

DAYANANDA SAGAR COLLEGE OF ENGINEERING*(An Autonomous Institute Affiliated to VTU, Belagavi)*

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Department of Telecommunication Engineering**Online Continuous Internal Assessment Test - III**Course: **MIMO Technologies**Course Code: **17TE7DCMTN**Semester: **VII - 'A' & 'B'**Date: **05/01/2021**Maximum marks: **50**Duration: **90 Min**

Note: Answer 5 full questions.		Marks
1	<p>a) Turbo codes are ----</p> <p>i) Convolution Code ii) FEC codes iii) Channel Code iv) all of mentioned</p> <p>b) -----, a technique for making forward error correction more robust with respect to burst errors</p> <p>i) Interleaving ii) puncturing iii) equalization iv) source coding</p> <p>c) Trellis termination is an important method for improving performance of ----- by periodically adding tail bits into information sequence.</p> <p>i) Reed Solomon Code ii) BCH Code iii) Hamming Code iv) Turbo Code</p> <p>d) ----- is where the fading process is approximately constant for a number of symbol intervals.</p> <p>i) Block fading ii) Flat fading iii) FS fading iv) Rayleigh Fading</p> <p>e) A channel can be ----- 'block-fading' when it is block fading in both the time and frequency domains.</p> <p>i) octople ii) single iii) quadruple iv) double</p> <p>f) Channel Tap is certain delay on delay line on -----.</p> <p>i) Time Axis ii) Frequency axis iii) Fourier Axis iv) Complex axis</p> <p>g) ----- is the time duration over which the channel impulse response is considered to be not varying</p> <p>i) Channel Time ii) Coherence time iii) Equalization Time iv) Interference Time</p> <p>h) The ----- algorithm is an algorithm for maximum a posteriori decoding of error correcting codes defined on trellises</p> <p>i) BCJR ii) Viterbi iii) MAP iv) Priori</p>	1x10

	<p>i) -----is process of adjusting the spatial attribute of a sound in order to perceive desired 3D sound sensation</p> <p>i) Spatial Equalization ii) Temporal Equalization iii) ISI iv) ISI-Tap</p> <p>j) -----is transceiver architecture for offering spatial multiplexing over multiple-antenna wireless communication systems</p> <p>i) D Blast ii) BLAST iii) V Blast iv) K-Blast</p>	
2	Write about SOVA Decoder in Concatenated STBC.	10
3	Verify the Frequency Selective Frequency Channel Information Rates with Gaussian Inputs.	10
4	Elaborate APP Decoder for Concatenated STBC	10
	(OR)	
5	Evaluate Full Diversity Code for MIMO FS Channels.	10
6	Demonstrate Detection Algorithms for Spatial Multiplexing Systems for Threaded STC.	10
	(OR)	
7	Verify Diversity/Multiplexing Gain Trade-off with plots and examples.	10

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No. of copies: 85

Solution Scheme

Subject: MIMO Technologies

Date: 05/01/2021

Course Code: 17TE7DCMTN

Sem: 7th 'A' & 'B'

- Q1. (a) All of the mentioned — (iv)
- (b) Interleaving — (i)
- (c) Turbo Code — (iv)
- (d) Block fading — (i)
- (e) double — (iv)
- (f) Time axis — (i)
- (g) Coherence time — (ii)
- (h) BCJR — (i)
- (i) Spatial Equalization — (i)
- (j) BLAST — (ii)

② SOVA Decoder in Concatenated STBC

Viterbi algorithm is a dynamic programming algorithm for finding the most likely sequence of hidden states — called the viterbi path — that results in a sequence of observed events, especially in the context of Markov

in the context of Markov information sources and hidden Markov models

Markov Model

In probability Model, a Markov model is a stochastic model used to model randomly changing systems. It is assumed that future states depend only on the current state, not on the events that occurred before.

Markov models can be used to recognise patterns, make predictions and to learn the statistics of sequential data. Markov model is a state machine with the state changes being probabilities. In hidden Markov model, we don't know the probabilities but we know outcomes.

HMM (Hidden Markov Models) in general both supervised and unsupervised are heavily applied to model sequences of data. HMM are a class of probabilistic graphical model that allows us to predict a sequence of unknown (hidden) variables from a set of observed variables.
Eg predicting weather (hidden) based on type of cloths someone wears.

Markov chains - is a discrete-time process for which the future behaviour given the past and present, only depends on the present and not on the past.

A Markov process is continuous-time version of a Markov chain.

SOVA is alternative to APP.

complexity is $\frac{1}{2}$ that of APP. at the expense of degradation of performance.

Cumulative metric $m_k(s)$ for state s at time k is updated according to

$$m_s(k) = \max \{ \lambda(s', s) + m_{k-1}(s'), \lambda(s'', s) + m_{k-1}(s'') \}$$

$\lambda(s', s) \rightarrow$ branch metric of transition from state s' to s

In iterative decoding, branch metric

when incorporating the a priori information

$$\lambda(s', s) = \frac{1}{2} L^e(b_k) b_k + 2\sqrt{P} y_k^b b_k + 2\sqrt{P} y_k^p P_k$$

$L^e(b_k) \rightarrow$ extrinsic information could be $L_{21}^e(b_k)$ or $L_{12}^e(b_k)$

$\Delta_k \rightarrow$ difference metric for state s at time k

$$\Delta_k = |(m_{k-1}(s') + \lambda(s', s)) - (m_{k-1}(s'') + \lambda(s'', s))|$$

$$\Delta_k \approx \log \frac{P(\text{correct})}{1 - P(\text{correct})}$$

$P(\text{correct}) \rightarrow$ probability of path decision of survivor at time k was correct
 $\therefore \Delta_k \rightarrow$ reliability that the path ending at state s at time k was correct.

$$\Delta_k^* = \min \{ \Delta_k, \Delta_{k+1}, \dots, \Delta_{k+\delta} \}$$

$\delta \rightarrow$ delay \rightarrow decoding depth.
 \rightarrow min is taken only over the non-surviving paths within the time window $[k, k+\delta]$ that would lead to different decision \hat{b}_k .

$L(b_k) \approx \hat{b}_k \cdot \Delta_k^* \leftarrow$
 Given \hat{b}_k and Δ_k^* the soft output for bit b_k is approximated by

Q3. Frequency Selective Channel Information Rates with Gaussian Inputs

System with $N_t \rightarrow$ transmit antennas
 $N_r \rightarrow$ Receiving antennas

channel has L Taps. ISI taps.

at j^{th} receive antenna at k^{th} tap is

given by

$$y_j(k) = \sqrt{P} \sum_{l=0}^{L-1} \sum_{i=0}^{N_t} h_{ij}^{(l)} x_i(k-l) + n_j(k)$$

$$I_{\text{Gauss}} = \int_0^{\infty} \log \left(\det \left(I_{N_r} + \frac{P}{N_t} \tilde{H}_f^T \tilde{H}_f \right) \right) df$$

$$\tilde{H}_f = \sum_{l=0}^{L-1} H_l \exp(-j2\pi fl)$$

$$H_l = \begin{bmatrix} h_{11}^{(l)} & h_{12}^{(l)} & \dots & h_{1N_r}^{(l)} \\ h_{21}^{(l)} & h_{22}^{(l)} & \dots & h_{2N_r}^{(l)} \\ \vdots & & & \vdots \\ h_{N_t 1}^{(l)} & \dots & \dots & h_{N_t N_r}^{(l)} \end{bmatrix}$$

where the time dependence of the fading coefficients is dropped since they are constants.

$$I = \lim_{N \rightarrow \infty} \frac{1}{N} I(X(1), X(2), \dots, X(N); Y(1), Y(2), \dots, Y(N))$$

$X(1), X(2), \dots, X(N) \rightarrow$ Sequence of channel inputs

$Y(1), Y(2), \dots, Y(N) \rightarrow$ Set of channel outputs

$$\frac{1}{N} H(Y(1), \dots, Y(N)) = -\frac{1}{N} E[\log(P(Y(1), \dots, Y(N)))]$$

$$y(k) = [y_1(k), y_2(k), \dots, y_{N_r}(k)]^T$$

$$P(y(1), y(2), \dots, y(N)) = \sum_{m=0}^{M-1} \alpha_N(m)$$

Q4 APP decoder for concatenated STBC.

APP Algorithm

$1/2$ RSC code

- (N) $b = [b_1, b_2, \dots, b_N]$ Input Sequence.
- (N) $P = [P_1, P_2, \dots, P_N]$ parity sequence
- (2N) $x = [x_1^b, x_1^p, \dots, x_N^b, x_N^p]$ codeword sequence multiplexed.
- (2N) $y = [y_1^b, y_1^p, \dots, y_N^b, y_N^p]$ received sequence
- (2N) $n = [n_1^b, n_1^p, \dots, n_N^b, n_N^p]$ noise sequence

decision $\rightarrow \hat{b}_k$ in logarithmic domain

Log posteriori probability $= L(b_k)$

$$L(b_k) \triangleq \log \left[\frac{P(b_k = +1 | y)}{P(b_k = -1 | y)} \right] \text{ for BPSK}$$

Forward recursion

$$\alpha_k(s) \triangleq P(s_k = s, y_1^k) \rightarrow (b_1^k, p_1^k)$$

Path branch metric

$$\gamma_k(s', s) \triangleq P(s_k = s, y_k | s_{k-1} = s')$$

Backward recursion

$$\tilde{\beta}_k(s) \triangleq P(y_{k+1}^N | s_k = s) \rightarrow (b_k, p_k)$$

$$(b_{k+1}^N, p_{k+1}^N)$$

$$\tilde{\alpha}_k(s) = \sum_{s'} \gamma_k(s', s) \cdot \tilde{\alpha}_{k-1}(s')$$

valid state transition $s' \rightarrow s$

$$\tilde{\gamma}_k(s', s) = P(b_k) p(y_k | b_k)$$

$$= \begin{cases} \frac{P(b_k)}{\pi} \exp\left(-\left(y_k - \sqrt{P} b_k\right)^2 - \left(y_k - \sqrt{P} P_k\right)^2\right) & \text{for valid } s' \rightarrow s \\ 0 & \text{for invalid } s' \rightarrow s \end{cases}$$

$$\tilde{\beta}_{k+1}(s') = \sum_s \tilde{\beta}_k(s) \cdot \tilde{\gamma}_k(s', s)$$

Once $\tilde{\alpha}_k(s)$, $\tilde{\gamma}_k(s', s)$ and $\tilde{\beta}_k(s)$ are computed one can easily compute

$$L(b_k) = \frac{\sum_{T+} p(s_{k-1}=s' s_k=s, y)}{\sum_{T-} p(s_{k-1}=s' s_k=s, y)}$$

$$= \frac{\sum_{T+} \tilde{\alpha}_{k-1}(s') \cdot \tilde{\gamma}_k(s', s) \tilde{\beta}_k(s)}{\sum_{T-} \tilde{\alpha}_{k-1}(s') \cdot \tilde{\gamma}_k(s', s) \tilde{\beta}_k(s)}$$

⑤ Full Diversity Code For MIMO F S Channels

Space time codeword matrix for flat fading

$x(1)$	0	0	...	0	0
$x(2)$	$x(1)$	0	...	0	0
$x(3)$	$x(2)$	$x(1)$...	0	0
\vdots					
$x(N_E)$	$x(N_E-1)$	$x(N_E-2)$		$x(1)$	
$x(N_E+1)$	$x(N_E)$	$x(N_E-1)$...	$x(2)$	
$x(N)$	$x(N-1)$...	$x(N-N_E)$		

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Flat fading

N_t transmit antenna system

uses delay diversity code is obtained

simply delayed versions of sequence

For FS Fading in Flat Fading

~~By~~ ^{by} delaying by single symbol for each transmit antenna will not work because some columns of

ST Codeword matrix are identical. Hence

code will not be full rank.

If we delay the transmitted symbols of 1st antenna by a number of ISI tap

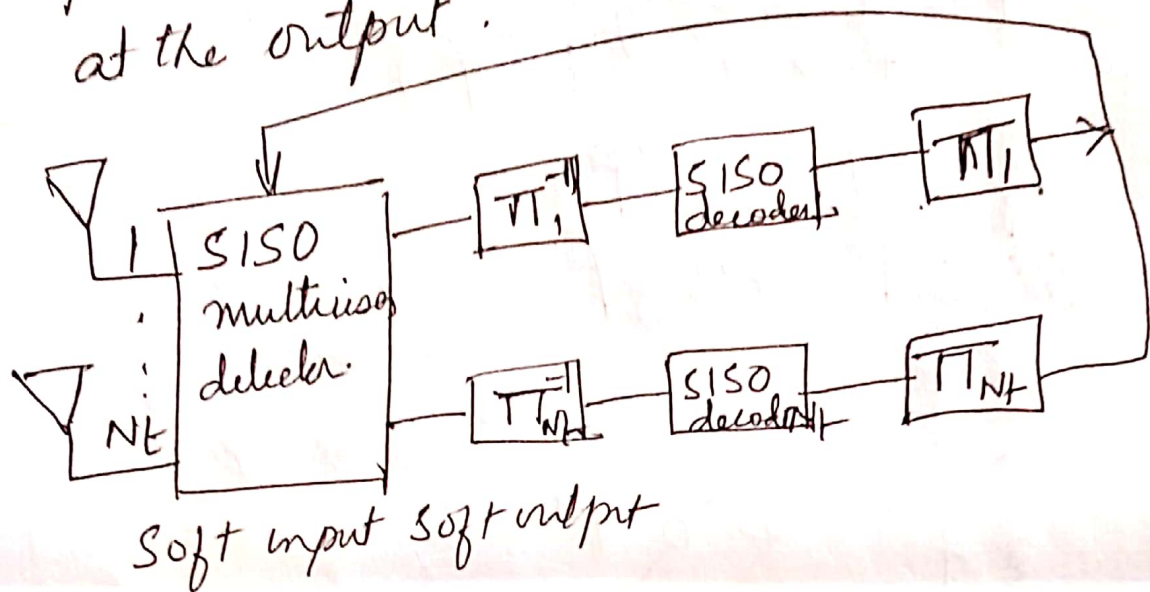
L instead of a single symbol we will achieve Full diversity code of order $N_t N_r L$

$$\begin{bmatrix} x(1) & 0 & 0 & 0 & \dots & 0 \\ x(2) & x(1) & 0 & 0 & \dots & 0 \\ x(3) & x(2) & x(1) & \dots & \dots & 0 \\ x(N_t L) & x(N_t L - 1) & x(N_t L - 2) & \dots & x(1) \\ x(N_t L + 1) & x(N_t L) & x(N_t L - 1) & \dots & x(2) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x(N) & x(N - 1) & x(N - 2) & \dots & x(N - N_t L + 1) \end{bmatrix}$$

Space time code word matrix for Frequency Selective Fading.

Q6. Detection Algorithm for Spatial Multiplexing for Threaded STC.

A Thread is a layer that extends over full spatial span N_t and full temporal span N , where $N \rightarrow$ codewords length at the output.



Explanation

① Greedy Algorithm (log likelihood in greedy fashion)

② Belief Propagation

③ Turbo-Blast Detection

④ Reduced complexity ZF/MMSE Detection
(W_{opt}^H is computed for every detected layer)

⑤ SPACE decoding.

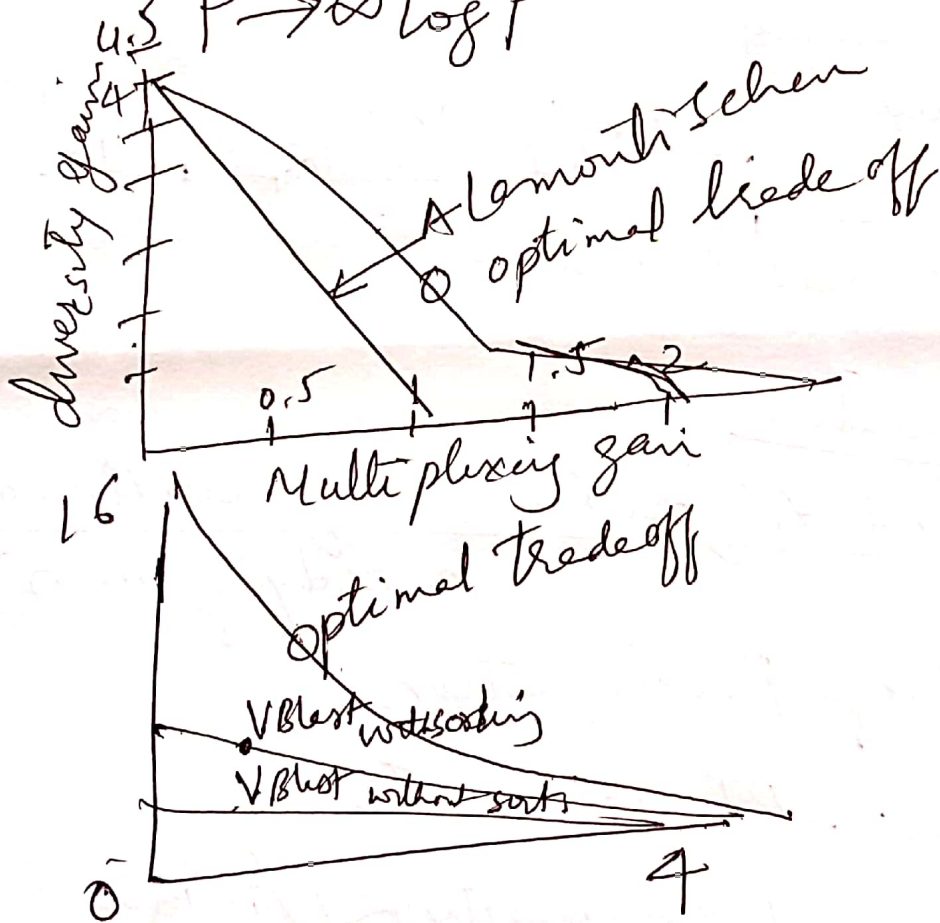
Q7. Diversity/Multiplexing Gain Tradeoff.

(d) Diversity gain \rightarrow negative asymptotic slope of error rate curve as function of SNR in log-log scale

Multiplexing gain (γ)

$$d = \lim_{P \rightarrow \infty} \frac{-\log P_e(P)}{\log(P)}$$

$$\gamma = \lim_{P \rightarrow \infty} \frac{R(P)}{\log P}$$



Explanation