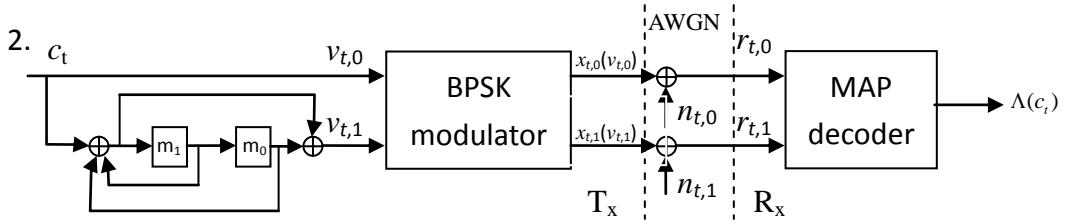


## 2008 Advanced Coding Theorem HW#1 Due date :2007.03.26

1. The same as the example in Lecture 2 except  $\frac{E_s}{N_0} = \frac{1}{2}$  and

$$\mathbf{r} = (1.4, -1.6, -0.9, -1.1, -0.3, 1.1, -0.3, 0.6).$$

- (a) Perform SOVA decoding.
- (b) Perform Max-log-MAP decoding.
- (c) Perform Log-MAP decoding.



- (a) Plot the trellis diagram.
- (b)  $\alpha_{t-1}(l) = ?$  for log-MAP and  $l = 0, 1, 2, 3$ . ( $l \equiv m_1 \cdot 2 + m_0$ )
- (c)  $\beta_t(l) = ?$  for  $l = 0, 1, 2, 3$ .
- (d)  $L(c_t) = ?$

	i=t-1	i=t	i=t+1
$\exp\left(-\frac{ r_{i,0} - x_{i,0}(0) ^2}{2\sigma^2}\right)$	0.1	0.25	0.4
$\exp\left(-\frac{ r_{i,0} - x_{i,0}(1) ^2}{2\sigma^2}\right)$	0.5	0.2	0.25
$\exp\left(-\frac{ r_{i,1} - x_{i,1}(0) ^2}{2\sigma^2}\right)$	0.4	0.5	0.1
$\exp\left(-\frac{ r_{i,1} - x_{i,1}(1) ^2}{2\sigma^2}\right)$	0.25	0.3	0.5
a priori probability $P_i(c_t=0)$	0.5	0.25	0.6

$\alpha_{t-2}(0) = 0.5$	$\beta_{t+1}(0) = 0.1$
$\alpha_{t-2}(1) = 0.2$	$\beta_{t+1}(1) = 0.5$
$\alpha_{t-2}(2) = 0.25$	$\beta_{t+1}(2) = 0.25$
$\alpha_{t-2}(3) = 0.4$	$\beta_{t+1}(3) = 0.2$