

# Lecture 5

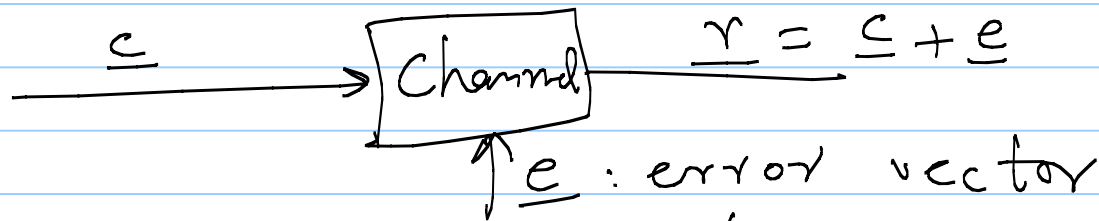
Note Title

1/10/2008

$(n, k, d)$  Code : linear, binary  
→ How do you construct?  
→ How do you decode?

Syndrome Decoder: nearest neighbour for linear codes

$(n, k, d)$  Code  $C$  :  $H \rightarrow$  a parity-check matrix



$$r_i = c_i + e_i$$

$$\Pr(e_i = 1) = p$$

$$\Pr(\underline{e} = \text{all-zero}) = (1-p)^n$$

$$\Pr(\underline{e}) = p^{wt(\underline{e})} (1-p)^{n-wt(\underline{e})}$$

Decoder:  $\min_{\underline{u} \in C} d_H(\underline{x}, \underline{u}) = \min_{\underline{e}: \underline{x} + \underline{e} \in C} \text{wt}(\underline{e})$

$\underline{r} = \underline{c} + \underline{e}$   
 knows this  $\downarrow$  find  $\underline{c}$  (or)  $\underline{e}$   
 $\downarrow$   
 $2^k$  possibilities

Syndrome:  $\underline{s} = H \underline{x}^T = H \underline{e}^T$

→ Any  $\underline{e}$  that satisfies  $H \underline{e}^T = \underline{s}$  will be s.t.  $\underline{x} + \underline{e} \in C$ .

$$\begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_{n-k} \end{bmatrix} \stackrel{H}{=} \begin{bmatrix} e_0 \\ e_1 \\ \vdots \\ e_{n-1} \end{bmatrix}$$

Decoder:  $\hat{\underline{e}} = \arg \min_{\underline{e}: \underline{s} = H\underline{e}^T} wt(\underline{e})$

$$\hat{\underline{c}} = \underline{r} + \hat{\underline{e}}$$

→ Finding minimum weight solution to the linear equation  $\underline{s} = H\underline{e}^T$

Ex: (6, 3, 3) code

$$H = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$

$$\underline{r} = [0 \ 1 \ 0 \ 1 \ 0 \ 1]$$

$$\hat{\underline{e}} = [0 \ 0 \ 1 \ 0 \ 0 \ 0]$$

$$\hat{\underline{c}} = [0 \ 1 \ 1 \ 1 \ 0 \ 1]$$

$$\underline{s} = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} e_0 \\ e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix}$$

Self-study → syndrome table

Complexity:  $(n, k, d)$  Code.

→ Try  $\underline{e}$  of wt  $\leq t$

$$\# = 1 + \binom{n}{1} + \binom{n}{2} + \dots + \binom{n}{t} \approx 2^{n_f(t)} \quad \begin{matrix} e \in [0, 1] \\ \downarrow \\ n_f(t) \end{matrix}$$

→ need to solve for  $\underline{s} = H\underline{e}^T$  smartly....

Ex:

$$H = \begin{matrix} & \begin{matrix} k & n & m \end{matrix} \\ \begin{matrix} n-k \\ \end{matrix} & \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

$$\begin{matrix} n = 8 \\ k = 4 \end{matrix}$$

$$\rightarrow \underline{r} = [1 \ 0 \ 0 \ 1 \ 1 \ 1 \ 0 \ 0] = \underline{\hat{e}} \quad \underline{s} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\rightarrow \underline{r} = [0 \ 0 \ 1 \ 1 \ 0 \ 1 \ 0 \ 1] \quad \underline{s} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \quad \begin{matrix} \underline{\hat{e}} = [1 \ 0 \ 0 \ 1 \ 0 \ 0 \ 0 \ 0] \\ \underline{e} = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 1] \end{matrix}$$

## Minimum Distance & H

$$\rightarrow \underline{c} \in C \quad H \underline{c}^T = \underline{0}$$

$$\begin{bmatrix} \underline{h}_0^T & \underline{h}_1^T & \dots & \underline{h}_{n-1}^T \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_{n-1} \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$c_0 \underline{h}_0^T + c_1 \underline{h}_1^T + \dots + c_{n-1} \underline{h}_{n-1}^T = \underline{0}$$

$\text{wt}(\underline{c}) = w$  &  $w$  1s in  $\underline{c}$  are at positions  $i_1, \dots, i_w$



$$c_{i_1} \underline{h}_{i_1}^T + c_{i_2} \underline{h}_{i_2}^T + \dots + c_{i_w} \underline{h}_{i_w}^T = \underline{0}$$

$\Rightarrow$  Columns  $i_1, i_2, \dots, i_w$  of  $H$  are linearly dependent.

$\Rightarrow$  Min dist of  $C = \min \#$  of linearly dependent columns of  $H$ .