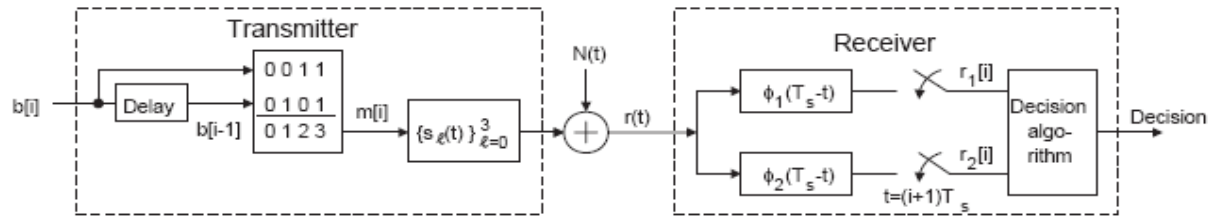
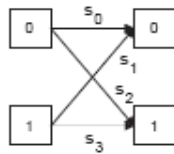


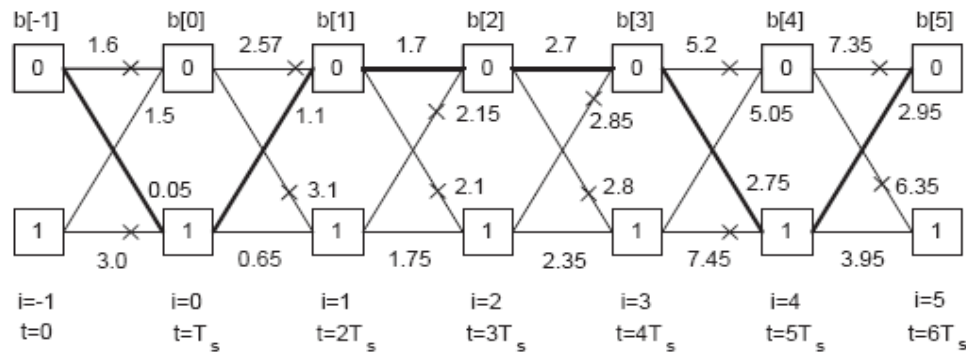
EXAMPLE 8.16



The table below gives the squared Euclidean distance increments $D_{inc}^2[i] = (r_1[i] - s_{j,1})^2 + (r_2[i] - s_{j,2})^2$, obtained between $\mathbf{r}[i]$ and s_j , $j = 0, 1, 2, 3$. The noise $N(t)$ is AWGN.



$i :$	0	1	2	3	4	5
s_0	1.6	1.07	0.6	1.0	2.5	2.3
s_1	1.5	1.05	1.5	1.1	2.7	0.2
s_2	0.05	1.6	1.0	1.1	0.05	1.3
s_3	3.0	0.6	1.1	0.6	5.1	1.2



$$r(t) = z(t) + N(t) = \sum_{n=-\infty}^{\infty} x_{m[n]}(t - nT_s) + N(t), \quad -\infty \leq t \leq \infty \quad (8.54)$$

$$x_\ell(t) = s_\ell(t) * h(t), \quad \ell = 0, 1, \dots, M_{tra} - 1 \quad (8.55)$$

$$\int_{-\infty}^{\infty} (r(t) - z(t))^2 dt \quad (8.56)$$

maximize the expression,

$$\int_{-\infty}^{\infty} \left(r(t)z(t) - \frac{z^2(t)}{2} \right) dt \quad (8.57)$$

$$\begin{aligned}
& \int_{-\infty}^{\infty} \left(r(t)z(t) - \frac{z^2(t)}{2} \right) dt = \int_{-\infty}^{\infty} \sum_{n=-\infty}^{\infty} r(t)x_{m[n]}(t - nT_s) dt - \\
& - \frac{1}{2} \int_{-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \sum_{\ell=-\infty}^{\infty} x_{m[n]}(t - nT_s)x_{m[\ell]}(t - \ell T_s) dt = \\
& = \sum_{n=-\infty}^{\infty} \underbrace{\int_{-\infty}^{\infty} r(t)x_{m[n]}(t - nT_s) dt}_{y_{m[n]}[n]} - \\
& - \frac{1}{2} \sum_{n=-\infty}^{\infty} \sum_{\ell=-\infty}^{\infty} \int_{-\infty}^{\infty} x_{m[n]}(t - nT_s)x_{m[\ell]}(t - \ell T_s) dt = \\
& = \sum_{n=-\infty}^{\infty} y_{m[n]}[n] - \frac{1}{2} \sum_{n=-\infty}^{\infty} \sum_{\ell=-\infty}^{\infty} \int_{-\infty}^{\infty} x_{m[n]}(\xi)x_{m[\ell]}(\xi + (n - \ell)T_s) d\xi = \\
& = \sum_{n=-\infty}^{\infty} y_{m[n]}[n] - \frac{1}{2} \sum_{n=-\infty}^{\infty} \sum_{\ell=-\infty}^{\infty} \underbrace{[x_{m[n]}(t) * x_{m[\ell]}(-t)]_{t=(\ell-n)T_s}}_{p_{m[n],m[\ell]}[\ell-n]} \quad (8.58)
\end{aligned}$$

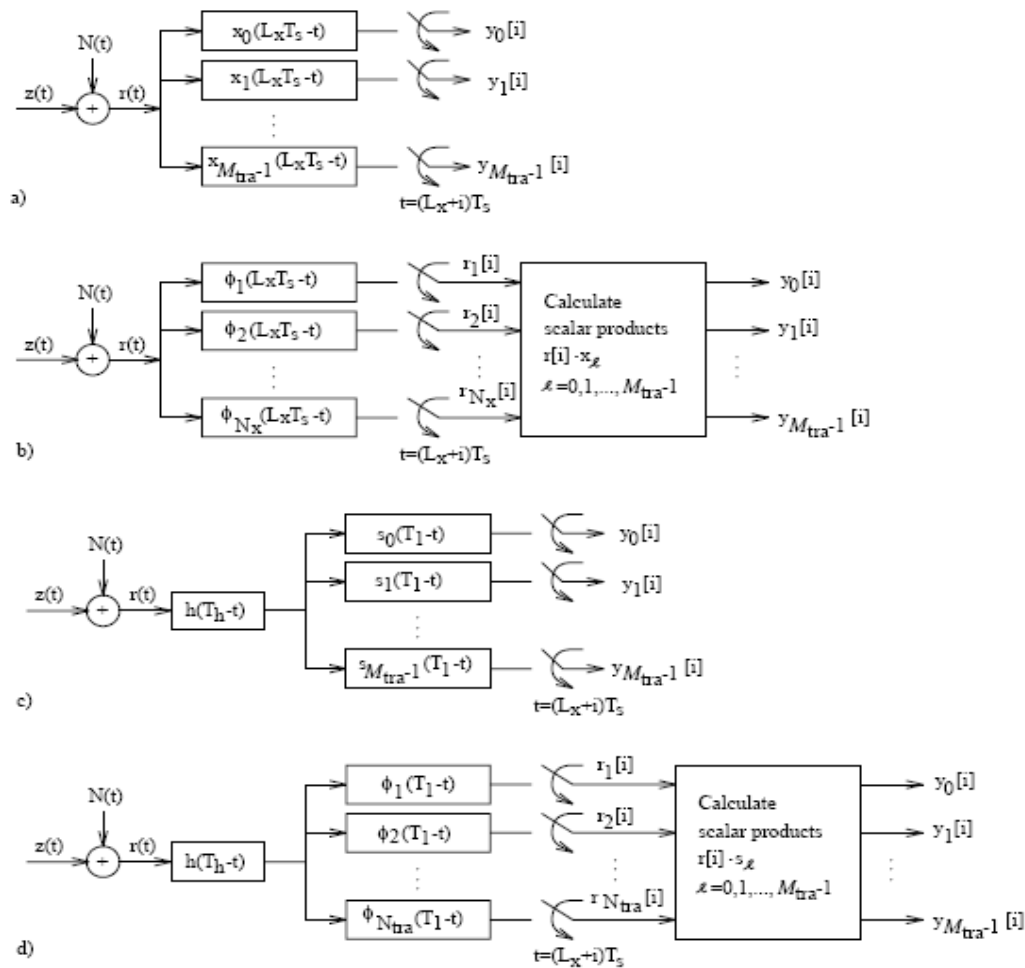


Figure 8.13: The first stage in the ML receiver. a) Filters matched to the signals $x_\ell(t)$; b) Filters matched to the N_x ($N_x \leq M_{tra}$) basis functions of $\{x_\ell(t)\}_{\ell=0}^{M_{tra}-1}$; c) Filters matched to $h(t)$ and to $s_\ell(t)$; d) Filters matched to $h(t)$ and to the N_{tra} basis functions of $\{s_\ell(t)\}_{\ell=0}^{M_{tra}-1}$.

8.4 Bit Error Probability for the ML Receiver in AWGN

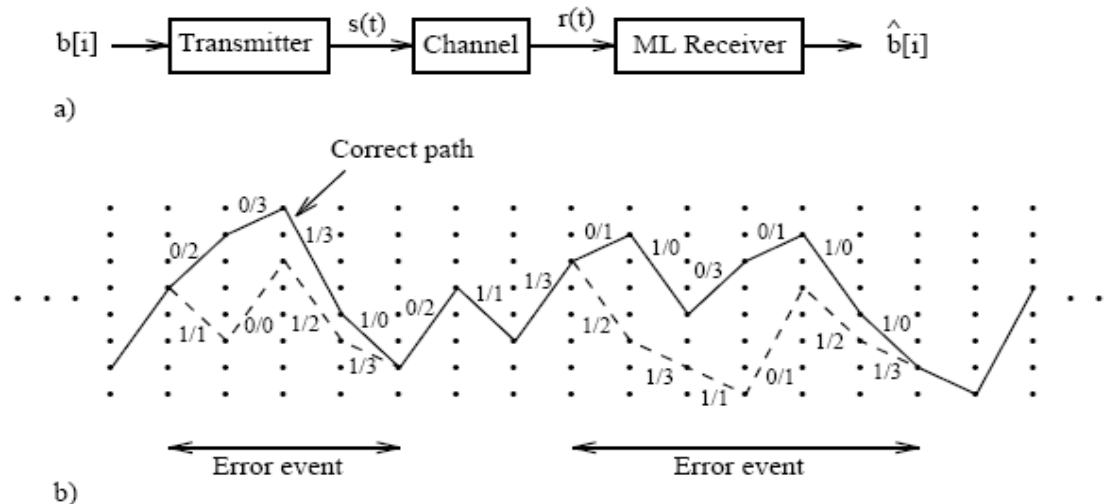


Figure 8.19: a) The digital communication system. b) Parts of the correct path (solid) and the decoded path (dashed) in the trellis.

Observe that by using trellis-coding, longer error events than in the uncoded case are possible. *So, with trellis-coding there is a potential for larger Euclidean distances than in the uncoded case.*

It can be shown that as a first approximation, at high signal-to-noise ratios, the bit error probability can be approximated with the expression

$$P_b \approx cQ \left(\sqrt{\frac{D_{\min}^2}{2N_0}} \right) = cQ \left(\sqrt{d_{\min}^2 \mathcal{E}_b / N_0} \right) \quad (8.102)$$

where D_{\min} is the smallest Euclidean distance *in the set of all error events*, and $d_{\min}^2 = D_{\min}^2 / 2\mathcal{E}_b$.