

Lecture 10

Note Title

2/13/2012

* The VA is an efficient codeword-based ML decoder. 6 different variants.

Q: Can we also design an efficient bit-based ML decoder?

Recall that

$$\begin{aligned} \prod_{t=1}^T f_t(\overline{p[t]}) &= P(\vec{Y}=y \mid \vec{V}=\vec{p}) \\ &= \frac{P(\vec{Y}=u, \vec{V}=\vec{p})}{2^{-\# u \text{ bits}}} \end{aligned}$$

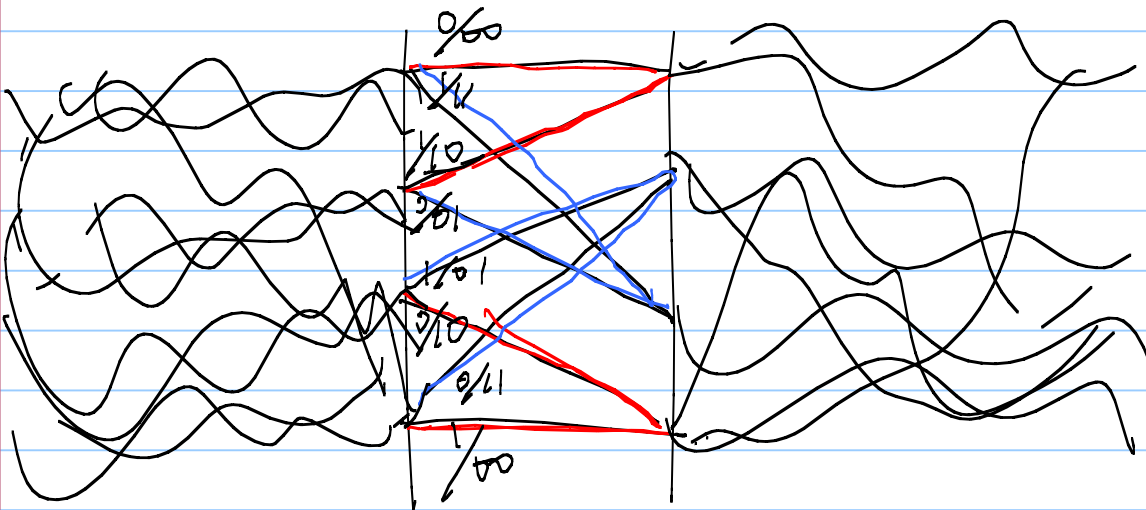
A codeword-based decoder finds

$$\arg \max_{\vec{p}} \prod_{t=1}^T f_t(\overline{p[t]})$$

A bit-based decoder of the i th bit finds

$$\arg \max_{x=0,1} \left(\sum_{\vec{p}:\vec{p}[i]=x} \prod_{t=1}^T f_t(\overline{p[t]}) \right)$$

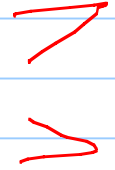
Illustration t
 V_{t-1}, V_t




If we are interested in

$$V_t = 0 \quad \text{vs.} \quad V_t = 1$$

$$\sum_{\text{all paths}} \prod_{t=1}^T P_{t \in \mathcal{E}}(\overline{P^{[t]}})$$

using 

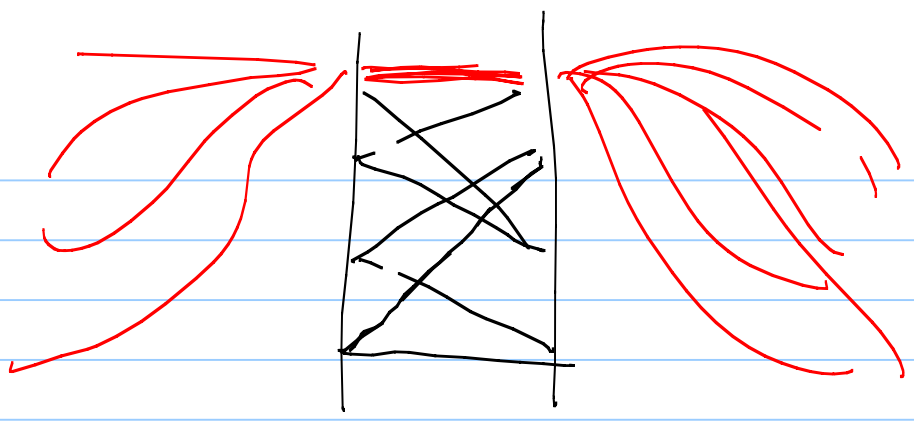
$$\text{vs.} \quad \sum_{\text{all paths}} \prod_{t=1}^T P_{t \in \mathcal{E}}(\overline{P^{[t]}})$$

using 

How to compute the sum efficiently?

We first look at each segment respectively.

Sum of



A generalized form of the distributive rule

$$a_1 b c_1 + a_2 b c_2 + a_1 b c_2 + a_2 b c_1 = (a_1 + a_2) b (c_1 + c_2)$$

=

$$\left(\sum_{s=1}^{t-1} \prod f_s(\overline{p[s]}) \right) \prod_{s=t}^p f_s(\overline{p[s]})$$

=

$$\left(\sum_{s=t+1}^p \prod f_s(\overline{p[s]}) \right)$$

Similarly

=

$$\left(\sum_{s=1}^{t-1} \prod f_s(\overline{p[s]}) \right) \prod_{s=t}^p f_s(\overline{p[s]})$$

=

$$\left(\sum_{s=t+1}^p \prod f_s(\overline{p[s]}) \right)$$

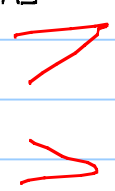
In summary, if we know

$$\left(\sum_{s=1}^{t-1} \prod f_s(\overline{p[s]}) \right), \left(\sum_{s=1}^{t-1} \prod f_s(\overline{p[s]}) \right) \dots$$


$$\& \left(\sum_{s=t+1}^T \prod f_s(\overline{p[s]}) \right) \left(\sum_{s=1}^T \prod f_s(\overline{p[s]}) \right)$$

then we can compute

$$\sum_{\text{all paths}} \prod_{t=1}^T f_t(\overline{p[t]})$$

using 

$$\text{vs. } \sum_{\text{all paths}} \prod_{t=1}^T f_t(\overline{p[t]})$$

using 

efficiently. Q: How to find

The BCJR algorithm 1974

* Again, we rely on iterative computation.

* Iteration must be performed in two ways.

** The forward iteration computes

$$\left(\sum_{s=1}^{t-1} \prod_{s=1}^{t-1} f_s(\overline{p[s]}) \right) \quad , \quad \left(\sum_{s=1}^{t-1} \prod_{s=1}^{t-1} f_s(\overline{p[s]}) \right) \dots$$

The diagram shows two large parentheses containing summation expressions. The first expression is $\sum_{s=1}^{t-1} \prod_{s=1}^{t-1} f_s(\overline{p[s]})$. The second expression is $\sum_{s=1}^{t-1} \prod_{s=1}^{t-1} f_s(\overline{p[s]})$ followed by an ellipsis. In both expressions, arrows point from the summation index $s=1$ to $s=t-1$ to the right, indicating the direction of the iteration.

** The backward iteration computes

$$\left(\sum_{s=t+1}^T \prod_{s=t+1}^T f_s(\overline{p[s]}) \right) \quad \left(\sum_{s=t+1}^T \prod_{s=t+1}^T f_s(\overline{p[s]}) \right)$$

The diagram shows two large parentheses containing summation expressions. The first expression is $\sum_{s=t+1}^T \prod_{s=t+1}^T f_s(\overline{p[s]})$. The second expression is $\sum_{s=t+1}^T \prod_{s=t+1}^T f_s(\overline{p[s]})$. In both expressions, arrows point from the summation index $s=t+1$ to $s=T$ to the right, indicating the direction of the iteration.

Detailed description of the BCJR algorithm

① Given the observation \vec{y}_{obs}
Compute the metric $m_{s,s'}$ for
each path segment from state $s \rightarrow$
state s' . (The same step as in
the VA.)

② The forward iteration. Compute the
forward state metric

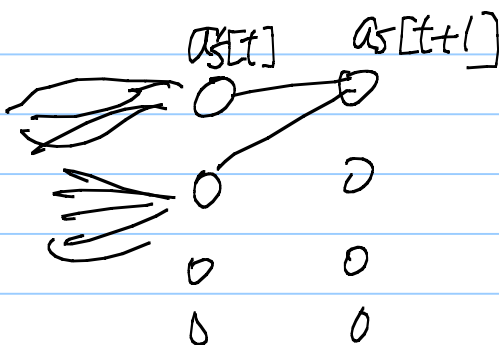
$$a_s[t]$$

Initialization

$$a_{s_0}[0] = 1, \quad a_s[0] = 0 \text{ for all } s \neq s_0$$

$$\forall t = 0, \dots, T-1,$$

$$a_s[t+1] = \sum_{s' \text{ that reaches } s} a_{s'}[t] \cdot m_{s',s}$$



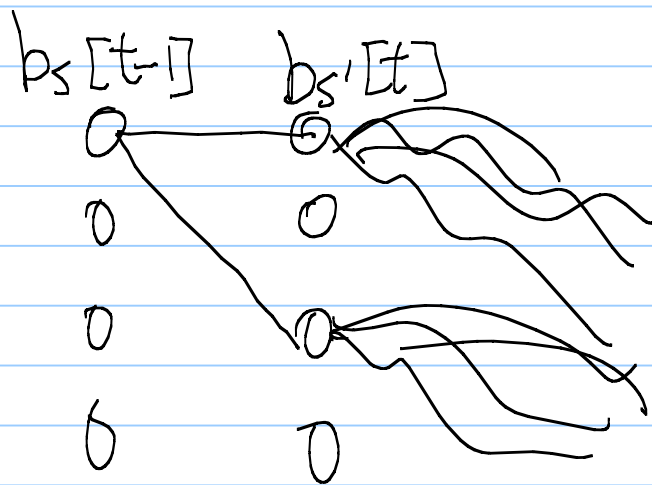
⑤ Backward iteration. Compute $b_s[t]$

Initialization

$$b_s[T] = 1 \text{ for all } s$$

$$\forall t = T, \dots, 1.$$

$$b_s[t-1] = \sum_{\substack{s' \text{ that is} \\ \text{reachable from } s}} b_{s'}[t] m_{s,s'}$$



⑥ Make the decision for the i -th bit by $a_s[t]$ $b_{s'}[t+1]$.

$$\sum_{\substack{P | i = x \\ t=1}} \prod_{t=1}^T P_t(P[t])$$

$$= \sum_{\substack{s, s' \text{ s.t.} \\ s \rightarrow s' | i = x}} a_s[t_0] m_{s,s'} b_{s'}[t_0 + 1]$$

where the i -th bit is in the t -th path segment.

Then we can find

$$\operatorname{argmax}_{x=0,1} \sum_{\overline{P}|_i=x} \prod_{t=1}^T p_t(\overline{P}[t])$$

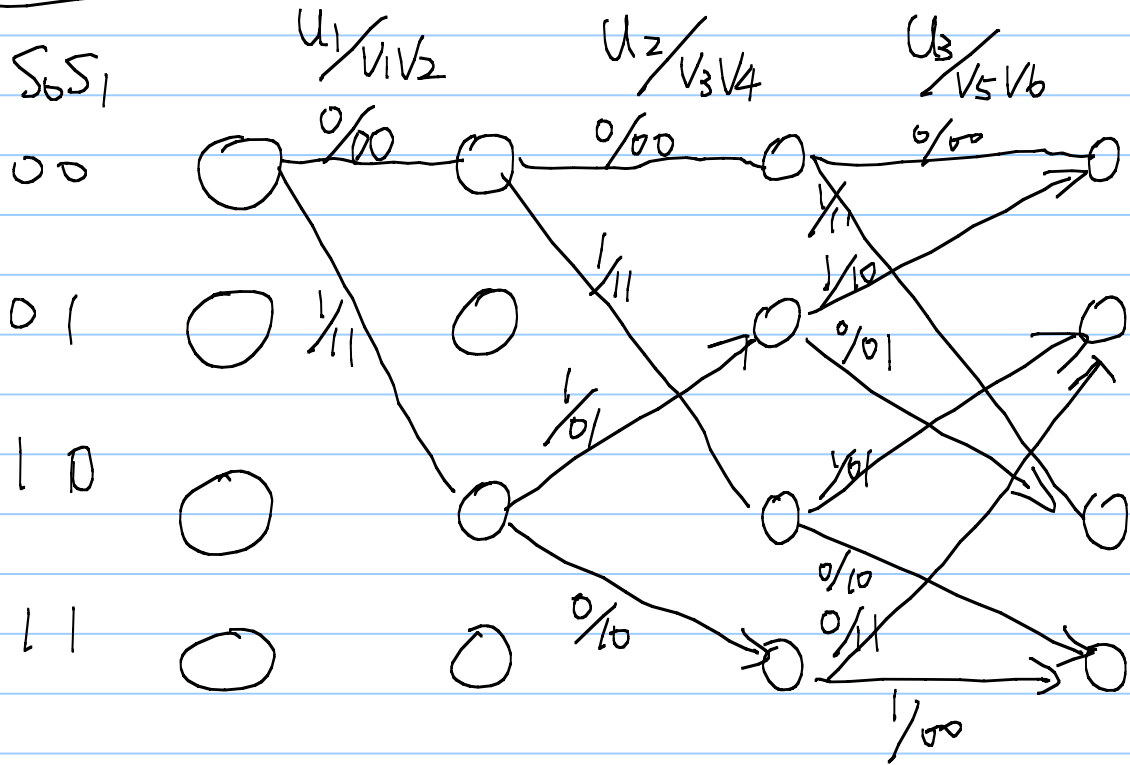
* BCJR finds the optimal bit-level

ML decoder.

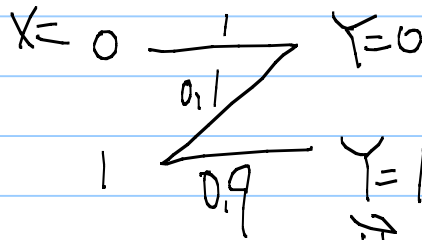
* The BCJR algorithm can be easily modified to be a ML decoder for the info bit u & for the coded bits v by carefully counting the paths in the final, comparison stage.

The principle of the BCJR decoder can be extended to take ^{into} account ^{of} different prior distributions.

An illustrating example With the addition into of the prior distri P_U



Channel: Z-channel



the input into bits

The observation is $\vec{y} = 00$ but suppose we know that u_1, u_2, u_3 are not uniform Bernoulli distributed but have $P(U=1)=0.05$ $P(U=0)=0.95$

Q: Use the BCJR algorithm to find the MAP u_2 value.

Ans: We are interested in maximizing

$$\underset{u_2=0 \text{ or } 1}{\operatorname{argmax}} \sum_{\text{all } \vec{u} \text{ w. the given } u_2 \text{ value}} \frac{P_{\vec{Y}, \vec{U}}(\vec{y}, \vec{u})}{\text{The joint prob.}}$$

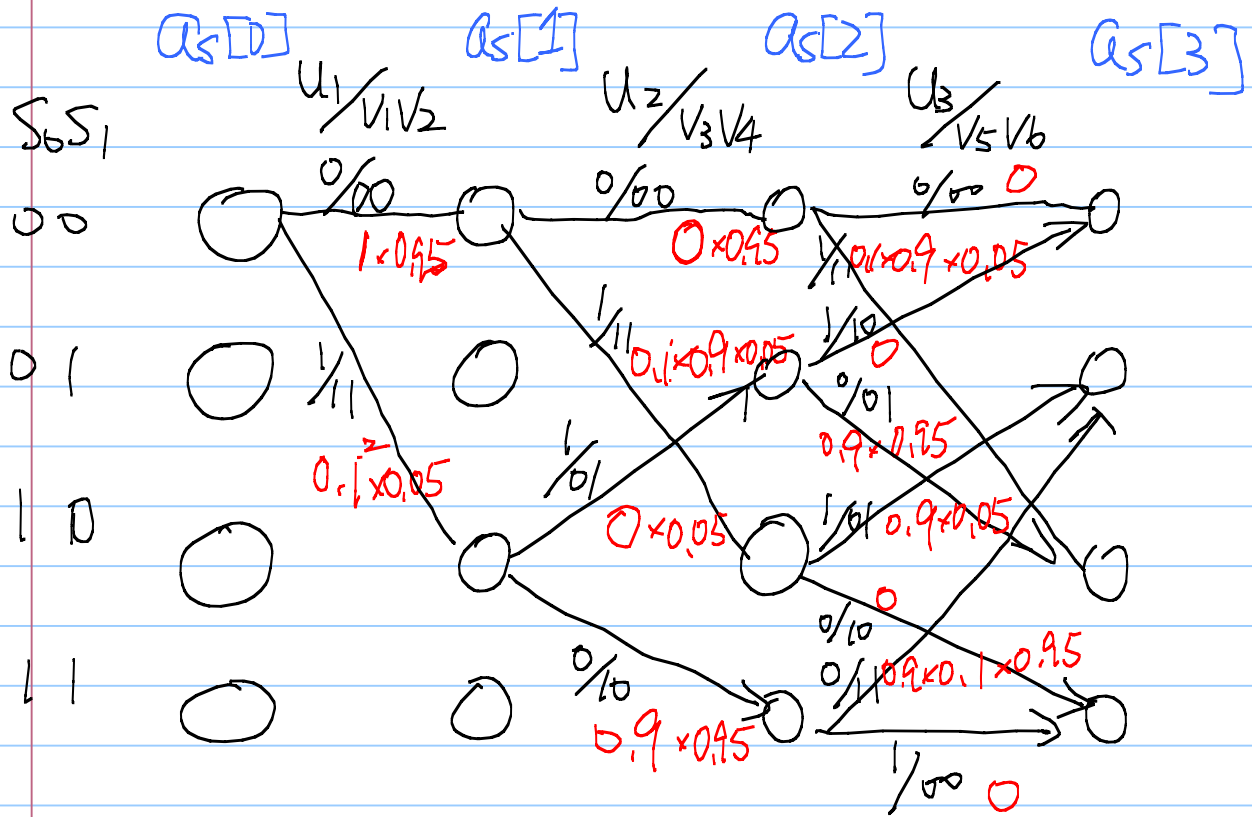
$$\equiv \underset{u_2=0 \text{ or } 1}{\operatorname{argmax}} \sum_{\text{all paths with the given } u_2 \text{ value}} \left(\prod_{t=1}^3 P_{Y_{2t+1} Y_{2t} | V_{2t+1} V_{2t}} \right) \cdot P_{u_1 u_2 u_3}$$

$$\equiv \underset{u_2=0 \text{ or } 1}{\operatorname{argmax}} \sum_{\text{all paths with the given } u_2 \text{ value}} \left(\prod_{t=1}^3 \left(P_{Y_{2t+1} Y_{2t} | V_{2t+1} V_{2t}} \cdot P_{u_t} \right) \right)$$

use $f_t(\cdot)$ instead

$$\equiv \underset{u_2=0 \text{ or } 1}{\operatorname{argmax}} \sum_{\text{all paths using } u_2} \prod_{t=1}^3 f_t(\overline{p[t]})$$

Observation: 00 10 01



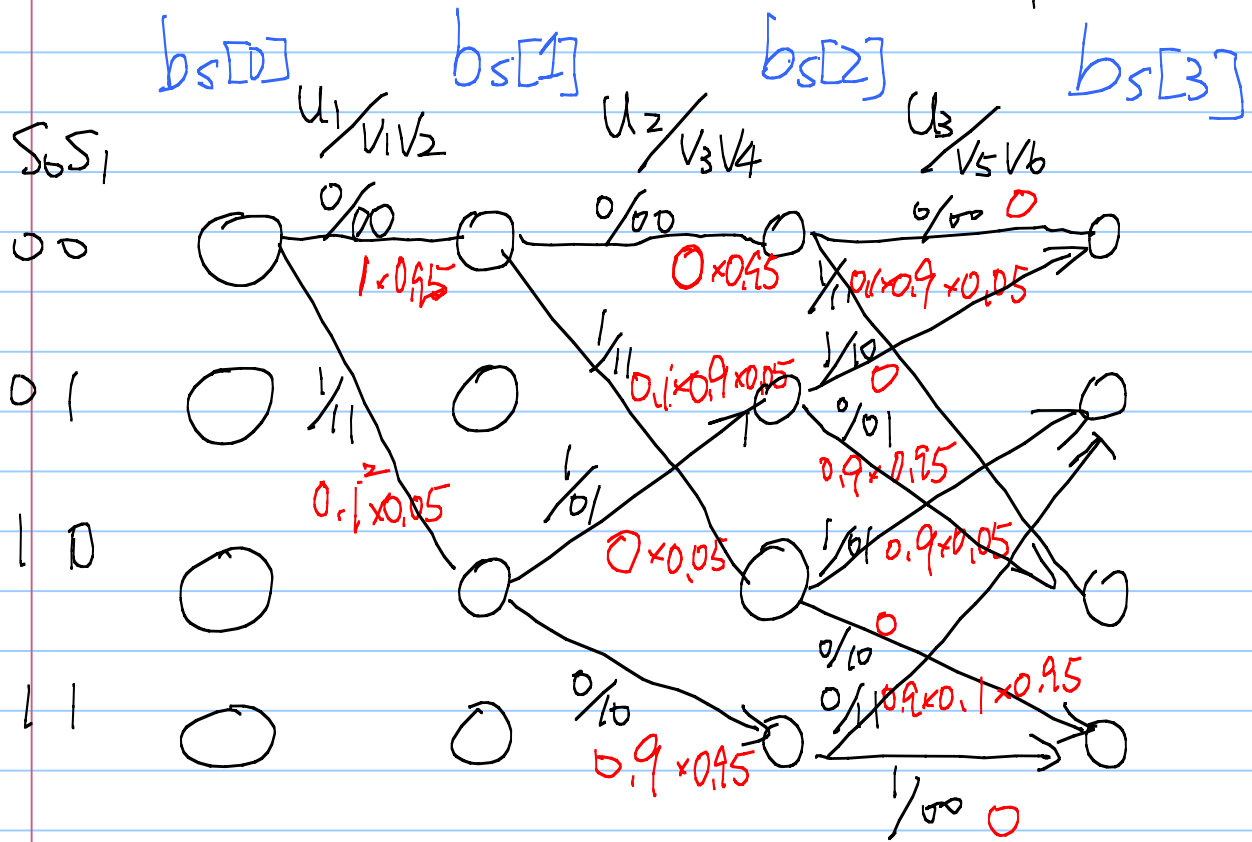
Step 1: Initialize the path metrics

Step 2: Compute the forward metrics,

$$a_s[t+1] = \sum_{s': s' \rightarrow s} a_{s'}[t] \cdot m_{s',s}$$

$a_s[t]$				
1	0.95	0	0	
0	0	0	5.57875×10^{-4}	
0	5×10^{-4}	4.215×10^{-3}	0	
0	0	4.215×10^{-4}	0	

Observation: 00 10 01



Step 3: Compute the backward metrics.

$$b_s[t-1] = \sum_{s'} b_{s'}[t] m_{s,s'}$$

$$b_s[t] \quad \quad \quad b_s[3]$$

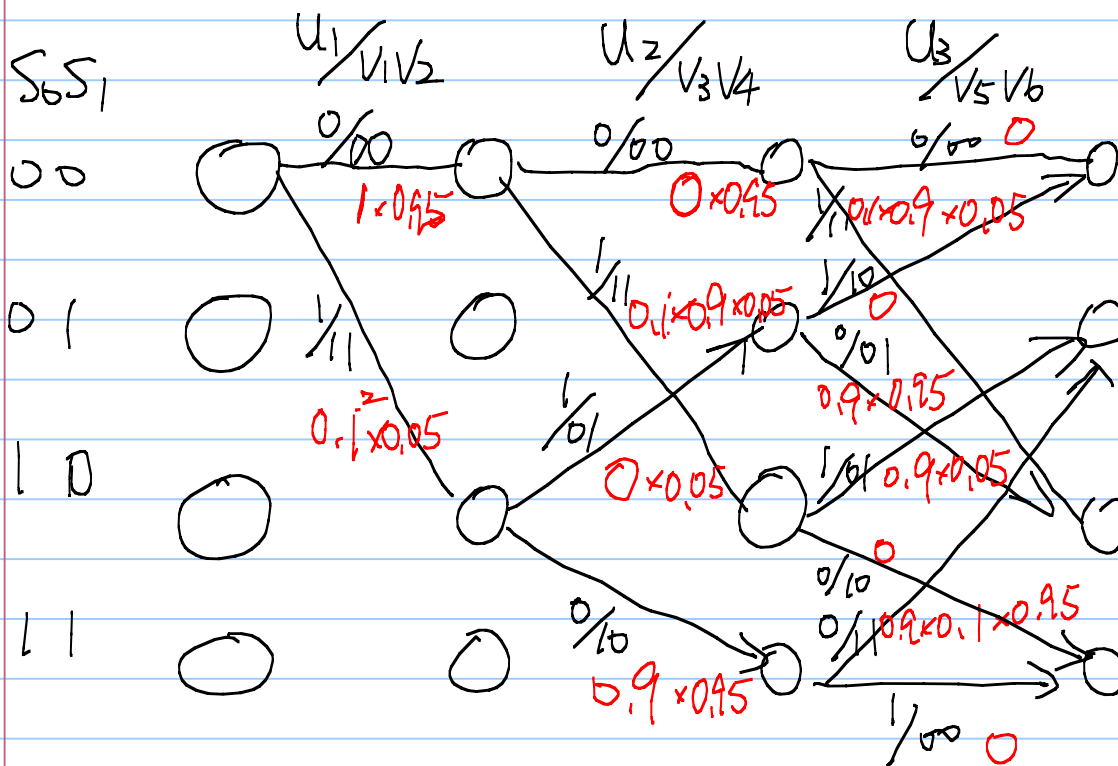
$$2,289,2625 \times 10^{-4} \quad \quad \quad 2,025 \times 10^{-4} \quad \quad \quad 4.5 \times 10^{-3} \quad \quad \quad 1$$

$$0 \quad \quad \quad 0 \quad \quad \quad 8.55 \times 10^{-4} \quad \quad \quad 1$$

$$0 \quad \quad \quad 7,310,25 \times 10^{-2} \quad \quad \quad 4.5 \times 10^{-2} \quad \quad \quad 1$$

$$0 \quad \quad \quad 0 \quad \quad \quad 8.55 \times 10^{-2} \quad \quad \quad 1$$

Observation: 00 10 01



Step 4: Make a decision on U_2

Two cases: $U_2=0$

$U_2=1$

$$a_{00}[1] \cdot M_{00,00} b_{00}[2]$$

$$+ a_{10}[1] \cdot M_{10,11} b_{11}[2]$$

$$= 3.655125 \times 10^{-5}$$

$$a_{00}[1] \cdot M_{00,10} b_{10}[2]$$

$$+ a_{10}[1] \cdot M_{10,01} b_{01}[2]$$

$$= 1.92375 \times 10^{-4}$$

\Rightarrow The MAP for U_2 is 1.

With high posterior prob.

$$\frac{1.92375 \times 10^{-4}}{3.655125 \times 10^{-5} + 1.92375 \times 10^{-4}} \doteq 84\%$$

* Although both being linear with respect to T , the complexity of BCJR \geq VA (needs a lot of multiplication & summation).

* In practice, BCJR is used less frequently as the bit error rate improvement over the VA does not justify the complexity.

* Recently, BCJR is getting more attention due to the invention of the "turbo codes."

* In practice, we simply "approximate" the iteration of the VA & the BCJR.

Common simplification techniques: ① quantized $a_s[t]$ & $b_s[t]$ values

② Incomplete sum / max / product: The best k rule.

③ Proper rescaling. $a'_s[t] \leftarrow \text{const} \cdot a_s[t]$

$b'_s[t] \leftarrow \text{const} \cdot b_s[t]$

since our decision, based on the relative values of $a_s[t]$ ($b_s[t]$) for different s values rescaling does not change the decision

④ Implementation by table look-up.