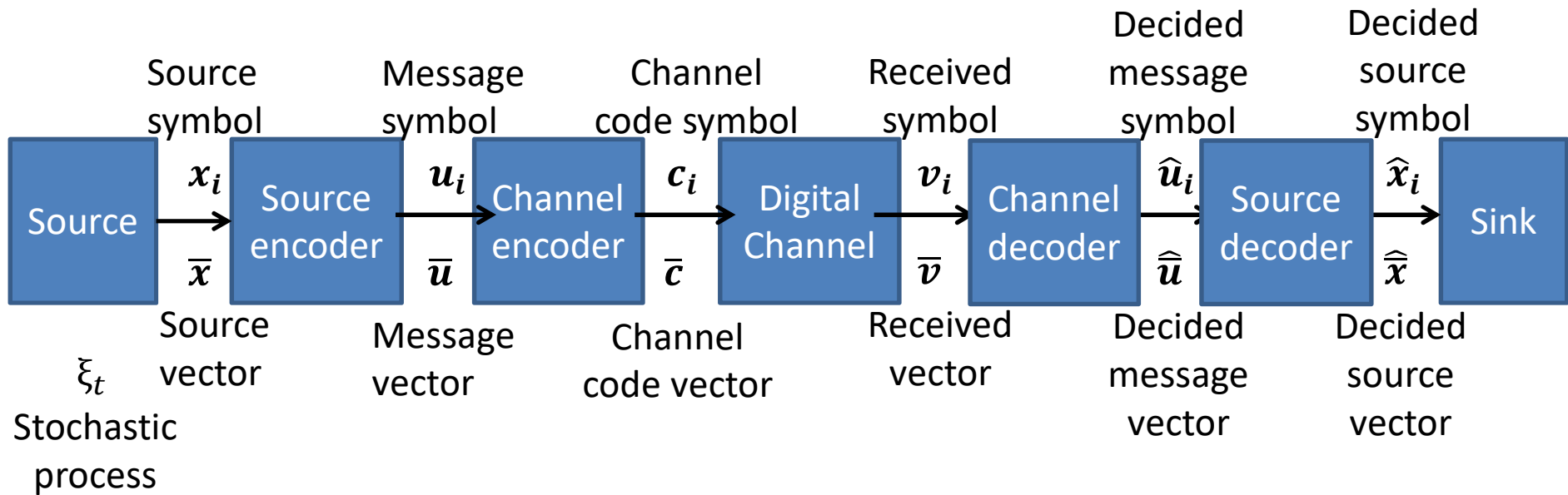


Úrkommunikáció
Space Communication
2023/6.

Channel Coding

Error correction coding



Channel encoding rule

$$\Omega(\bar{u}) = \bar{c}$$

Decoding in 2 steps

$$D(\bar{v}) = \hat{\bar{c}}$$

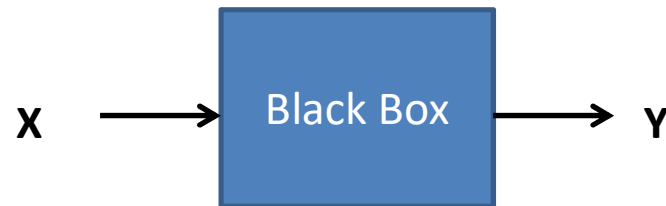
1. Decision

$$\Omega^{-1}(\hat{\bar{c}}) = \hat{\bar{u}}$$

2. Invers operation

Digital transmission channel

Input X and output Y are discrete random variables



How many information can we gather about X by observing Y?

- a-posteriori Entropy
- Mutual Information

Definition: a-posteriori Entropy [bit/symbol, Shannon/symbol]

$$H(X|Y) = E\{I(x|y)\} = \sum_x \sum_y p(x, y) \log_2 \frac{1}{p(x|y)}$$

Digital transmission channel

Definition: Mutual Information of two random events

$$I(x_i; y_j) = \log_2 \frac{p(x_i | y_j)}{p(x_i)} = \log_2 \frac{p(y_j | x_i)}{p(y_j)} \quad [\text{bit}, \text{Shannon}]$$

Using Bayes's theorem:

$$\log_2 \frac{p(x_i | y_j)}{p(x_i)} = \log_2 \frac{p(x_i | y_j)p(y_j)}{p(x_i)p(y_j)} = \log_2 \frac{p(x_i, y_j)}{p(x_i)p(y_j)} = \log_2 \frac{p(y_j | x_i)p(x_i)}{p(x_i)p(y_j)} = \log_2 \frac{p(y_j | x_i)}{p(y_j)}$$

Definition: Average mutual information $[\text{bit}/\text{symbol}]$, $[\text{Shannon}/\text{symbol}]$

$$\begin{aligned} I(X; Y) &= E\{I(x_i; y_j)\} = \sum_x \sum_y p(x_i, y_j) I(x_i; y_j) = \\ &= \sum_x \sum_y p(x_i, y_j) \log_2 \frac{p(x_i, y_j)}{p(x_i)p(y_j)} = D(p(x, y) || p(x) \cdot p(y)) = \\ &= \sum_x \sum_y p(x_i, y_j) \log_2 \frac{p(x_i | y_j)}{p(x_i)} = \sum_x \sum_y p(x_i, y_j) \left[\log_2 \frac{1}{p(x_i)} - \log_2 \frac{1}{p(x_i | y_j)} \right] = \\ &= \sum_x \sum_y p(x_i, y_j) \log_2 \frac{1}{p(x_i)} - \sum_x \sum_y p(x_i, y_j) \log_2 \frac{1}{p(x_i | y_j)} = \\ &= \quad H(X) \quad - \quad H(X | Y) \end{aligned}$$

Channel Capacity

$$I(X ; Y) = H(X) - H(X | Y) = H(Y) - H(Y | X) = D(p(x, y) || p(x) \cdot p(y))$$

Definition: Channel capacity $\left[\frac{\text{Shannon}}{\text{channel use}} \right], \left[\frac{\text{bit}}{\text{channel use}} \right], e.g. [\text{bit/sec}]$

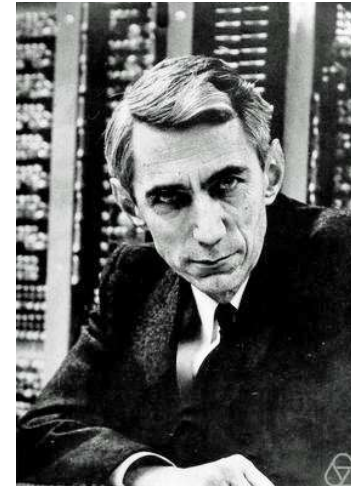
$$\begin{aligned} C &= \max_{p(x)} I(X ; Y) = \\ &= \max_{p(x)} [H(X) - H(X|Y)] = \\ &= \max_{p(x)} [H(Y) - H(Y|X)] = \\ &= \max_{p(x)} D(p(x_i, y_j) || p(x_i) \cdot p(y_j)) \end{aligned}$$

Channel capacity is bounded

$$0 \leq C \leq \max_{p(x)} [H(X)] = H_0 = ld n$$

X and Y are independent:
 $H(X) = H(X|Y)$

Error free channel: $y_i = x_i$
 $p(x_i | y_i) = 1$
 $H(X|Y) = 0$



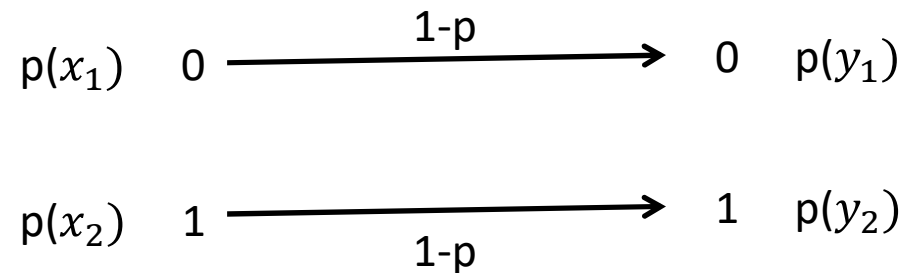
Forrás: www.techzibits.com

Ideal Binary Channel

- Ideal: No parameter, that is only one parameter: error probability $p=0$

- Binary in- and output:

$$X = \{x_1, x_2\} \text{ e.g: } \{0,1\} \quad Y = \{y_1, y_2\} \text{ e.g: } \{0,1\}$$



- How much is the capacity?

Starting form here:

$$C_{ideal\ binary}(p=0) = \max_{p(x)} [H(X) - H(X|Y)] = H_0(X) = \log_2 2 = 1 \left[\frac{\text{bit}}{\text{channel use}} \right]$$

Capacity of BSC

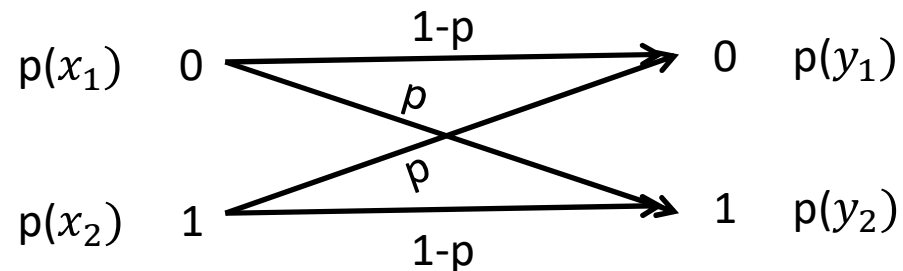
BSC: **B**inary **S**ymmetric **C**hannel

- One parameter: error probability p

- Binary in- and output:

$$X = \{x_1, x_2\} \text{ pl.: } \{0,1\} \quad Y = \{y_1, y_2\} \text{ pl.: } \{0,1\}$$

- Symmetric:



- How much is the capacity?

Starting now form here:
$$C_{BSC}(p) = \max_{p(x)} [H(Y) - H(Y|X)]$$

Capacity of BSC

$$C_{BSC}(p) = \max_{p(x)} [H(Y) - H(Y|X)] \quad [\text{bit/channel use}]$$

- $H(Y)$ maximal if Y is uniformly distributed: $p(y_1) = p(y_2) = 1/2$
And then $H(Y)=1$ [bit/binary symbol]

- In the case of BSC the output is uniformly distributed for example when the input is a such:

$$p(x_1) = p(x_2) = \frac{1}{2}, \text{ then:}$$

$$\begin{aligned} p(y_1) &= p(x_1) \cdot (1 - p) + p(x_2) \cdot p = p(x_1) - p(x_1) \cdot p + (1 - p(x_1)) \cdot p = \\ &= p(x_1) - 2 \cdot p(x_1) \cdot p + p = p(x_1) - p + p = 1/2 \end{aligned}$$

- $H(Y|X) = \sum_x \sum_y p(x, y) \log_2 \frac{1}{p(y|x)} = \sum_x \sum_y p(x) p(y|x) \log_2 \frac{1}{p(y|x)} =$
 $= \sum_x p(x) \sum_y p(y|x) \log_2 \frac{1}{p(y|x)}$

$$p(x_1) \cdot \left[(1 - p) \cdot \log_2 \frac{1}{(1-p)} + p \cdot \log_2 \frac{1}{p} \right] + p(x_2) \cdot \left[p \cdot \log_2 \frac{1}{p} + (1 - p) \cdot \log_2 \frac{1}{(1-p)} \right] =$$

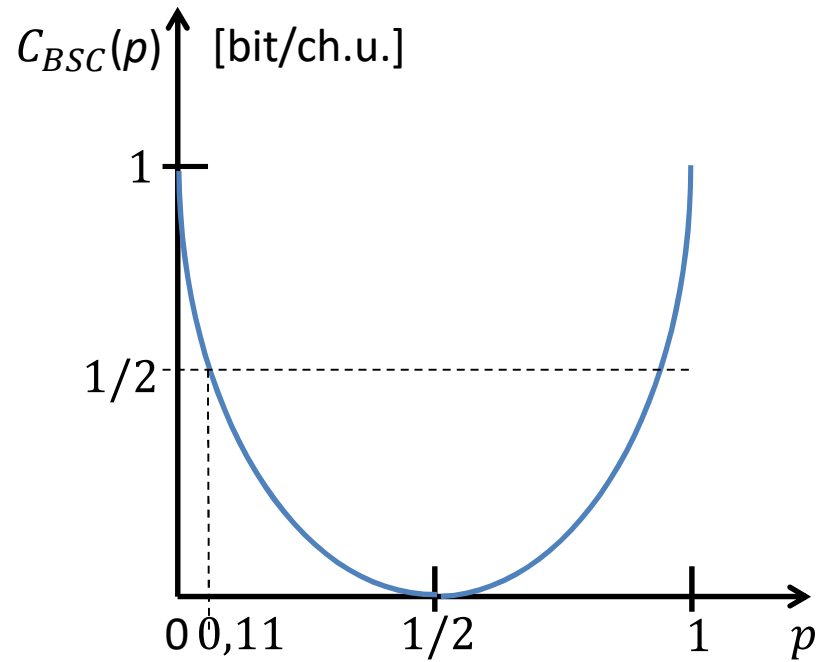
$$\underbrace{(p(x_1) + p(x_2))}_1 \cdot \underbrace{\left[(1 - p) \cdot \log_2 \frac{1}{(1-p)} + p \cdot \log_2 \frac{1}{p} \right]}_{h(p) \text{ binary entropy function}}$$

1

$h(p)$ binary entropy function

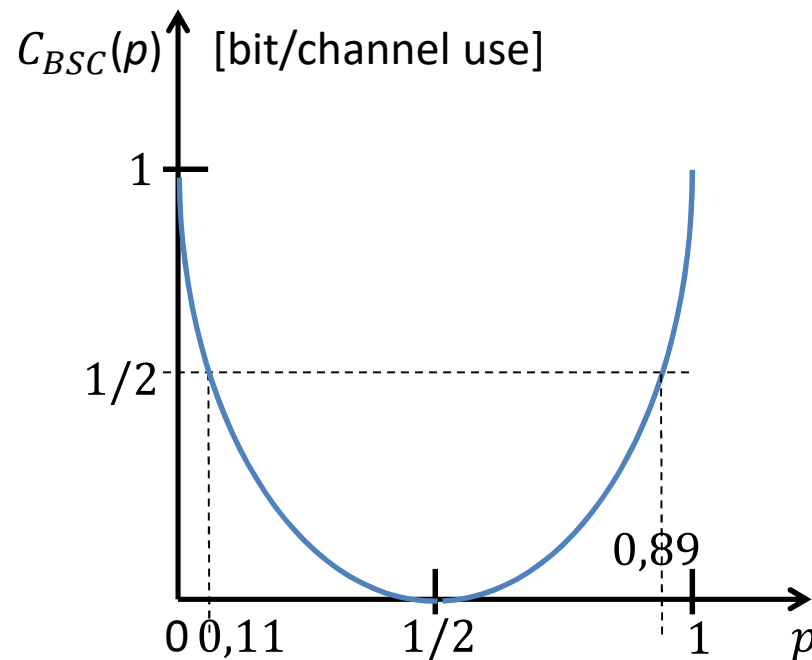
Capacity of BSC

$$C_{BSC}(p) = \max_{p(x)} [H(Y) - H(Y|X)] = 1 - h(p) \quad [\text{bit/channel use}]$$

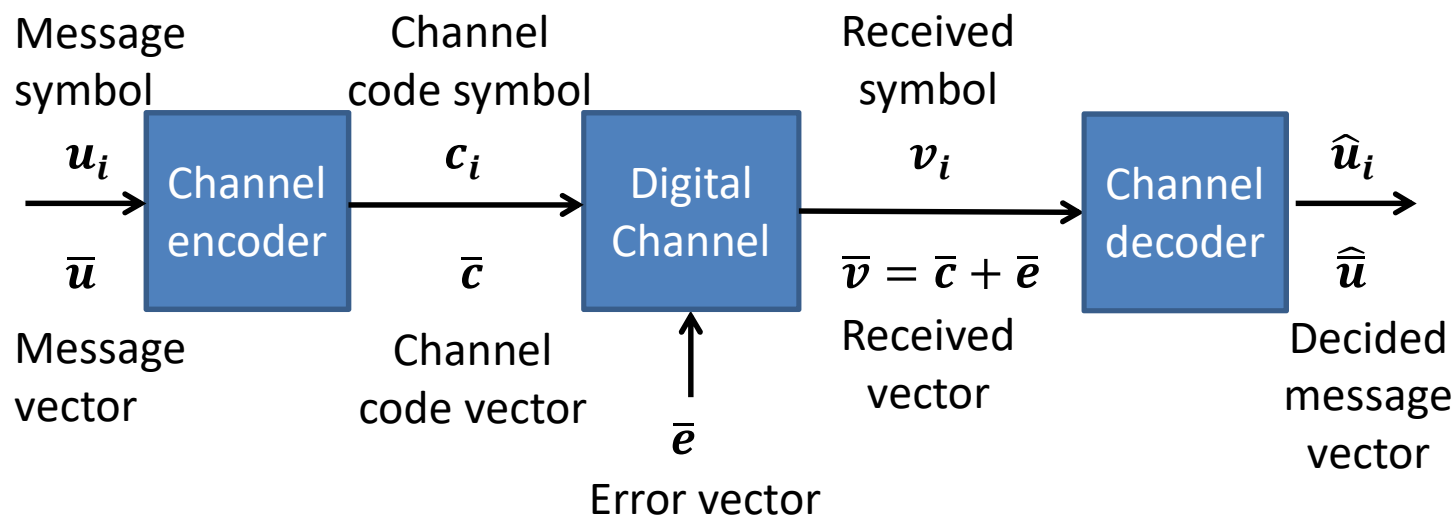


Shannon's Channel coding theorem, 1948

- If $H(X) < C$ then exists $\Omega(X) = \tilde{X}$ transformation (coding, modulation, method) so that $P_e \rightarrow 0$ until $H(\tilde{X}) < C$ holds.
- An other formulation (case of Block coding):
A vector (block) of K message symbols extended to a vector of N code symbols
 $P_e \rightarrow 0$, until $\lim_{K \rightarrow \infty} K/N < C$ remains valid.



Coding, Construction of codes



Channel encoding rule

$$\Omega(\bar{u}) = \bar{c}$$

Decoding in 2 steps

$$D(\bar{v}) = \hat{c}$$

1. Decision

$$\Omega^{-1}(\hat{c}) = \hat{u}$$

2. Invers operation

Coding rule: Mutually obvious transformation of Message space into Code space

Coding, Construction of codes

Message space: $\bar{U} = \{\bar{u}_i\}$

Message vector: $\bar{u}_i = [u_1, u_2, \dots, u_k, \dots, u_K]$

Message symbol: $u_k = \{0, 1, 2, \dots, r-1\}$

Dimension K, r-ary vector space

r^K possible message vector

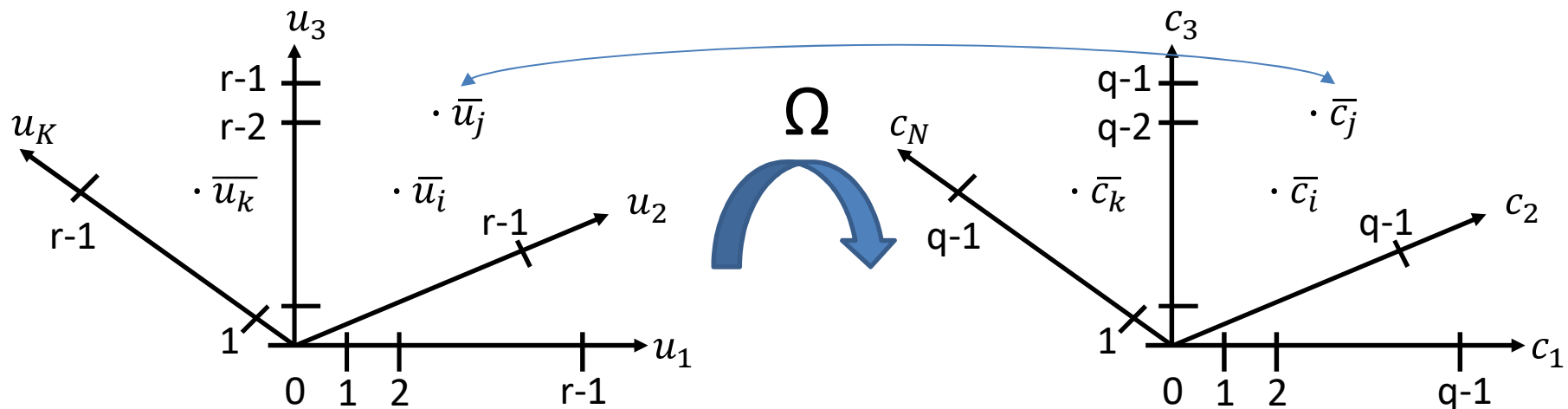
Code space: $\bar{C} = \{\bar{c}_i\}$

Code vector: $\bar{c}_i = [c_1, c_2, \dots, c_n, \dots, c_N]$

Code symbol: $c_n = \{0, 1, 2, \dots, q-1\}$

Dimension N, q-ary vector space

q^N possible code vector



Decoding

Error vector: $\bar{e} = [e_1, e_2, \dots, e_n, \dots, e_N]$

Example: two error events: $\bar{e} = [0, 0, \dots, e_i, \dots, e_j, \dots, 0, \dots, 0]$ if the events happens at position i and j, and the symbol values of the events are e_i and e_j .

Four unknown (positions and values) should be determined.

Example for „deleted” error type: $\bar{e}_{del} = [0, 0, \dots, *j, \dots, 0, \dots, 0]$.

The position is known.

Received vector: $\bar{v} = \bar{c} + \bar{e} = [v_1, v_2, \dots, v_n, \dots, v_N] = [c_1 + e_1, \dots, c_n + e_n, \dots, c_N + e_N]$

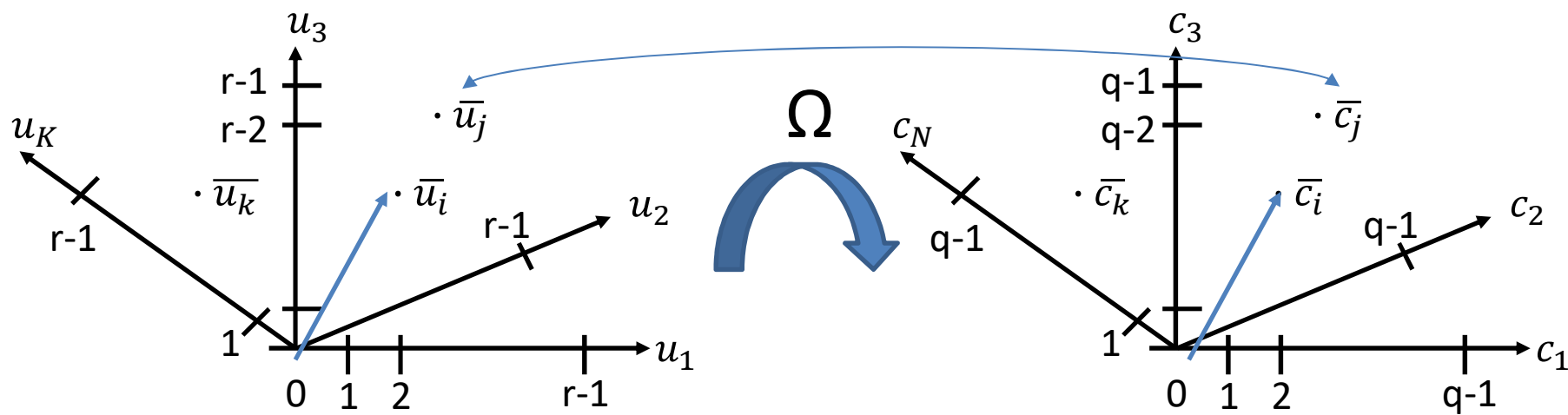
Based on the received vector: \bar{v} Decoding in 2 steps

1. Decision $D(\bar{v}) = \hat{\bar{c}}$

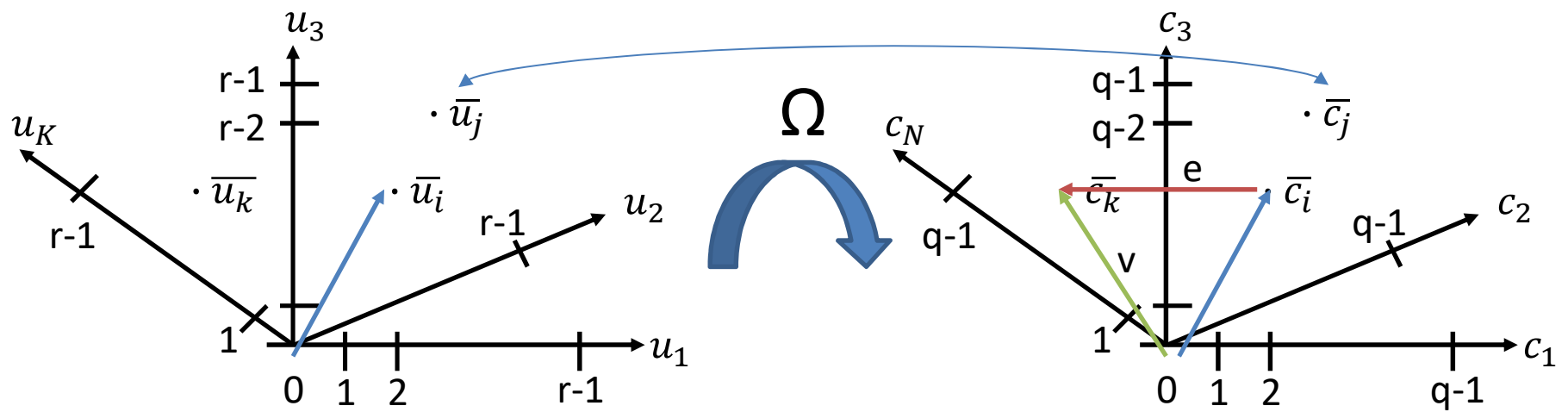
2. Invers operation $\Omega^{-1}(\hat{\bar{c}}) = \hat{\bar{u}}$

- Trivial: $\bar{v} = \bar{c}_i$
- Unsolvable: $\bar{v} = \bar{c}_j \neq \bar{c}_i$ that we sent
- Solvable with the possibility of wrong decision: $\bar{v} \neq \bar{c}_i \forall i$

Trivial



Unsolvable



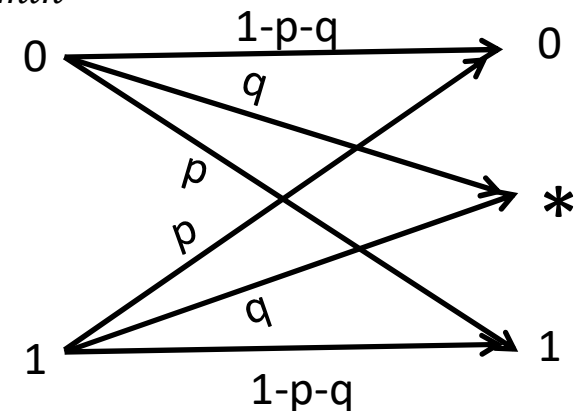
Coding, Construction of codes

Definitions for calculation in vector space:

- Hamming distance: $d(\bar{c}_i, \bar{c}_j) = \sum_{n=1}^N \chi(c_{i_n} \neq c_{j_n})$
- Code distance: $d_{min} = \min_{i,j \neq i} \{d(\bar{c}_i, \bar{c}_j)\}$
- Code weight: $w = \min_{i, \bar{c}_i \neq 0} \{\sum_{n=1}^N \chi(c_{i_n} \neq 0)\}$

Type of errors:

- Number of detectable error: $t_{det} < d_{min}, t_{det_{max}} = d_{min} - 1$
- Number of correctable errors: $t_{corr} = \left\lfloor \frac{d_{min}-1}{2} \right\rfloor$
- Number of „deleted” type errors: $t_{del} = d_{min} - 1$

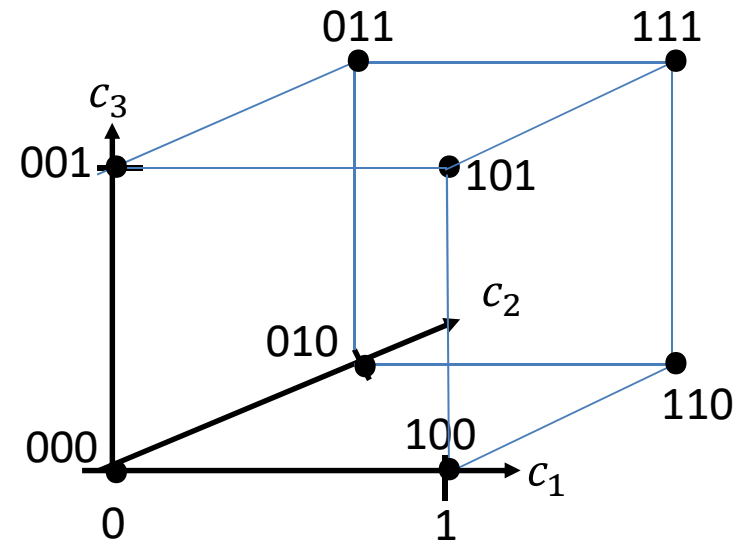
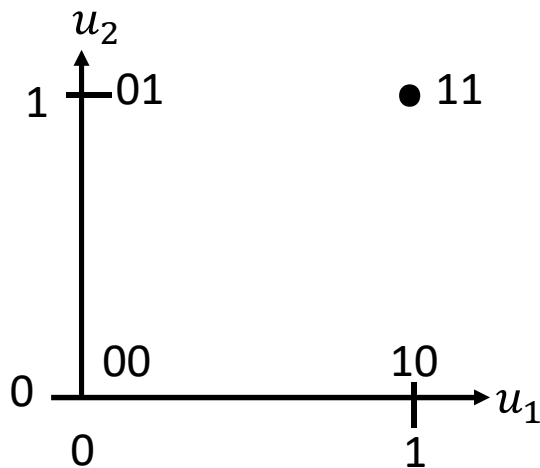


Basic types of Channel coding (error correction coding)

- Block codes (N,K,q): Hamming, Cyclic, Reed-Solomon, etc.
- Convolutional coding (Trellis codes, Viterbi coding/decoding)

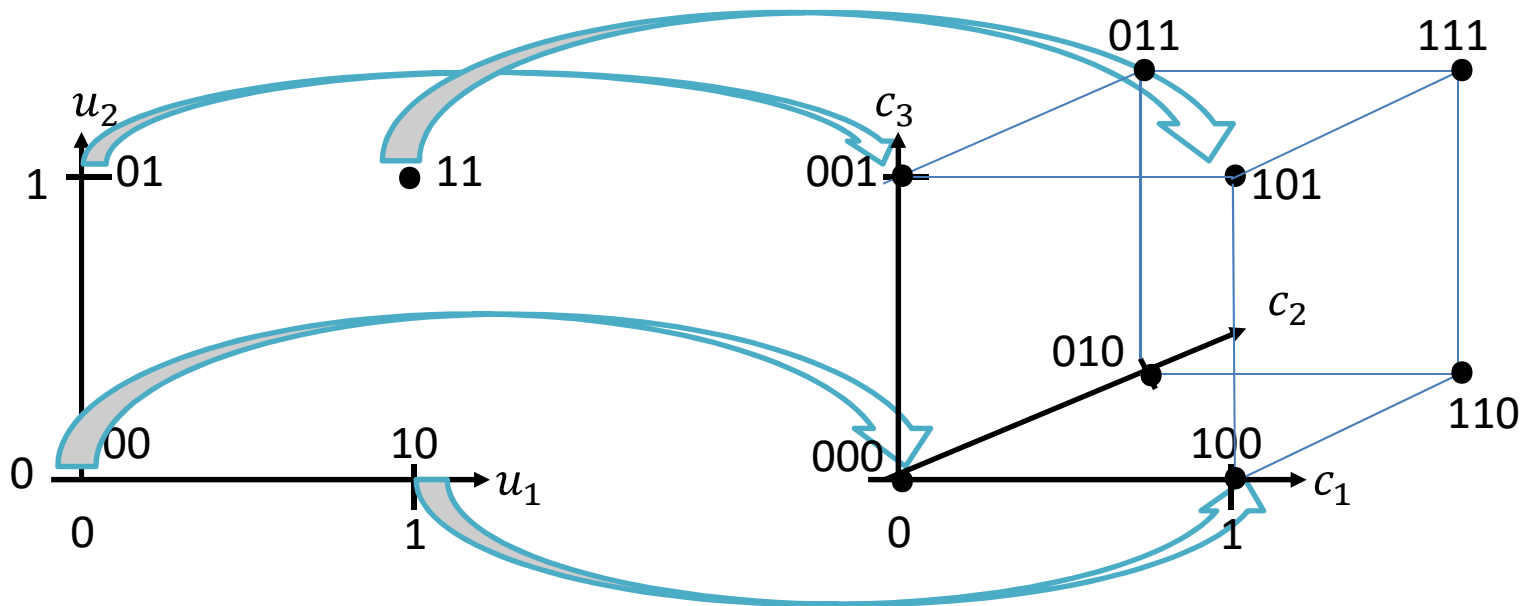
Let us start the encoding heuristically

(N=3, K=2, q=r=2) binary message and vector space, +1 redundant binary symbol



Construction heuristically

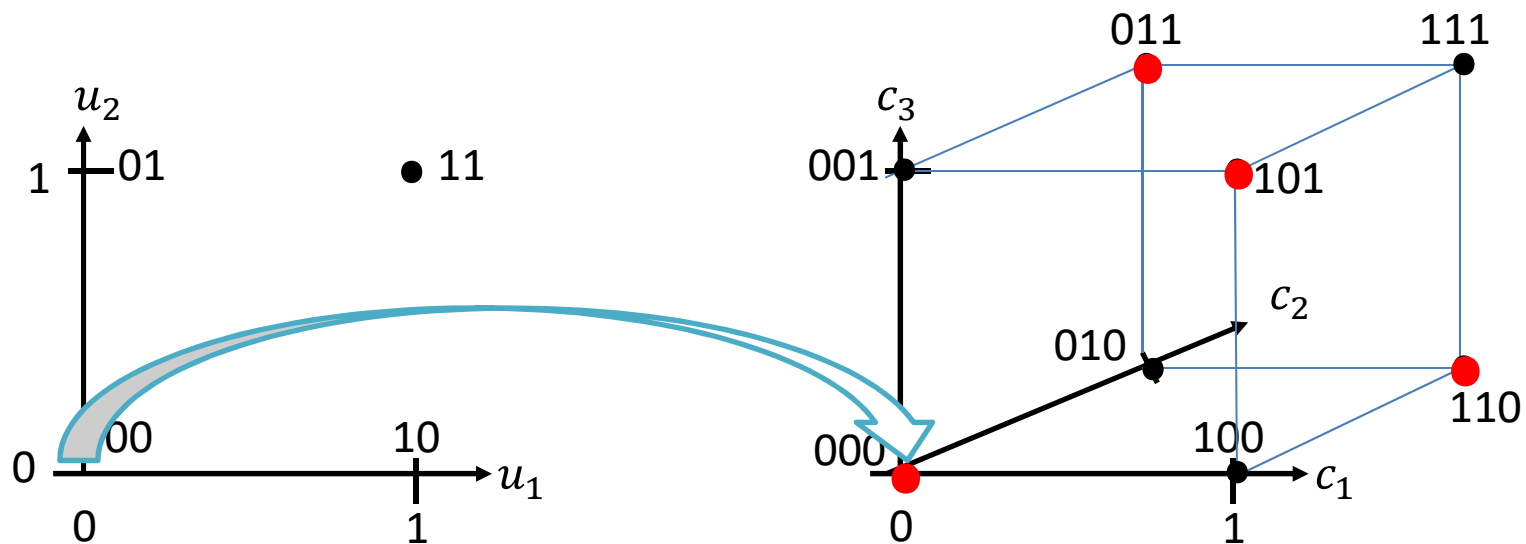
($N=3$, $K=2$, $q=r=2$) binary message and vector space, +1 redundant binary symbol



Although mutually obvious transformation, not appropriate, because the distances remaining the same

Construction heuristically

($N=3, K=2, q=r=2$) binary message and vector space, +1 redundant binary symbol

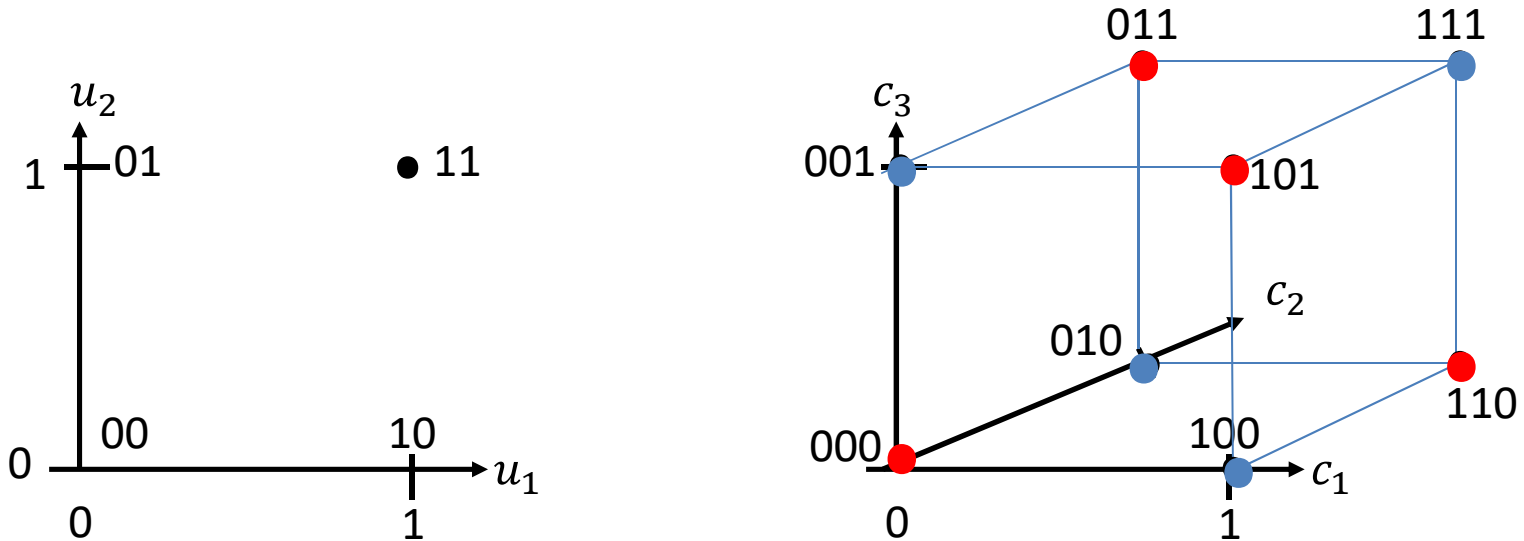


Mutually obvious transformation, appropriate, because increasing distances, $d_{\min}=2$
 (Parity check coding, cyclic coding heuristically!)

Message vector	Code vector
0 0	0 0 0
1 0	1 0 1
0 1	0 1 1
1 1	1 1 0

Construction heuristically

(N=3, K=2, q=r=2) binary message and vector space, +1 redundant binary symbol



Vector set represented by blue dots could detect one error because mutually obvious and increasing distances, $d_{min}=2$, however, they are not a linear subspace (see later) of the vector space, therefore not appropriate for calculations.

(cyclic but not parity check vector set!)

Another transformation	Message vector	Code vector
	0 0	1 0 0
	1 0	0 0 1
	0 1	0 1 0
	1 1	1 1 1

Example: correcting one

„delete” error, red also

$c=1$	0	0
$v=*$	0	0
$v=1$	*	0
$v=1$	0	*

Algebraic code construction rules

Singleton bound for $(N, K, q=r)$ block codes: $M \leq q^{N-d_{min}+1}$

The number of possible code vectors (therefore message vectors) related to the code attributes d_{min} , N and q .

Proof:

K dimensional q -ary space: $M \leq q^K$, max. $M = q^K \Rightarrow d_{min} = 1$,

Extended to N dimensional: $K \rightarrow N \Rightarrow q^{N-K}$ times more point, max. $d_{min} = N - K + 1$

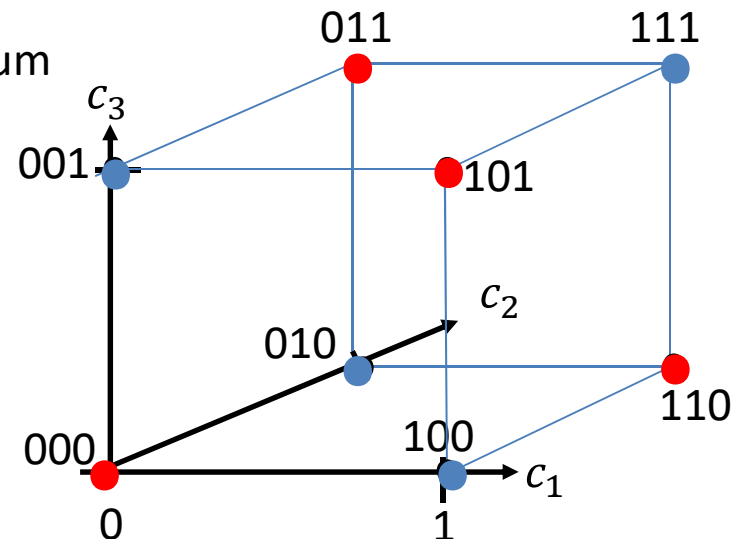
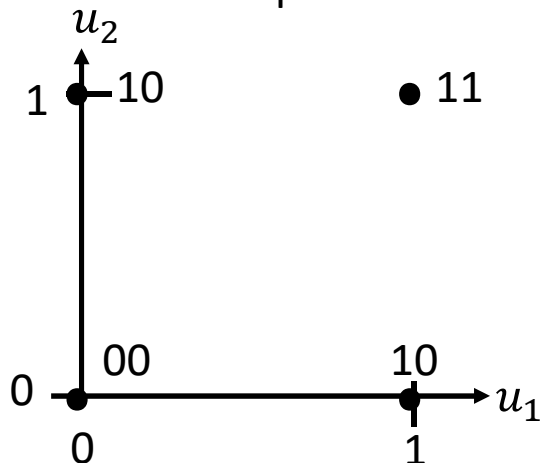
$$d_{min} \leq N - K + 1, \quad K \leq N - d_{min} + 1, \quad M \leq q^K \leq q^{N-d_{min}+1}$$

MDS code (Maximum Distance Separable): $M = q^{N-d_{min}+1}$

Or equivalently: $K = N - d_{min} + 1$ or $d_{min} = N - K + 1$

MDS example : $(N=3, K=2, q=2)$, $d_{min} = 2$

$M = q^{N-d_{min}+1} = 4$ possible messages maximum



Code construction rules cont.

Hamming bound (sphere-packing bound): t_{corr} required, $(N, K, q=r)$?

Determine the number of points in a decision subspace around a valid code vector; the size of decision subspace,

needed for correction of $t_{corr} = \left\lfloor \frac{d_{min}-1}{2} \right\rfloor$ errors

$$1 + N \cdot (q - 1) + \binom{N}{2} \cdot (q - 1)^2 + \dots + \binom{N}{t_{corr}} \cdot (q - 1)^{t_{corr}} = \sum_{i=0}^{t_{corr}} \binom{N}{i} \cdot (q - 1)^i$$

$$\max. M = q^K$$

$$q^K \cdot \sum_{i=0}^{t_{corr}} \binom{N}{i} \cdot (q - 1)^i \leq q^N ;$$

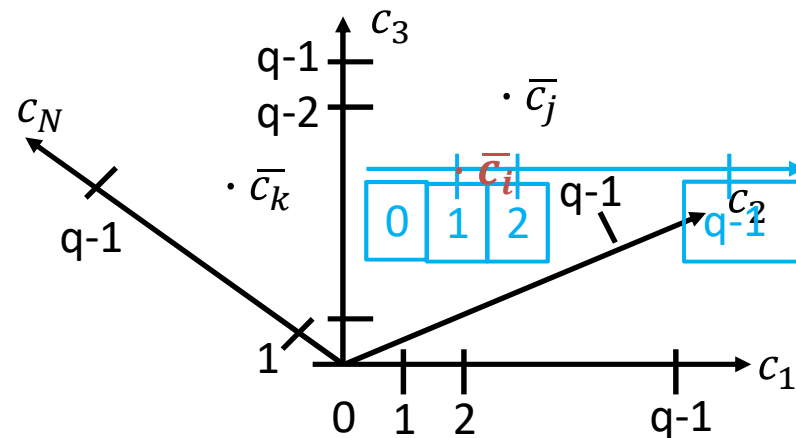
Hamming bound:

$$\sum_{i=0}^{t_{corr}} \binom{N}{i} \cdot (q - 1)^i \leq q^{N-K}$$

Binary case: $\sum_{i=0}^{t_{corr}} \binom{N}{i} \leq 2^{N-K}$

Perfect code:

$$\sum_{i=0}^{t_{corr}} \binom{N}{i} \cdot (q - 1)^i = q^{N-K}$$



Code construction rules cont.

Hamming bound (sphere-packing bound): t_{corr} required, $(N,K,q=r)$?

HOWEVER: Not only the size but also the form of the decision subspaces counts.

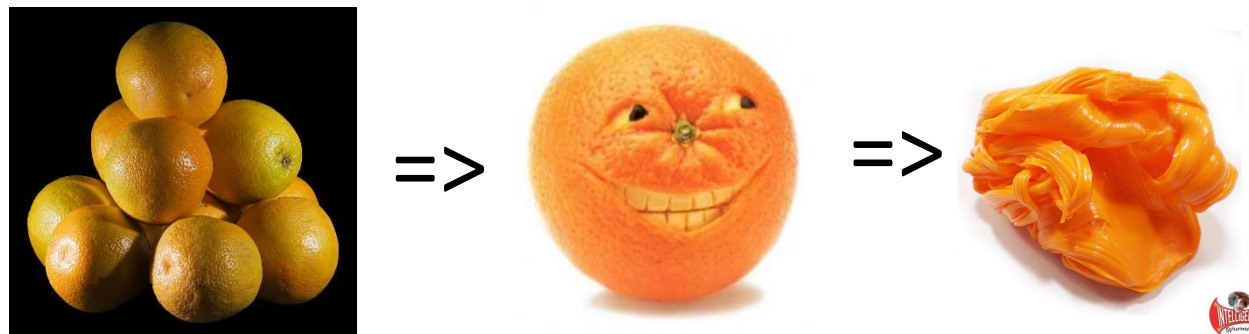
Example: Perfect; $t_{corr} = \left\lfloor \frac{d_{min}-1}{2} \right\rfloor = 2; q = 2;$

$$1 + N \cdot (q - 1) + \binom{N}{2} \cdot (q - 1)^2 = q^{N-K}$$

$$1 + N + \frac{N \cdot (N - 1)}{2} = 2^{N-K}$$

$N=90$ and $K=78$ solves the equation, however, doesn't exist in a 90 dimensional space $302231454903657293676544 (=2^{78})$ portion of disjoint decision subspaces so that every one vector of the space (2^{90} piece) is part of one and only one decision subspace.

The Hamming bound is just a bound for the size.



Examples: Hamming bound, perfect code

Hamming code (N,K,q): Perfect code, that capable to correct maximum one error, and detect max. two errors. In the practice mostly binary, however non-binary Hamming codes also exists.

$$t_{corr} = \left\lfloor \frac{d_{min}-1}{2} \right\rfloor = 1; \Rightarrow d_{min} \geq 3$$

For $t_{corr} = 1$ the Hamming bound is performed perfectly with the (N,K,q) set $\forall m \geq 2$

$$\left(N = \frac{q^m - 1}{q - 1}, K = N - m, q \right) \quad (m=1 \Rightarrow K=0 !!!)$$

Proof:

$$1 + N \cdot (q - 1) = q^{N-K} = q^m; \quad N \cdot (q - 1) = q^m - 1; \quad N = \frac{q^m - 1}{q - 1}$$

MDS \Rightarrow
 $d_{min} = N - K + 1 = 3$

	q=2			q=3			q=5		
m	N	K	$R_c = K/N$	N	K	$R_c = K/N$	N	K	$R_c = K/N$
2	3	1	1/3	4	2	1/2	6	4	2/3
3	7	4	0,57	13	10	0,77	31	28	0,9
4	15	11	0,73	40	36	0,9	156	152	0,97
5	31	26	0,84	121	116	0,96	⋮	⋮	⋮

Example: MDS, perfect code

Hamming (N=3, K=1, q=r=2) code

MDS: $d_{min} = N - K + 1 = 3$; $t_{corr} = \left\lfloor \frac{d_{min}-1}{2} \right\rfloor = 1$

Perfect: $1 + N \cdot (q - 1) = q^{N-K} = 1 + 3 = 2^2 = 4$ elements in a decision subspace

Message: 0 or 1

Code vector: 000 or 111

That is a simple repetition coding!

