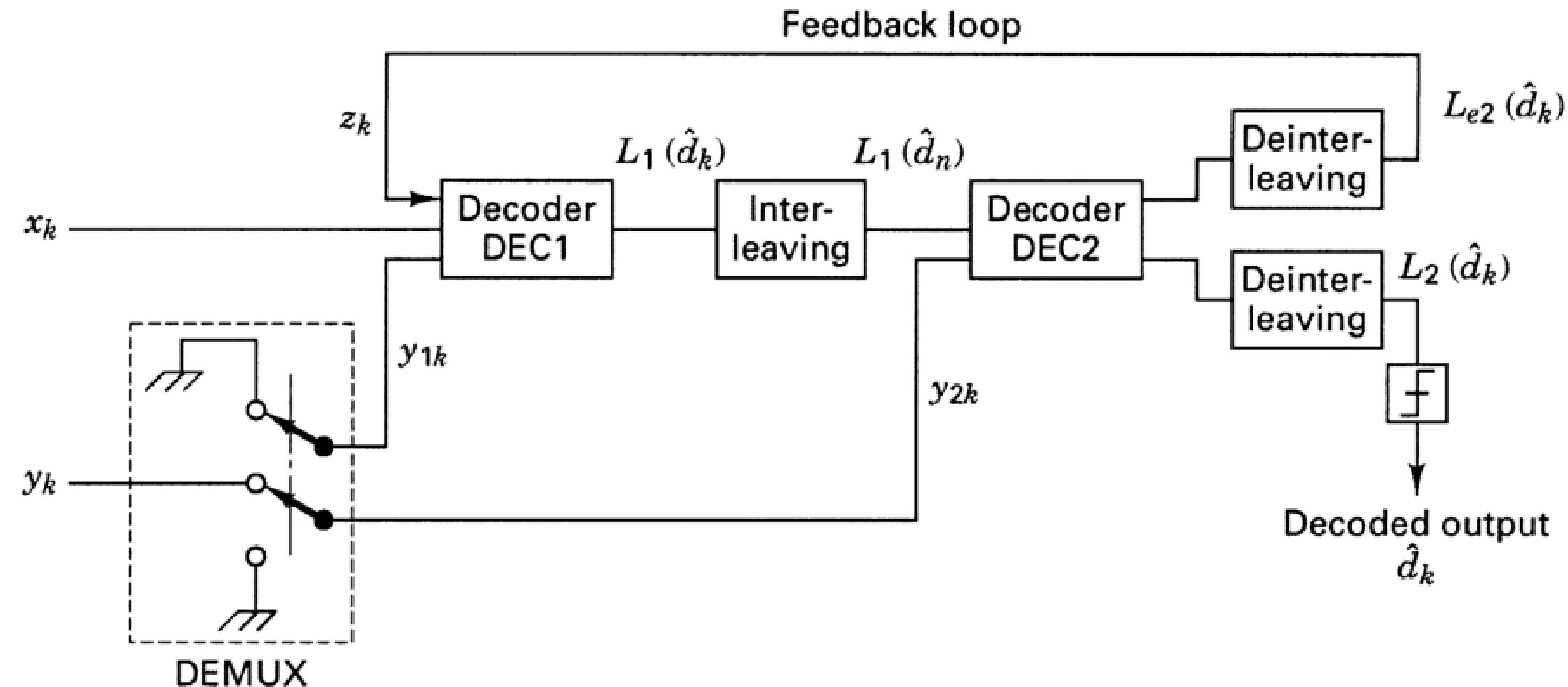


# Turbo Codes : Decoder

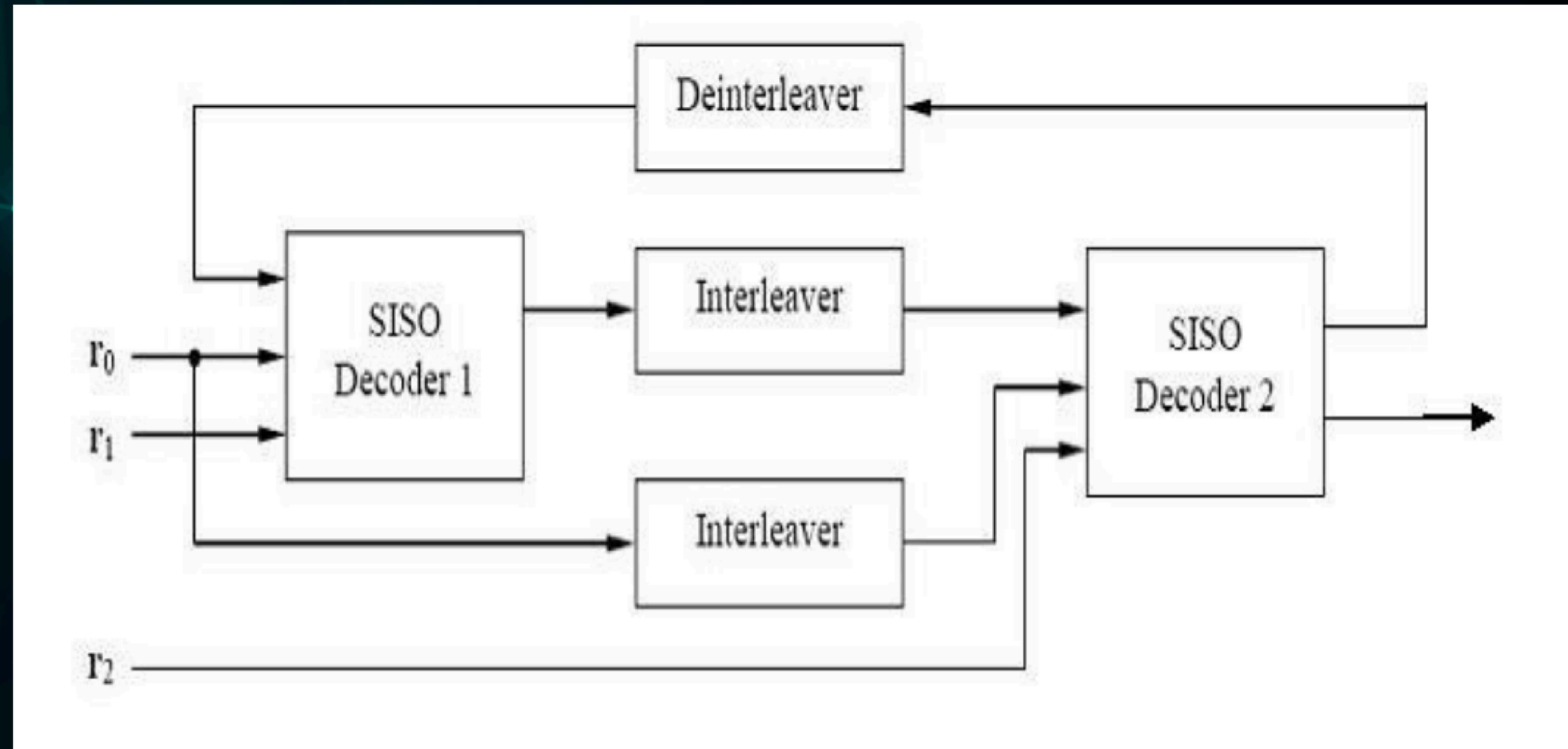
## Block diagram : Decoder With Puncturing



**Figure 8**

Feedback decoder.

*Decoder block diagram without puncturing*



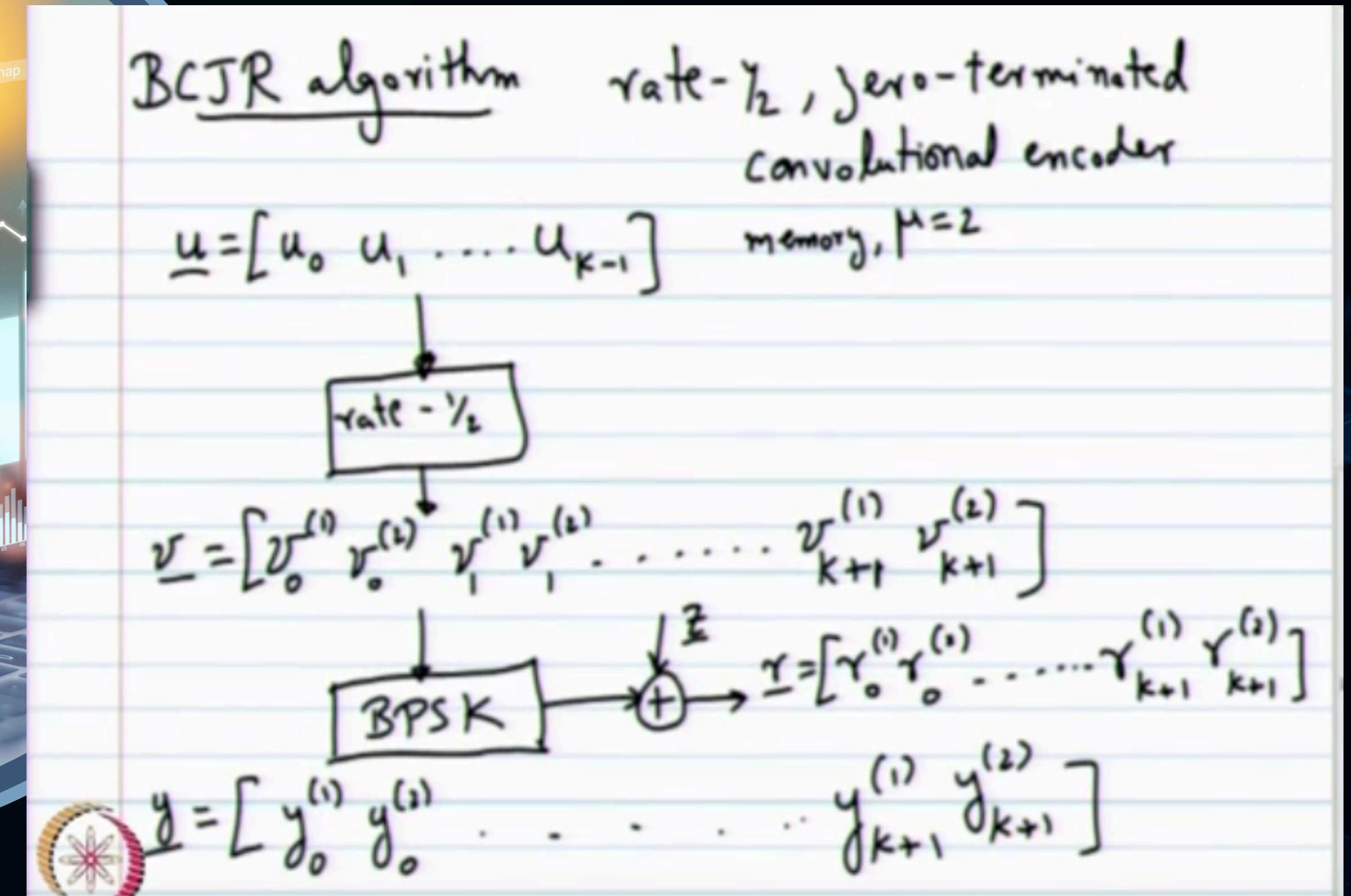
# REASONS FOR TURBO CODE'S AWESOME PERFORMANCE

- weight : no of 1's in a message
- low weight messages doesn't usually give low weight codewords
- size of codewords are pretty large  $> 500$
- turbo decoder has a feedback loop, similar to turbo charger in automobiles, info gets passed from one decoder to the other as soft outputs. i,e, llr's are sent.
- finally, after some iterations/ convergence criterion, we end the loop and take the output

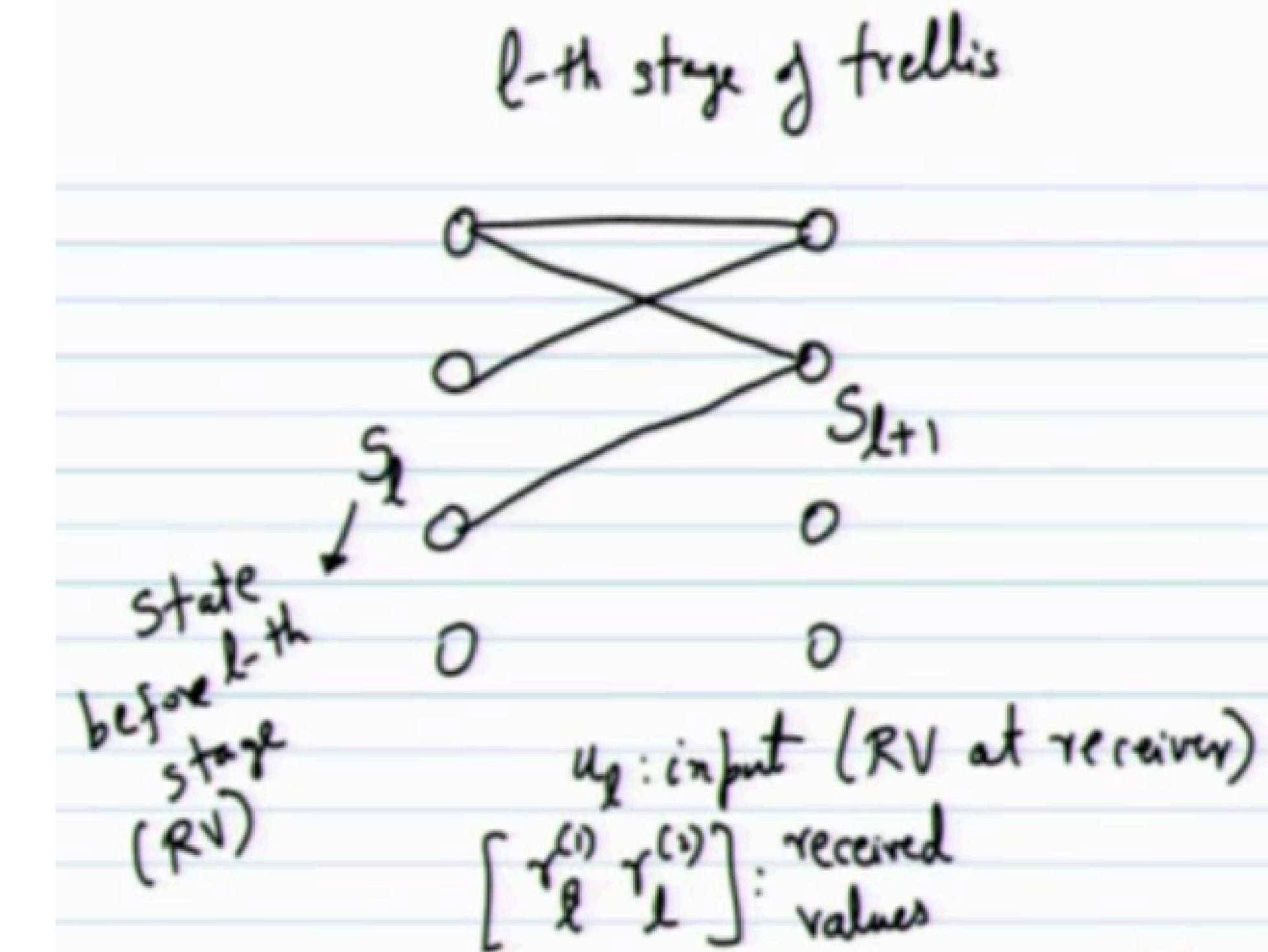
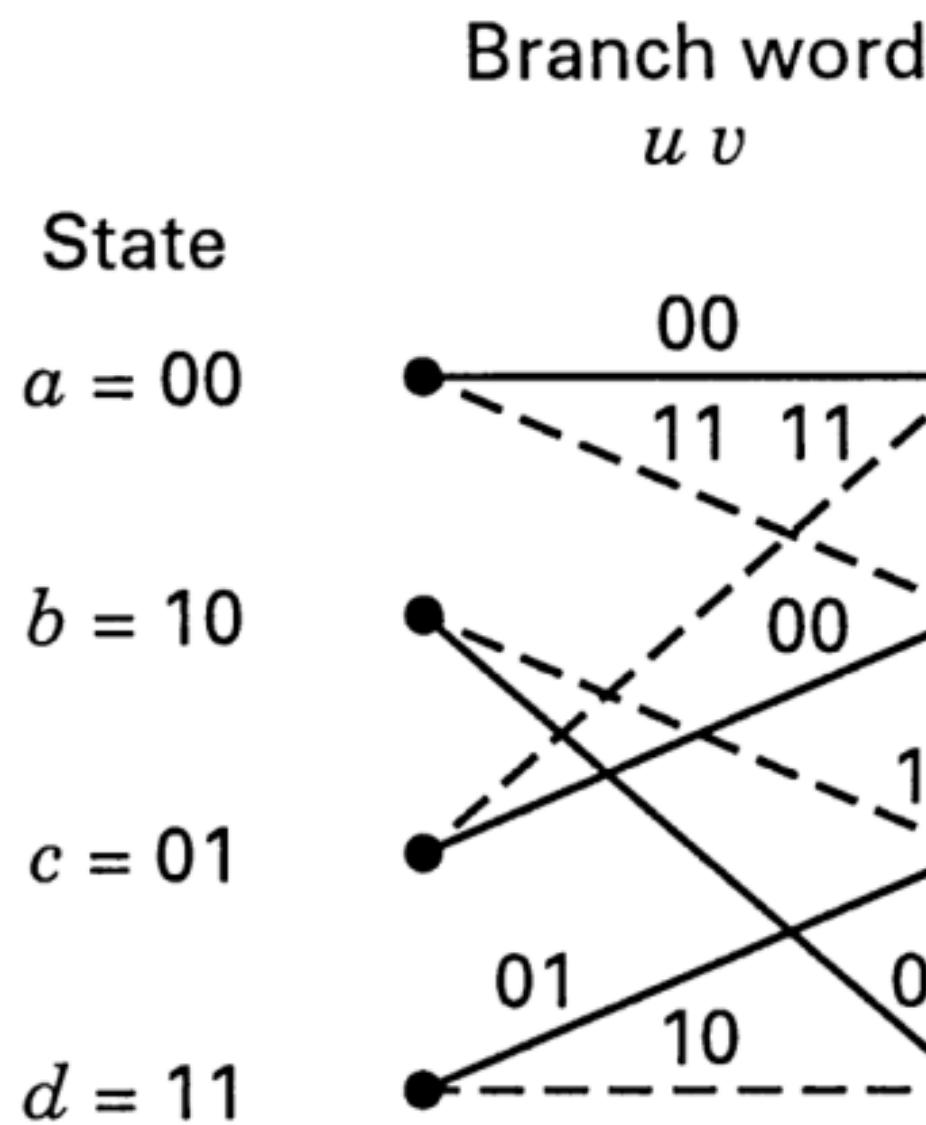


# BCJR ALGORITHM

Maximum A Posteriori(MAP) Decoding Algorithm



# SOME MORE NOTATION



$$P(u_l = +1 | r) = \frac{P(u_l = +1, r)}{P(r)}$$

HERE THE DENOMINATOR HAS NOT BEEN WRITTEN AS WE ARE ANY WAY GOING TO COMPUTE LLR WHERE THE DENOMINATOR GETS CANCELLED

ALPHA'S AND BETA'S GET CALCULATED BY RECURSION GAMMA'S COULD BE DIRECTLY CALCULATED

# THE RESULT

$$p(u_l = 0, r) = \sum_{\substack{s' \rightarrow s: \\ \text{labelled with} \\ \text{input } u_l = 0}} p(S_l = s', S_{l+1} = s, r)$$

we will  
compute this for  
each  $(s', s)$

Main result

$$p(S_l = s', S_{l+1} = s, r) = d_{l-1}(s') \gamma_l(s', s) \beta_l(s),$$

$$d_{l-1}(s') = p(S_l = s', r^{l-1}_0)$$

$$\gamma_l(s', s) = p(S_{l+1} = s, \gamma_l^{(1)}, \gamma_l^{(2)} | S_l = s')$$

$$\beta_l(s) = p(r^{k+1}_{l+1} | S_{l+1} = s)$$

# DERIVATION

Pf:

$$\begin{aligned} p(s_l = s', S_{l+1} = s, \underline{\gamma}) &= p(s', s, \underline{\gamma}_0^{l-1}, \gamma_l, \underline{\gamma}_{l+1}^{k+1}) \\ &= p(\underline{\gamma}_{l+1}^{k+1} | s', s, \underline{\gamma}_0^{l-1}, \gamma_l) \cdot p(s', s, \underline{\gamma}_0^{l-1}) \\ &= p(\underline{\gamma}_{l+1}^{k+1} | s) \cdot p(s, \gamma_l | s', \underline{\gamma}_0^{l-1}) \cdot p(s', \underline{\gamma}_0^{l-1}) \\ &= p(s_l = s', \underline{\gamma}_0^{l-1}) \cdot p(S_{l+1} = s, \gamma_l | S_l = s') \cdot \\ &\quad p(\underline{\gamma}_{l+1}^{k+1} | S_{l+1} = s) \end{aligned}$$

# RECURSION

$\alpha, \beta$ -recursions

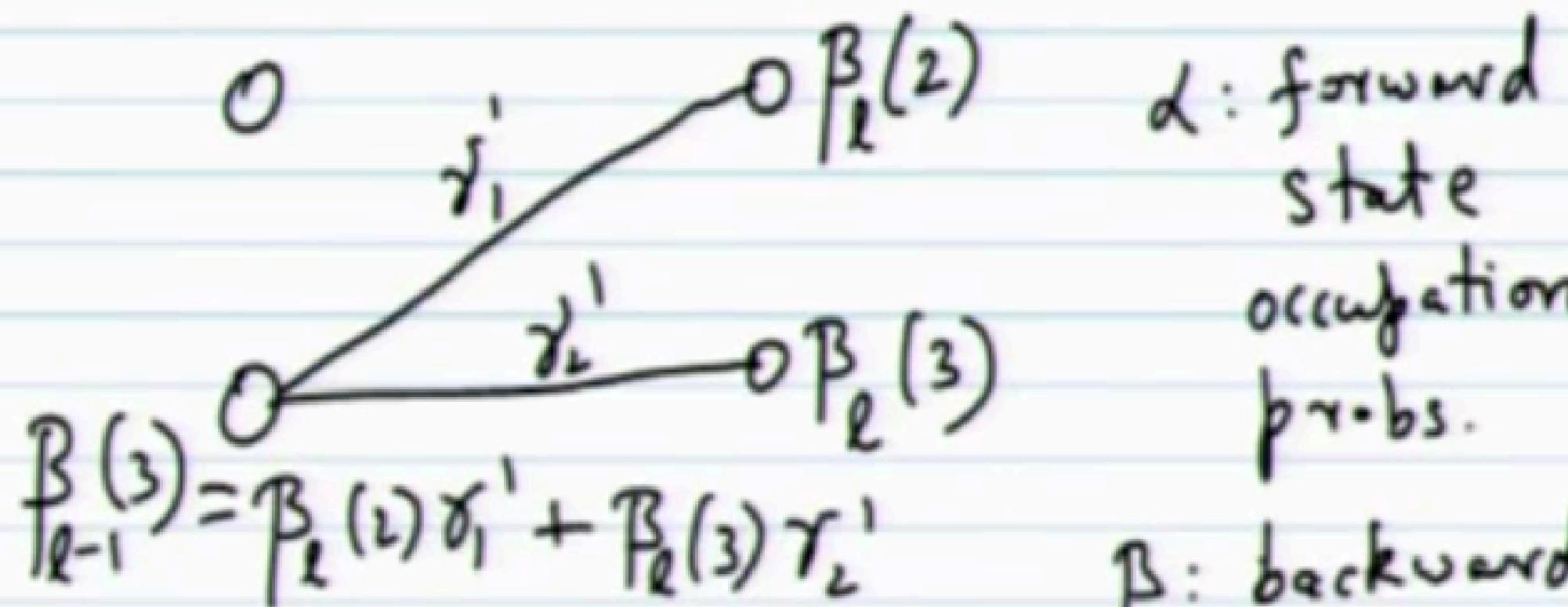
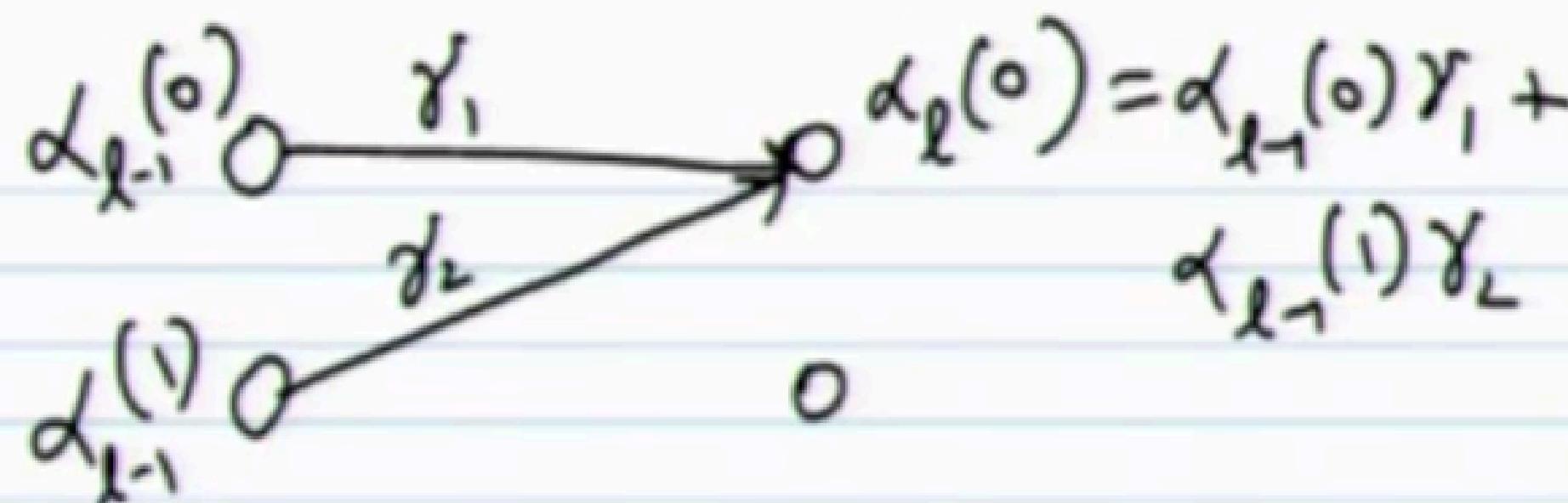
$$d_l(s) = \sum_{s' : s' \rightarrow s \text{ is a branch}} \gamma_l(s', s) d_{l-1}(s')$$

$$p(S_{l+1} = s, \underline{\gamma}_0)$$

$$= \sum_{s'} p[S_{l+1} = s, S_l = s', \underline{\gamma}_0, \gamma_l] d_{l-1}(s')$$

$$= \sum_s p[S_{l+1} = s, \gamma_l | S_l = s, \cancel{\gamma_0}, \cancel{\gamma_l}] d_{l-1}(s')$$

# SEEING IN THE TRELLIS DIAGRAM



$\gamma$ 's: branch metrics

$\alpha$ : forward state occupation

$\beta$ : backward state occupation

$\gamma$ 's: branch prob.

# GAMMA CALCULATION

Branche metric:  $\gamma_L(s', s) = p(S_{L+1} = s, \gamma_L | S_L = s')$

$$= p(S_{L+1} = s | S_L = s') \cdot p(\gamma_L | S_L = s', S_{L+1})$$
 $p(u_L = u(s' \rightarrow s)) = u_L$ 

w/o any  
other info

 $p(\gamma_L^{(1)}, \gamma_L^{(2)} | v^{(1)}(s' \rightarrow s), v^{(2)}(s' \rightarrow s))$ 

$v^{(1)}(s' \rightarrow s), v^{(2)}(s' \rightarrow s)$

$y^{(1)}(s' \rightarrow s), y^{(2)}(s' \rightarrow s)$

IN IMPLEMENTATION, WE STORE THE LOG VALUES OF ALPHA, BETA, GAMMA:

ALPHA'S AND BETA'S GET CALCULATED BY RECURSION, THERE VALUES BECOME VERY SMALL, COMPUTERS CAN'T STORE THEM.

WE ALSO USE WINDOWING AS A LARGE NO OF VALUES HAVE TO BE STORED.

$$\text{LLR}(u_l) = \log \frac{\sum_{\substack{(s', s): \\ \text{input} = 0}} \alpha_{l-1}(s') \gamma_l(s', s) \beta_l(s)}{\sum_{\substack{(s', s): \\ \text{input} = 1}} \alpha_{l-1}(s') \gamma_l(s', s) \beta_l(s)}$$

- implementation: Use  $\log \alpha$ ,  $\log \gamma$ ,  $\log \beta$  +

suitably change recursion.

$$*(z, y) = \log(e^z + e^y) = \text{mix}(z, y) + \log(1 + e^{-|z-y|})$$

INTERLEAVER:  
SHOULD BE ONE-ONE FOR  
INVERTIBILITY BUT BE CLOSE  
TO RANDOM.

IN THE INFO EXCHANGE,  
WHATEVER IT KNOWS ALREADY  
SHOULD NOT BE EXCAHNGED.  
I.E, INTRINSIC LLR'S AND  
EFFECTS OF INFO COMING  
FROM THE PREVIOUS DECODER  
HAVE TO BE SUBTRACTED.,

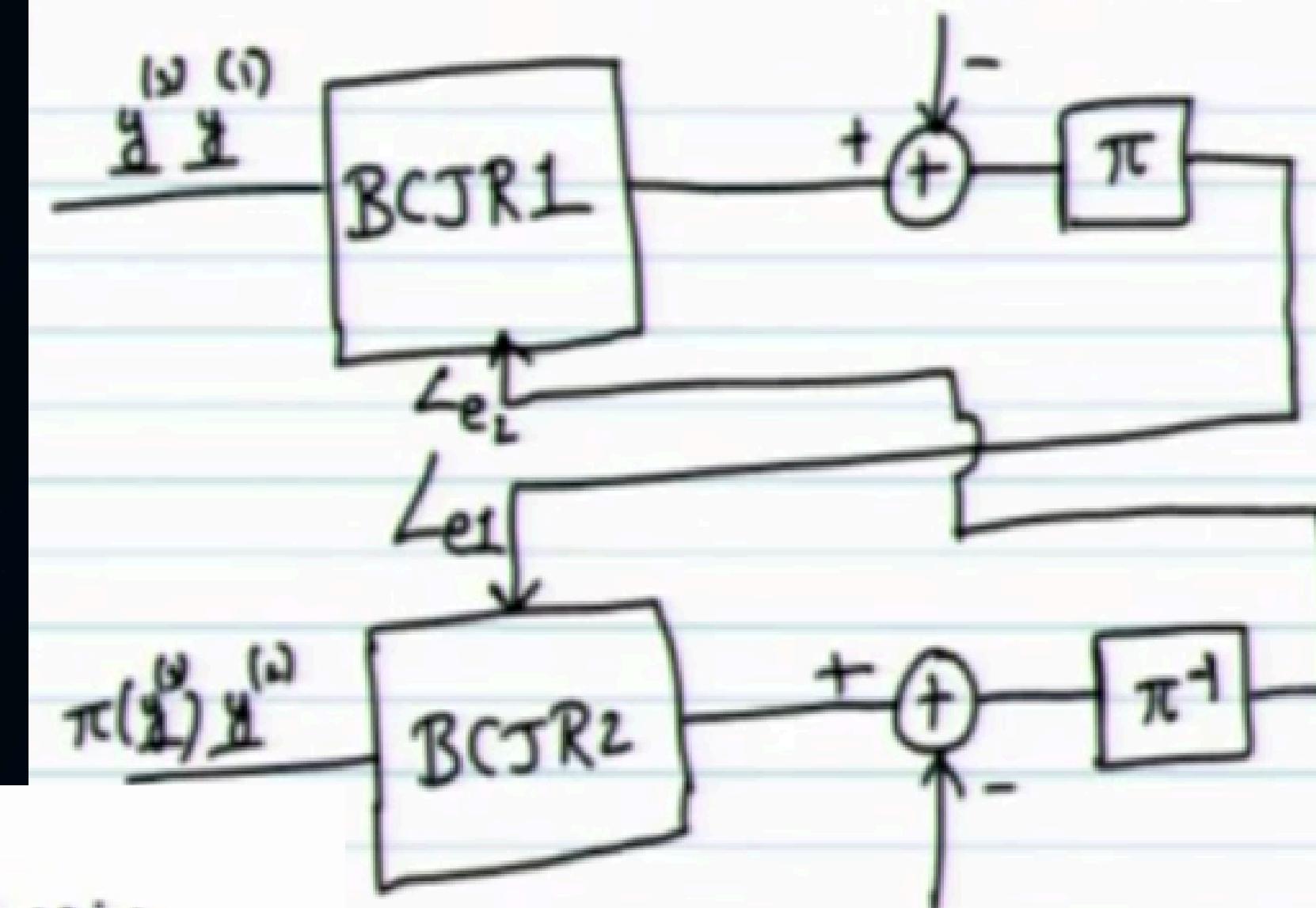
Quadratic Permutation Polynomial (QPP)

interleaver:

$$\pi(i) = f_1 i + f_2 i^2 \rightarrow K$$

$$i=0, 1, \dots, K-1$$

LLR +  
intrinsic (?)



intrinsic +  
effect of  $\underline{e}_L$  (?)

$$\text{Output LLR} = \underline{\delta}_L + \frac{2 \underline{y}_L^{(2)}}{\sigma^2} + \text{extrinsic output LLR}$$

# THANK YOU!

FOR YOUR ATTENTION/PATIENCE