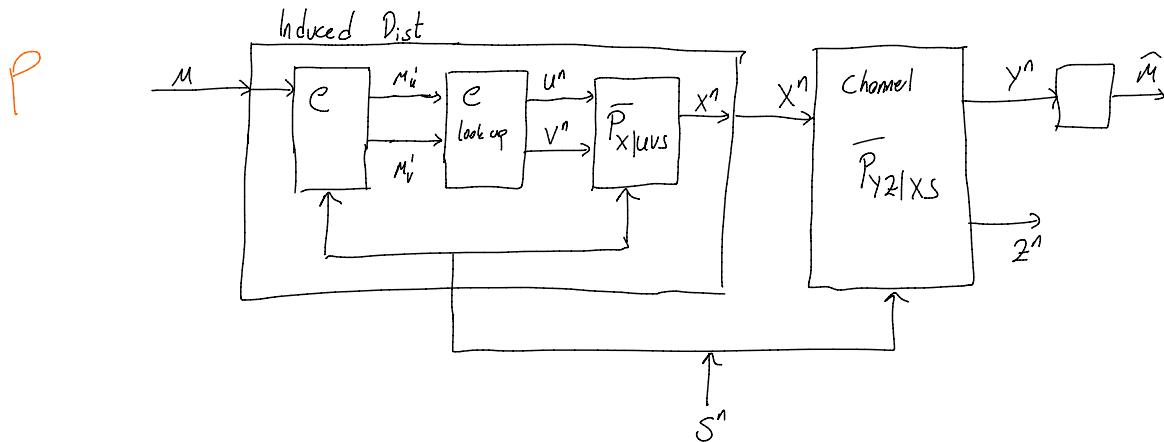
12/06/2016

Tuesday

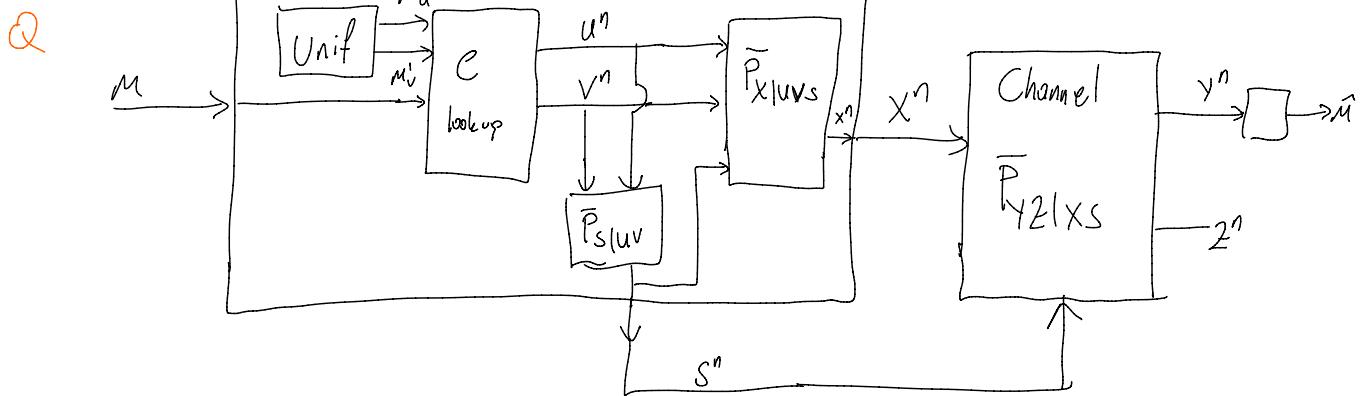
Superposition SCL



G-P WTC



|ideal|:



Difference: $P_{S^n M_u^1 M_v^1 / M} \neq Q_{S^n M_u^1 M_v^1 / M}$

Notice: $P_{S^n / M} = \bar{P}_{S^n}$ by problem definition

$P_{M_u^1 M_v^1 / S^n M}$ is anything we want (will be LE)

By Superposition Soft Covering Lemma:

$$\text{If } R_u' > I(u; s)$$

$$R_u' + R_v' > I(u, v; s)$$

$$\text{then } Q_{S^n/M} \approx \bar{P}_{S^n}$$

see the first picture
today!

$$\text{Define } P_{M_u' M_v' | S^n M} = Q_{M_u' M_v' | S^n M} \propto \bar{P}_{S^n | U^n V^n} (s^n | u^n(m_u'), v^n(m_u', m_v', m))$$

↑

↑ bayes rule

C_V has
3 inputs!

This is LE.

Therefore given rate constraints $P \approx Q$ (the entire systems!)

↳ This is due to scl.

According to Q :

$$\text{Socreg: } \begin{cases} R_u' > I(u; Z) \\ R_u' + R_v' > I(u, v; Z) \end{cases} \Rightarrow Q_{Z^n/M} \approx \bar{P}_{Z^n}$$

for superposition
this is better.

OR

$$R_v' > I(v; Z|u) \Rightarrow Q_{Z^n/M} = 2^{-nR_v'} \sum_{m_u'} P_{Z^n | U^n} (z^n | u^n(m_u'))$$

$$\text{Reliability: } (u^n(m_u'), v^n(m_u', m_v', m), y^n) \in T_\epsilon^{(n)}$$

R_u'

R_v'

R

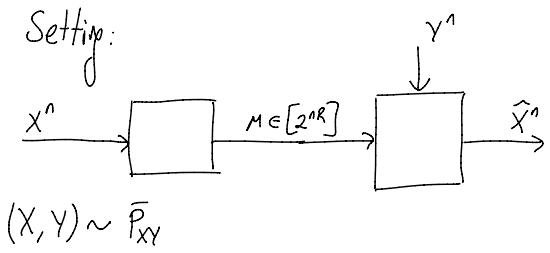
M_u'	M_v'	m	distribution	rate bound
wrong	right		$Q_{UV} Q_Y$	$R_u' < I(u, v; Y)$
right	wrong		$Q_{UV} Q_{Y U}$	$R_v' + R < I(V, Y U)$
wrong	wrong		$Q_{UV} Q_Y$	$R_u' + R_v' + R < I(u, v; Y)$

doesn't matter
because of this!

Wyner-Ziv:

Loosy compression with ^{non-causal} ✓ side information at decoder.

Setting:



Equivalently,

$$P_{X^n Y^n M \hat{X}^n} = \bar{P}_{X^n Y^n} P_{M|X^n} P_{\hat{X}^n|M, Y^n} \Rightarrow M - X^n - Y^n \\ X^n - (M, Y^n) - \hat{X}^n$$

Theorem: (Wyner-Ziv 1976)

$$R(D) = \min_{\substack{P_{u|x} \\ \hat{x} = f(y, u)}} I(u; x) - I(u; y) \\ I(u; x|y) \\ E[d(x, \hat{x})] \leq D$$

$$\left\{ \begin{array}{l} \text{distribution used in theorem} \quad P_{X^n Y^n M \hat{X}^n} = \bar{P}_{X^n Y^n} P_{u|x} P_{\hat{X}^n|yu} \\ \uparrow \text{function of } f \end{array} \right. \implies \begin{array}{l} u - x - y \\ x - (uy) - \hat{x} \end{array}$$

↳ Similar to distribution constraints (on sequence) in problem statement

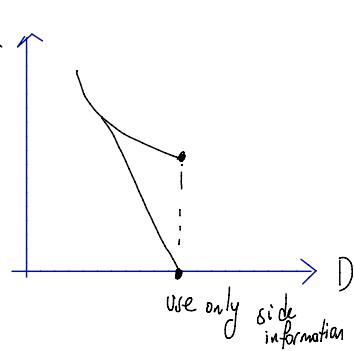
Main Idea of achievability:

Codebook of U sequences to cover X .

Y boosts the effect in two ways:

- 1.) Random binning: The codeword can be only partially transmitted. ($-I(U; Y)$)
- 2.) Decoder uses U and Y to produce \hat{X} .

In binary symmetric:



Proof of achievability: (specific)

$$\text{Fix } \bar{P} \text{ s.t. } R > I_{\bar{P}}(U; X|Y) = I(U; X) - I(U; Y)$$

$$D > E_{\bar{P}}(d(X; \bar{X}))$$

Choose $R' \in (I(U; X) - R, I(U; Y))$

$$\text{Let } m' \in [2^{nR'}]$$

Codebook $C = \{u^n(m, m')\} \sim \bar{P}_U$

Encoder: Observe X^n and produce M (somehow)

Decoder: Receive M . Decode $u^n(n, m')$ (somehow)

Compute, for each i , $\hat{X}_i = f(Y_i, U_i)$

Given by \bar{P} (i.e. $\bar{P}_{\bar{X}|YU}$)

As long as sequences are $\underbrace{\text{J.T.}}_{\text{Jointly typical}} \sim \bar{P}$, distortion satisfied.