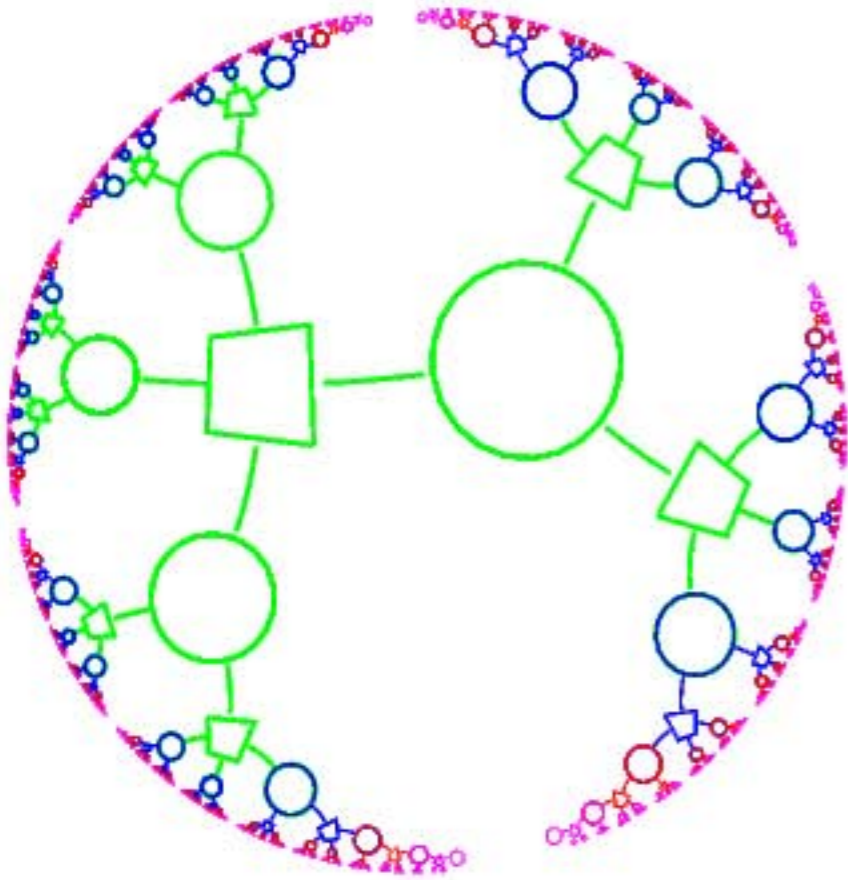
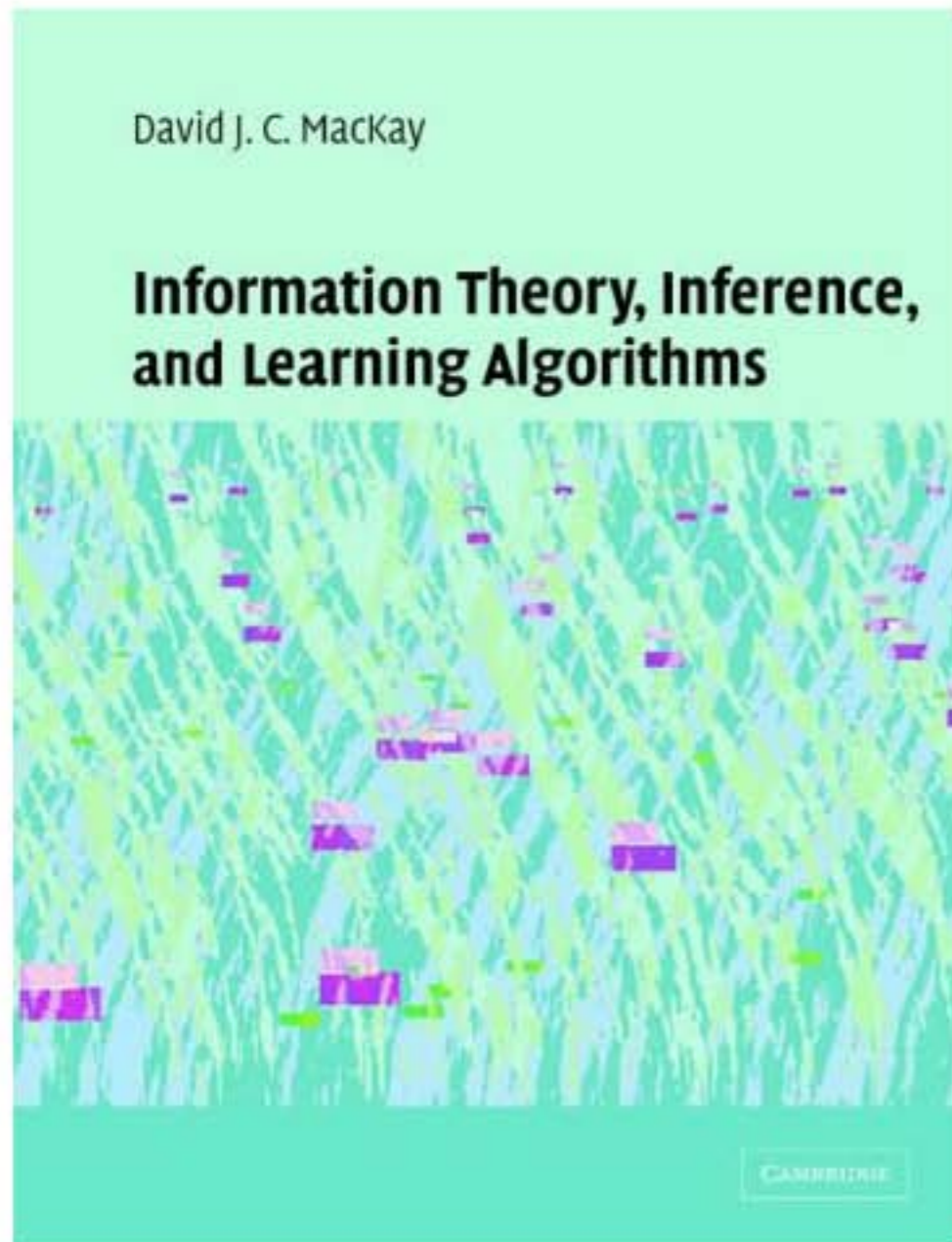


Information theory

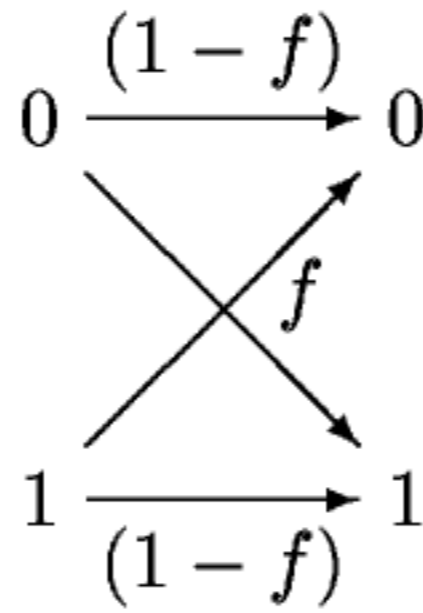


Lecture notes - Chapter 1

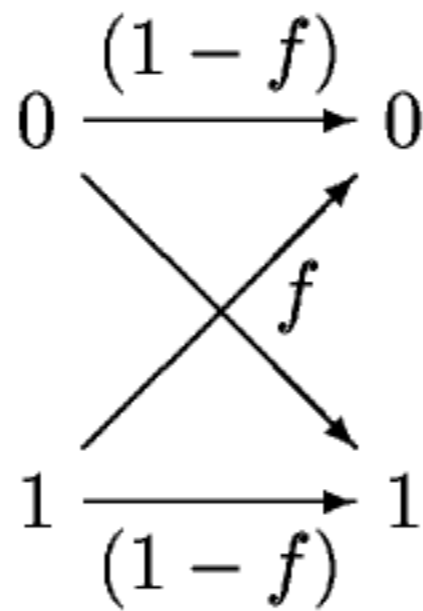


- Cambridge University Press
- 640 pages, 35 pounds
- Also available **free online**

www.inference.phy.cam.ac.uk/mackay/itila/



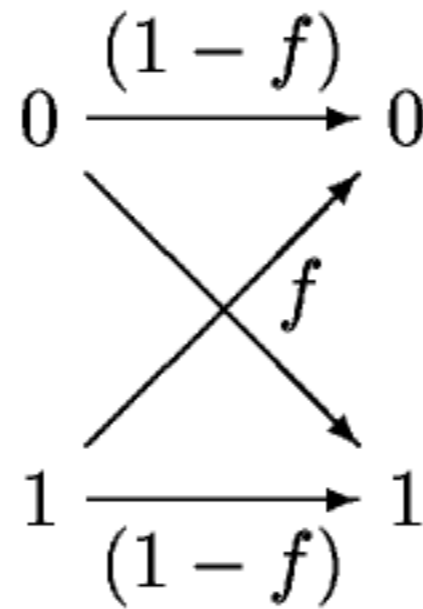
$$f = 0.1$$



Q: A file of $N = 10\,000$ bits is stored on this disc drive (with $f = 0.1$), then read.

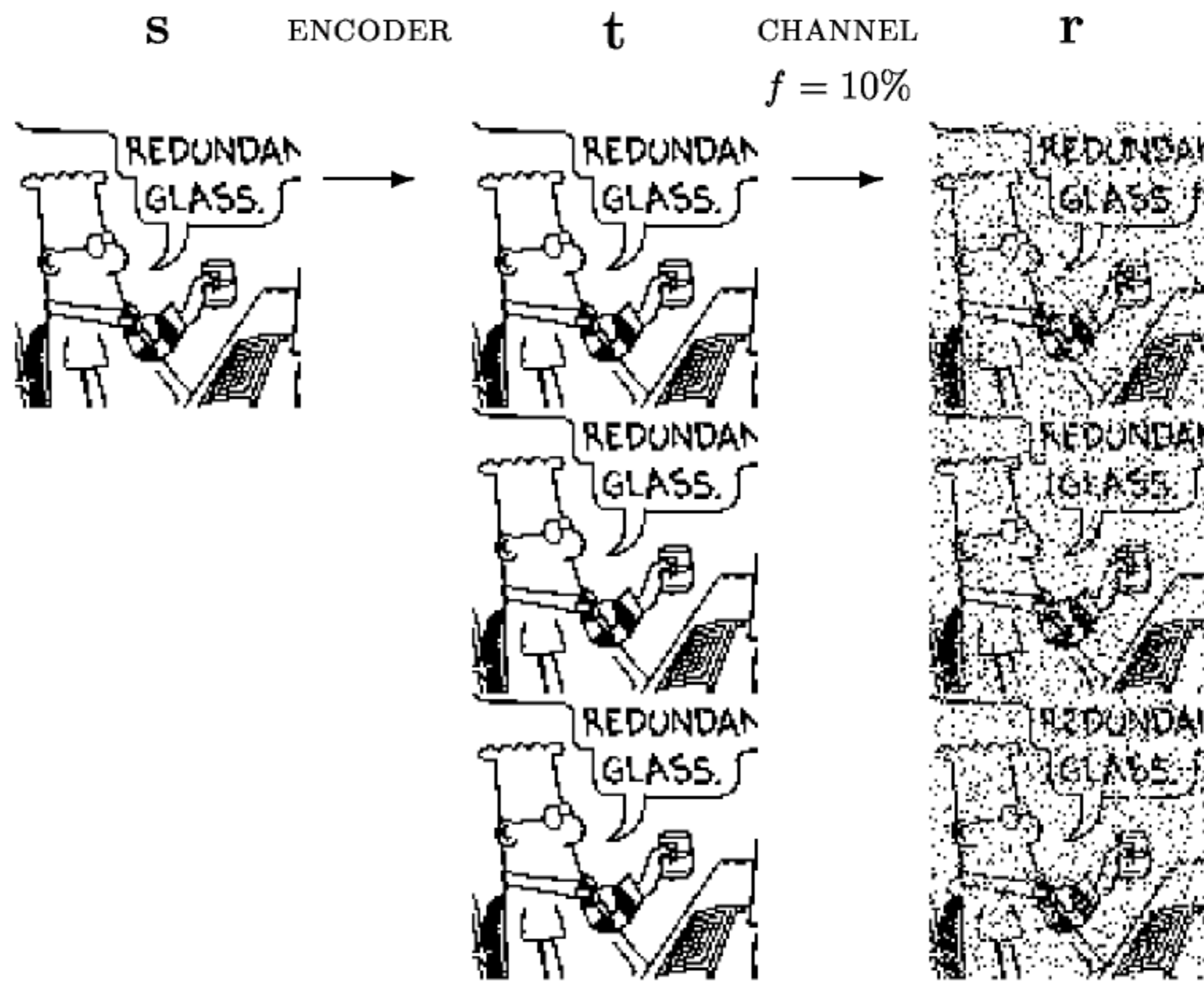
Roughly how many bits are flipped?

\pm



Q: To make a successful business selling 1 Gigabyte disc drives, how small does the flip probability f need to be?

Repetition code 'R3'



Repetition code 'R3'

S

ENCODER

t

CHANNEL

r

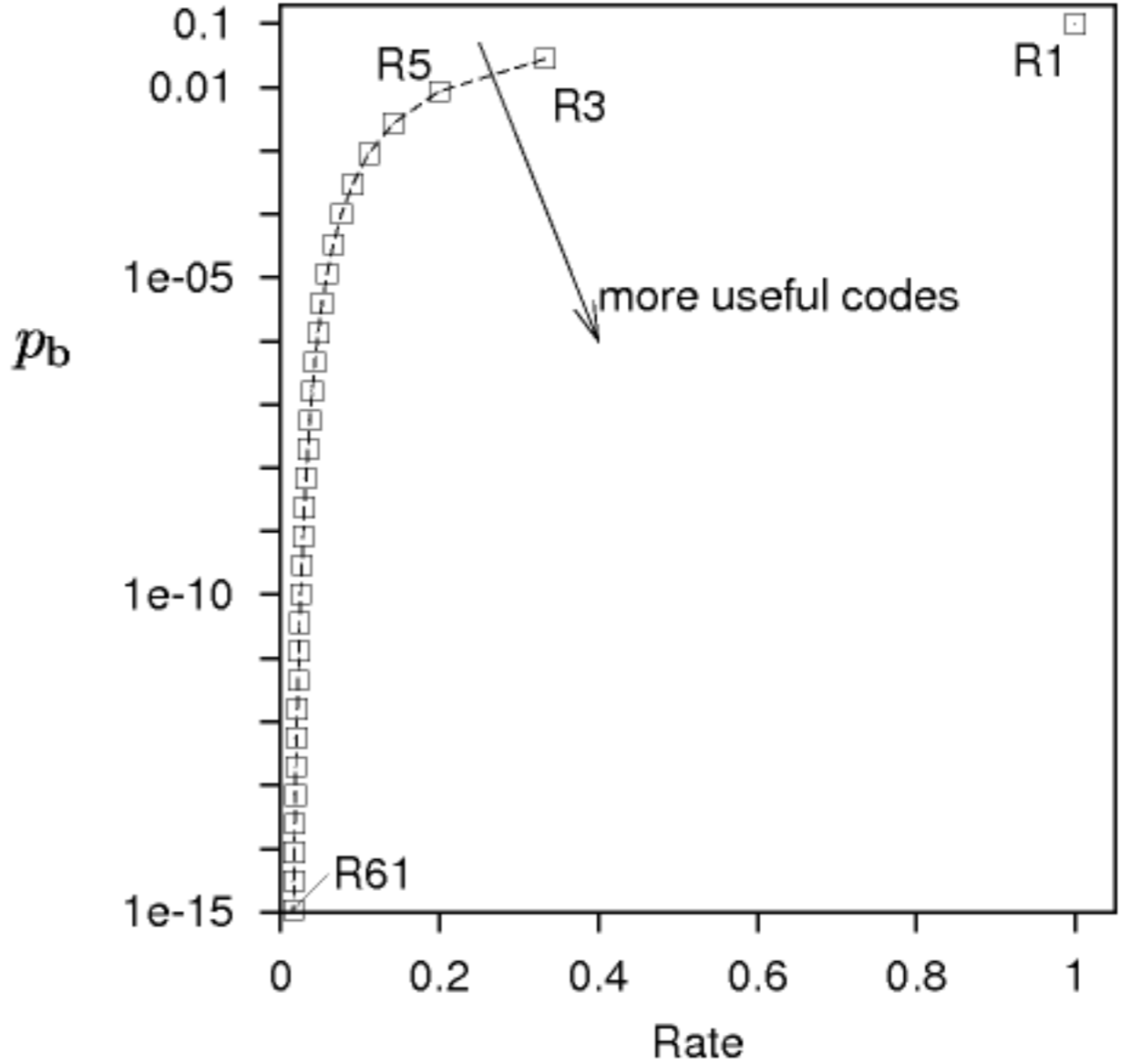
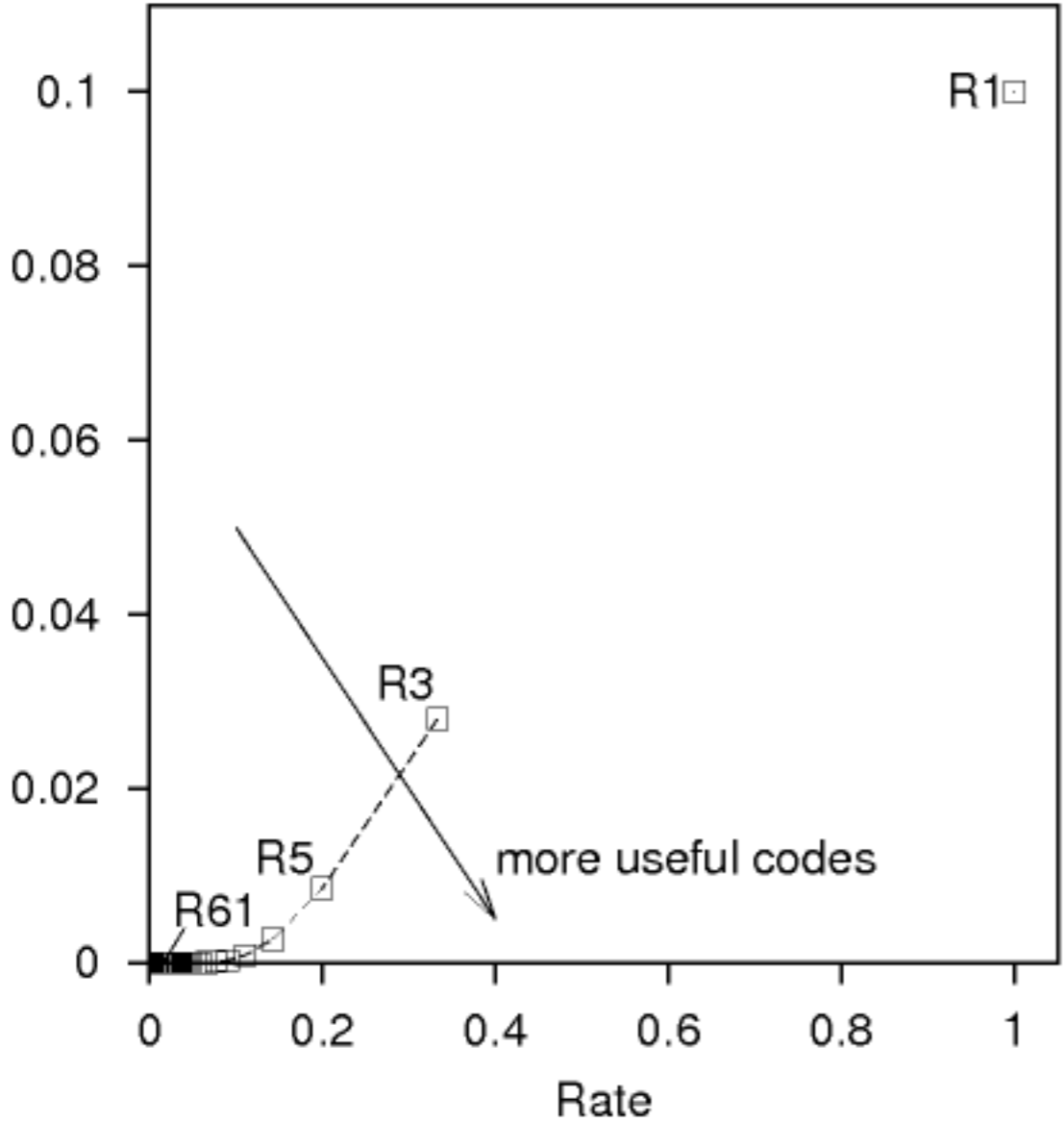
DECODER

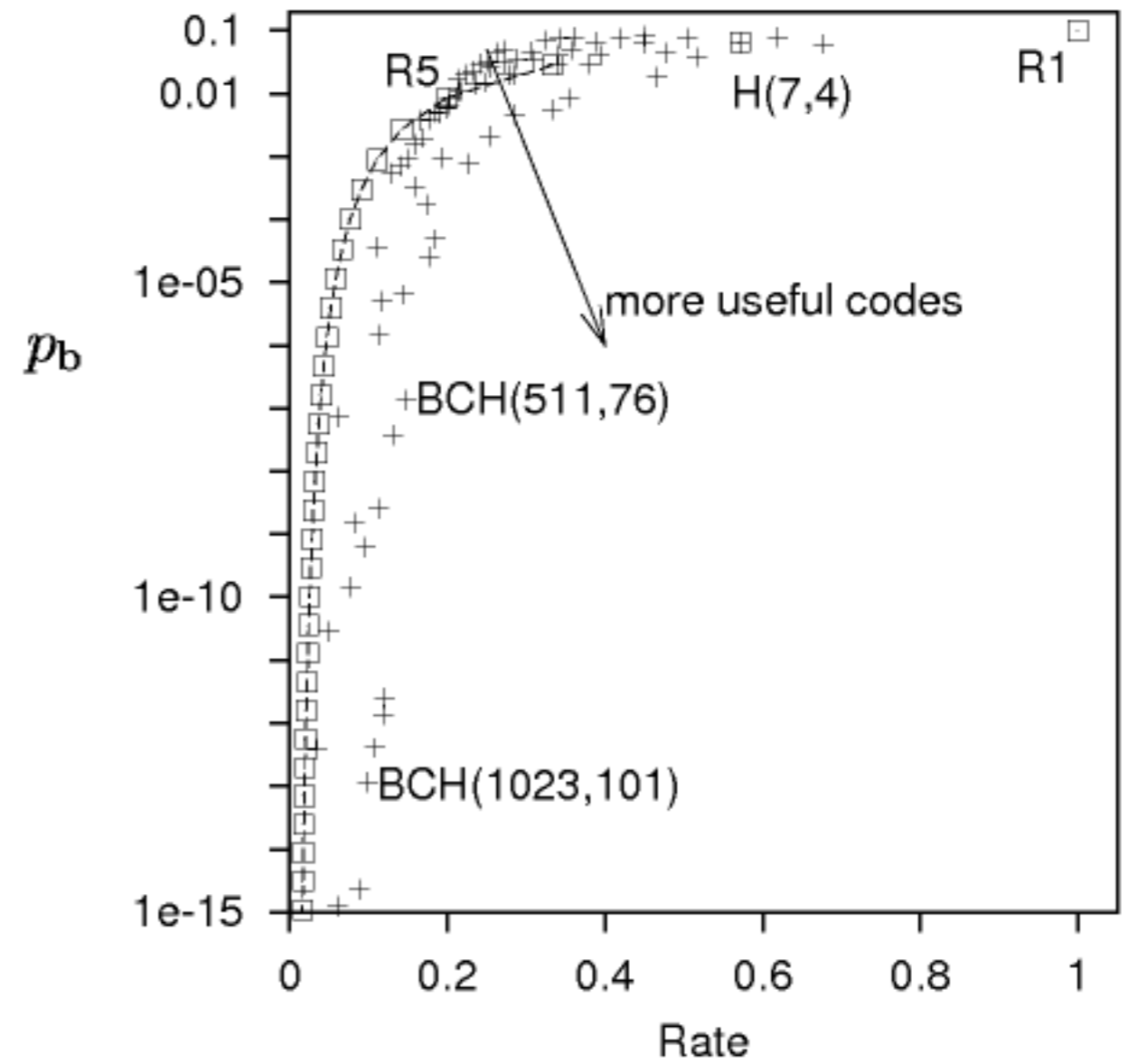
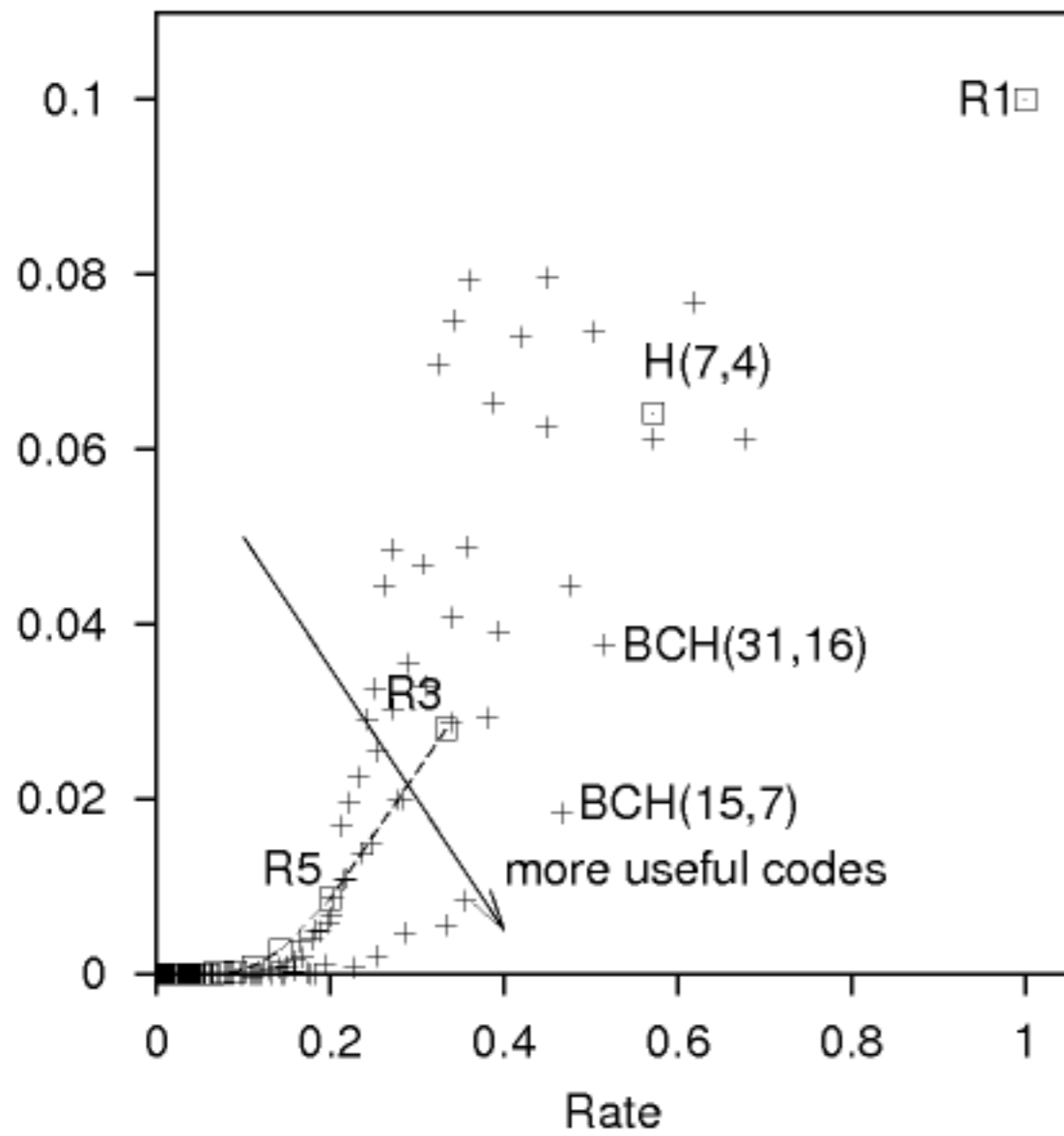
\hat{S}

$f = 10\%$



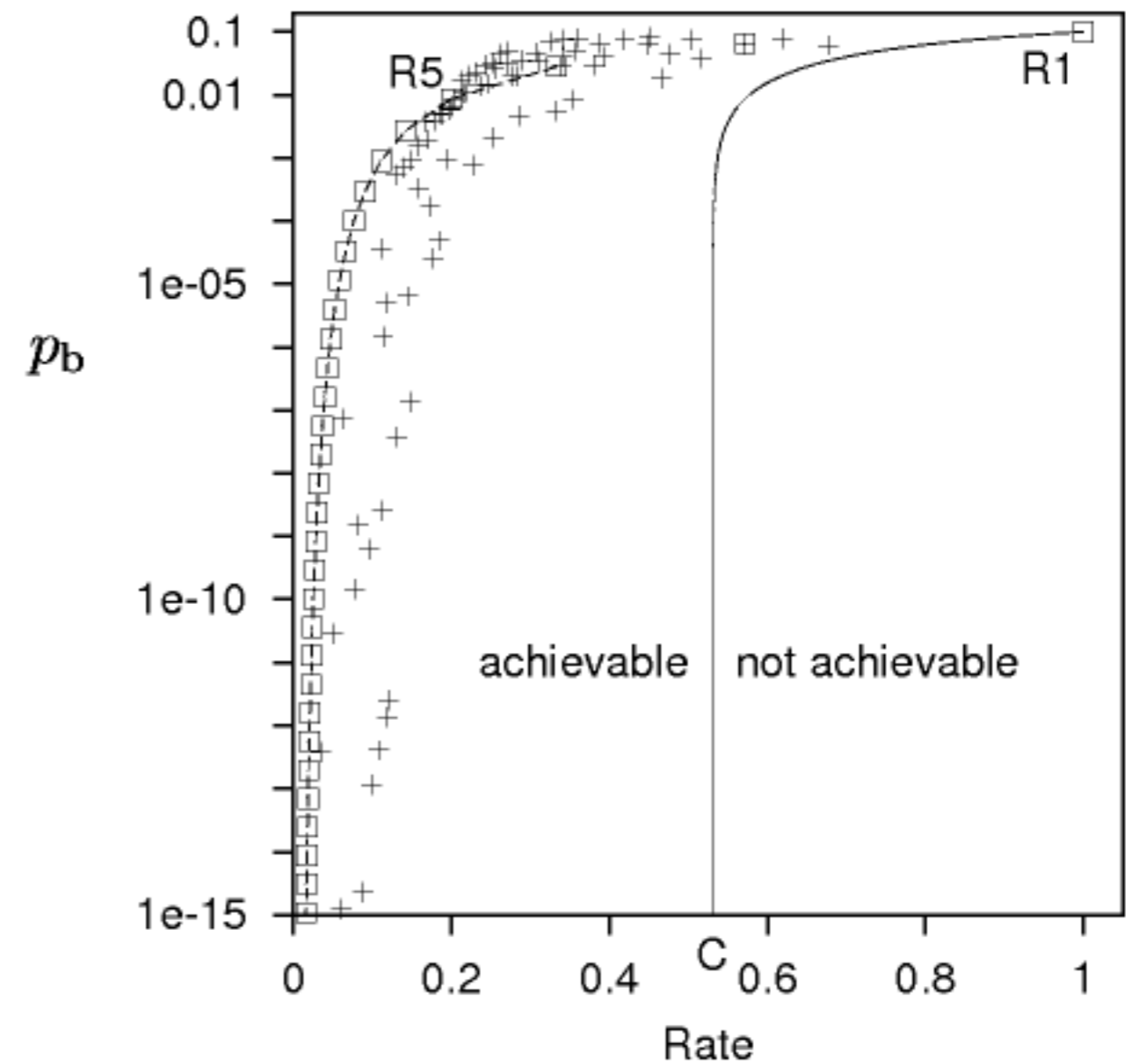
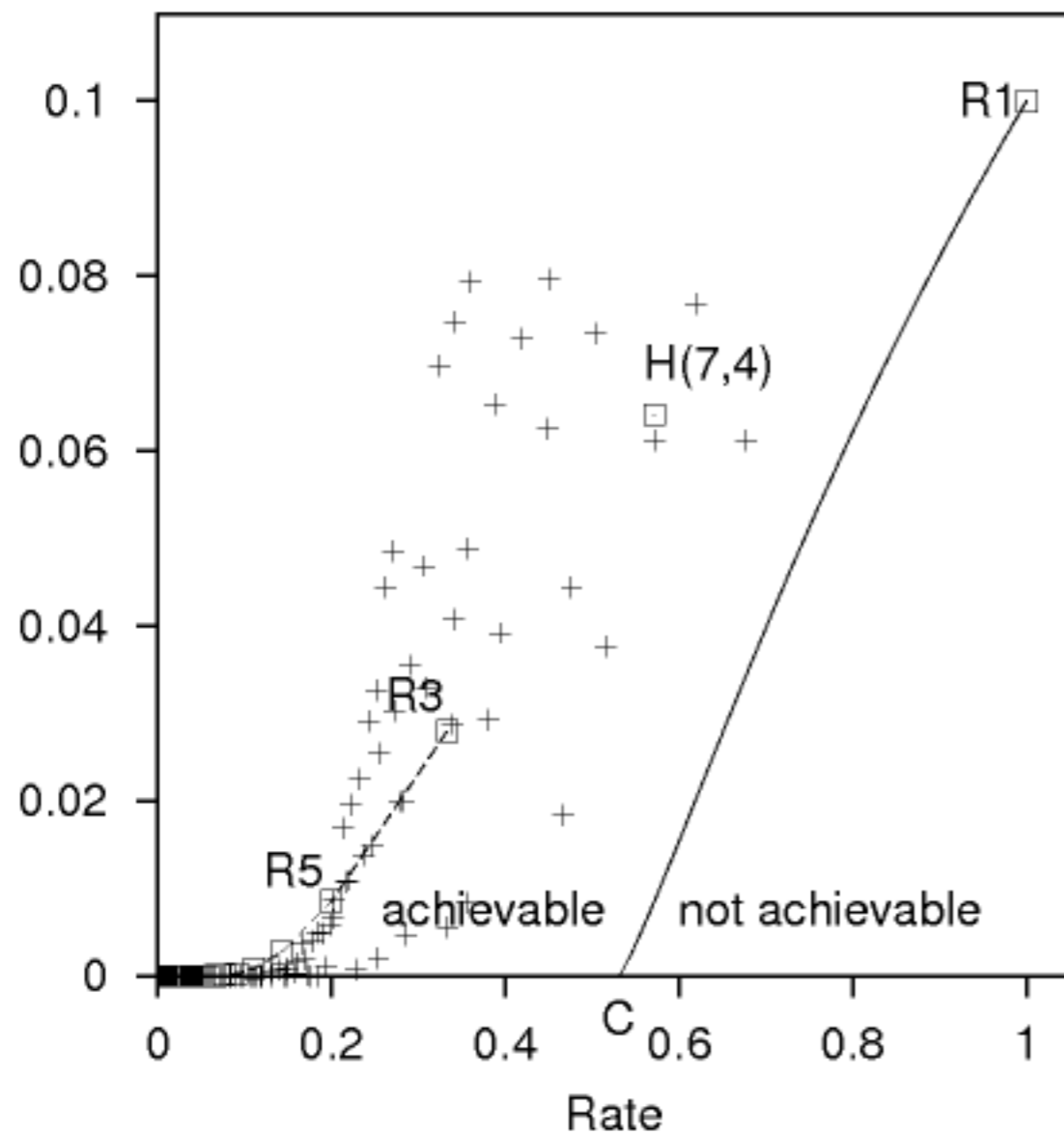
Performance of repetition codes





What's achievable?

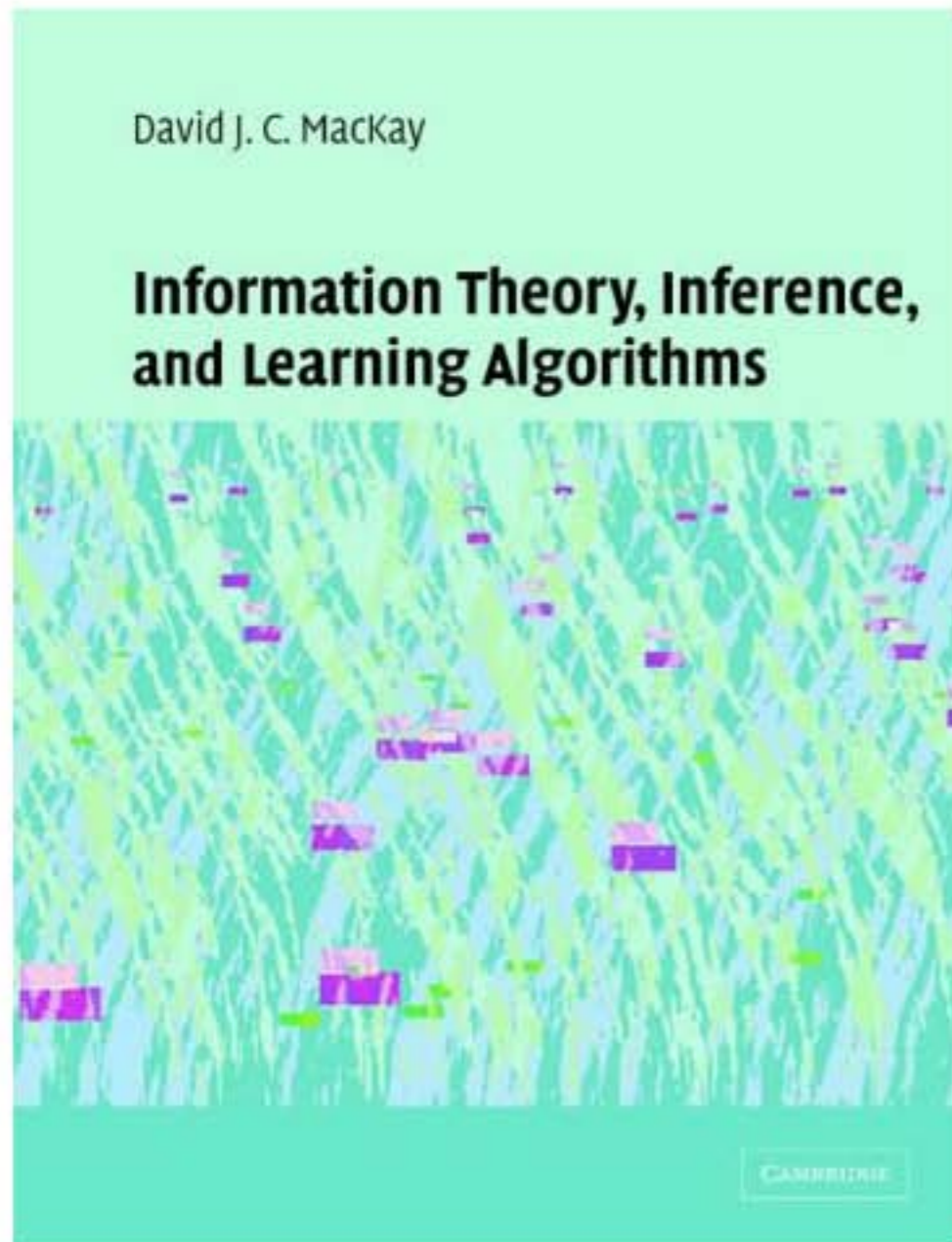
Shannon's noisy-channel coding theorem



$$C_{\text{BSC}}(f) = 1 - H_2(f)$$

$$H_2(f) = f \log_2 \frac{1}{f} + (1 - f) \log_2 \frac{1}{1 - f}$$

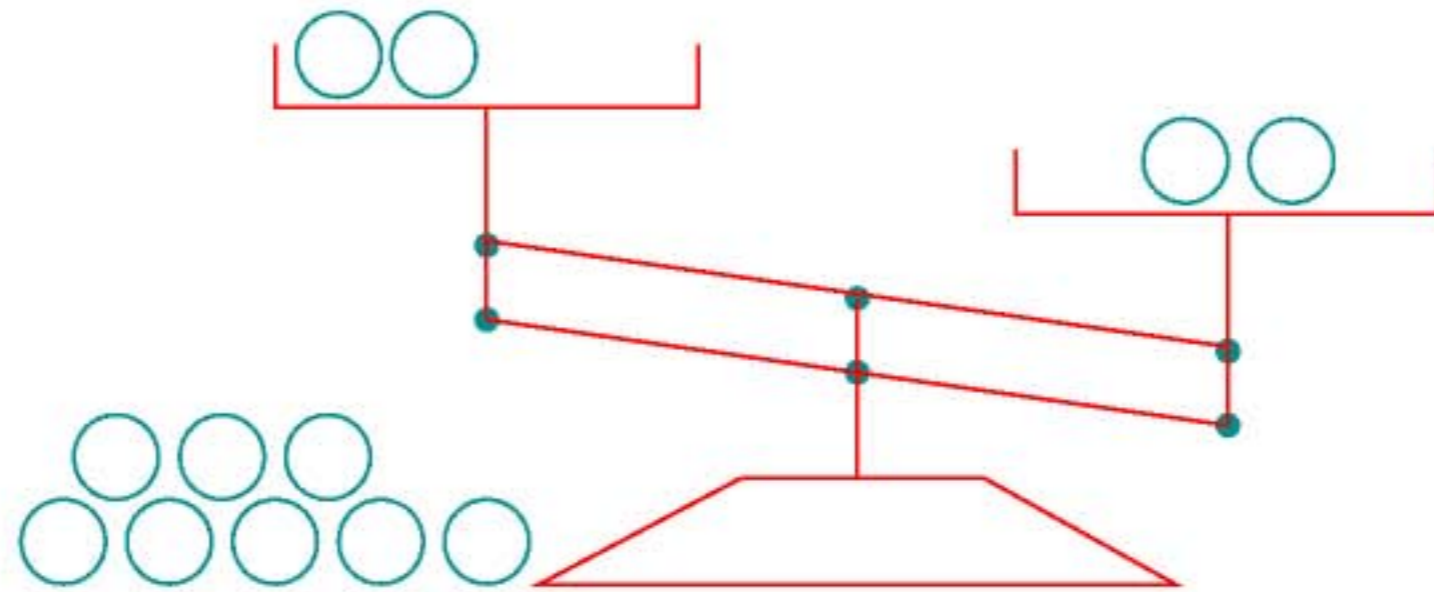
Outline of next lecture



● Information Theory

- Source coding (Data compression)
- Noisy-channel coding
 - ▶ the theorem
 - ▶ state-of-the art error-correcting codes

The weighing problem



You are given 12 balls, all equal in weight except for one that is either heavier or lighter.

Design a strategy to determine
which is the odd ball

and whether it is heavier or lighter,

in as few uses of the balance as possible.

Another book

Sustainable Energy –
without the hot air

David JC MacKay



“THIS BOOK IS A
TOUR DE FORCE ...
AS A WORK OF
POPULAR SCIENCE
IT IS EXEMPLARY”
THE ECONOMIST

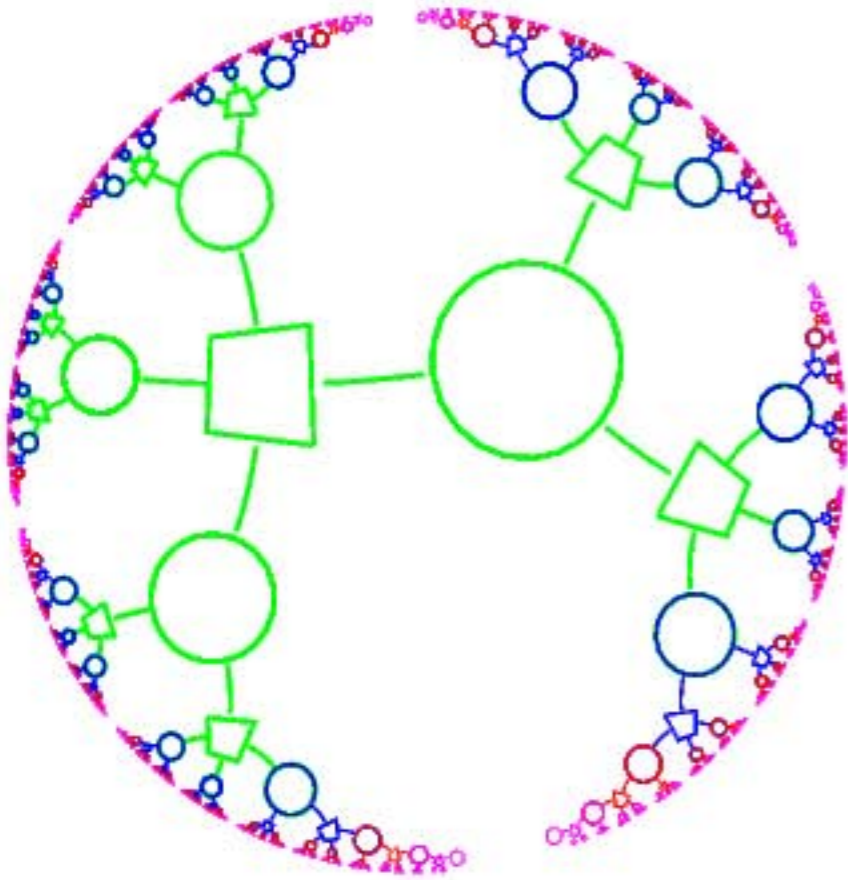
“THIS IS TO
ENERGY AND CLIMATE
WHAT FREAKONOMICS
IS TO ECONOMICS.”
CORY DOCTOROW,
BOINGBOING.NET



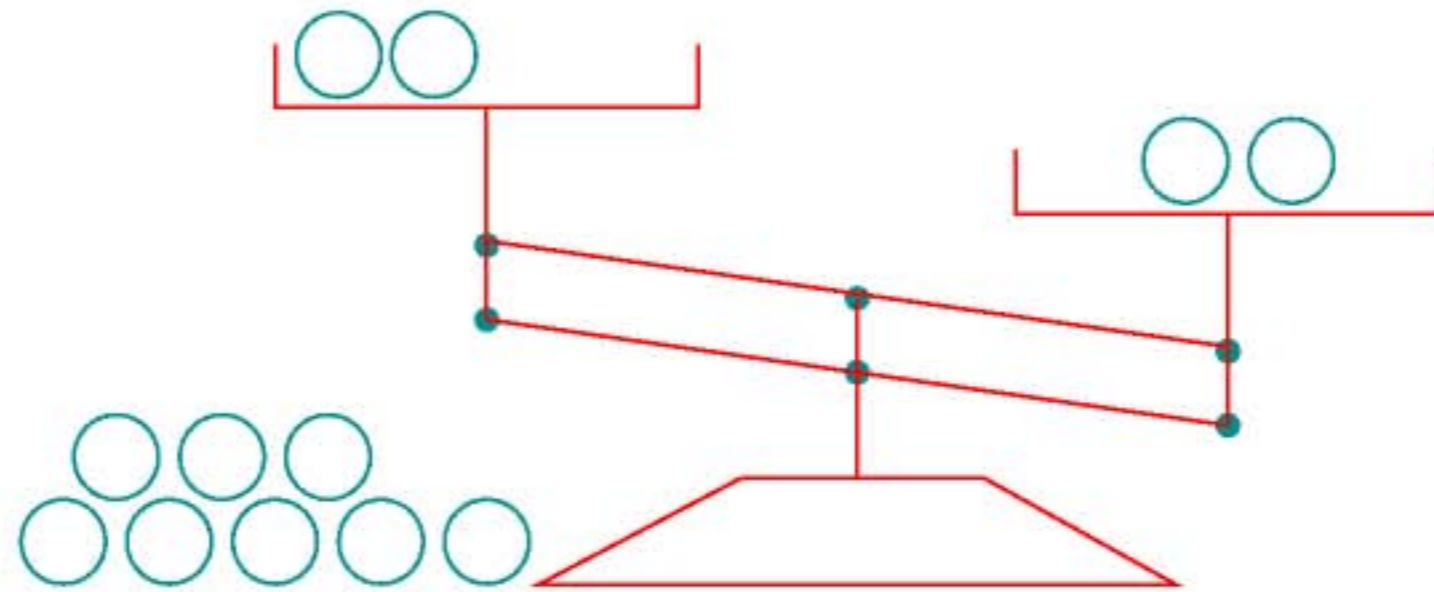
**SUSTAINABLE ENERGY –
WITHOUT THE HOT AIR**

David JC MacKay

Information theory II



The weighing problem



You are given 12 balls, all equal in weight except for one that is either heavier or lighter.

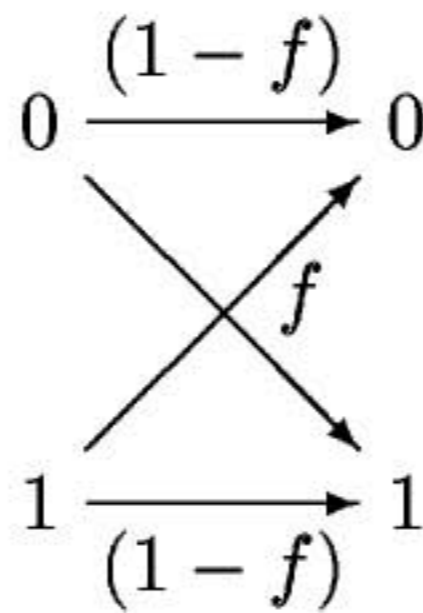
Design a strategy to determine
which is the odd ball

and whether it is heavier or lighter,

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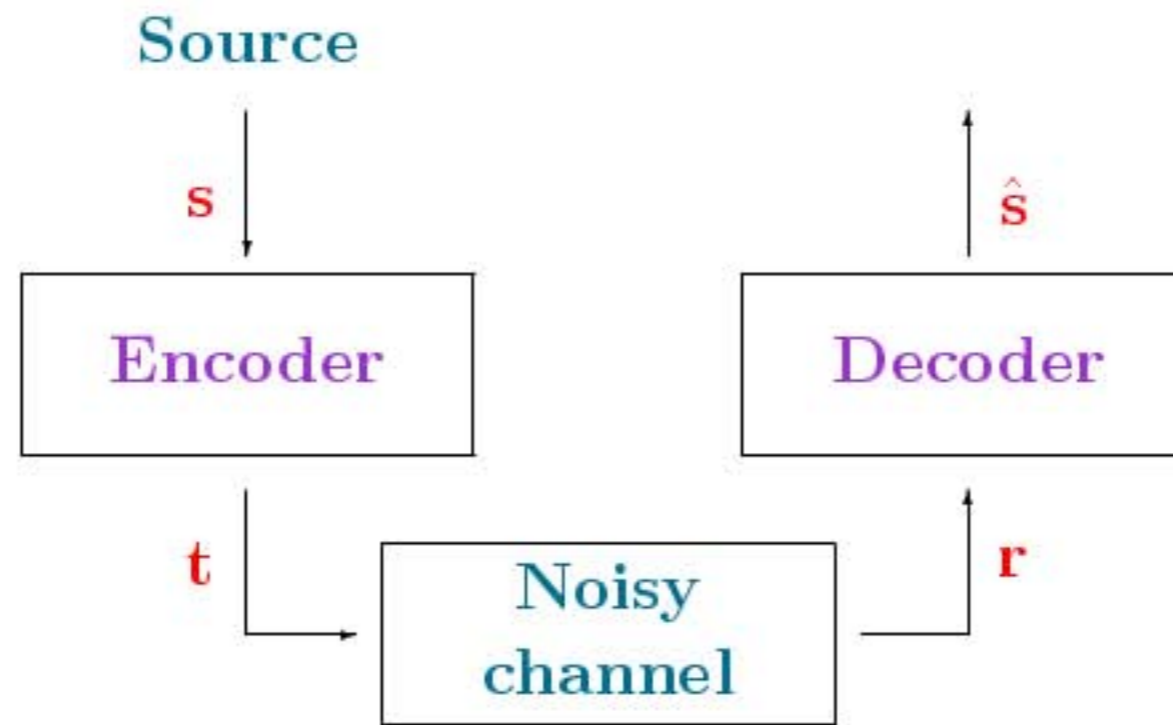
Purpose: reliable communication over unreliable channels

eg, Binary symmetric channel



$$f = 0.1$$

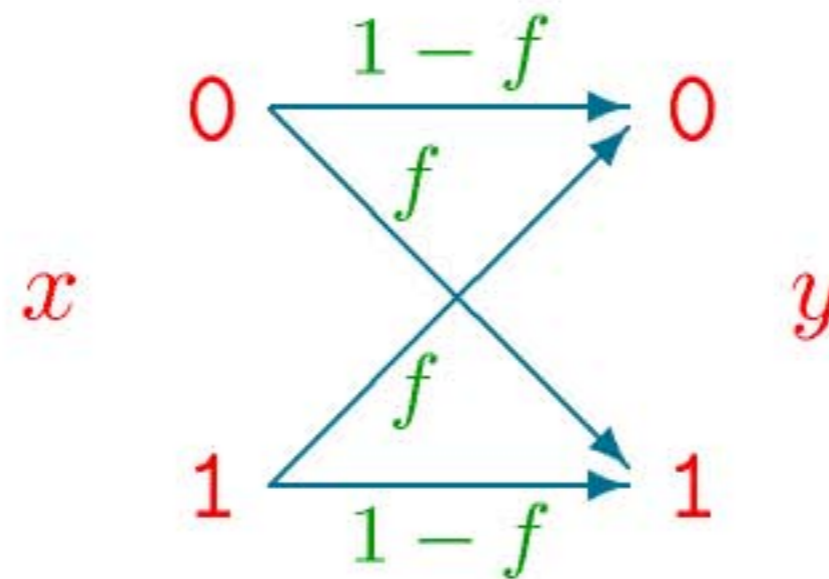
System solution



adds **redundancy**

does **inference**

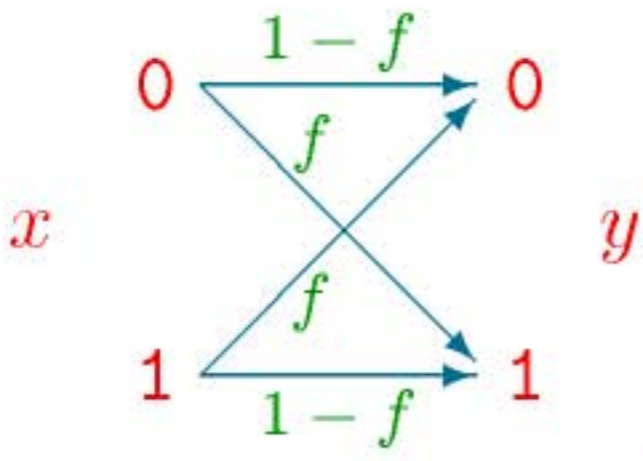
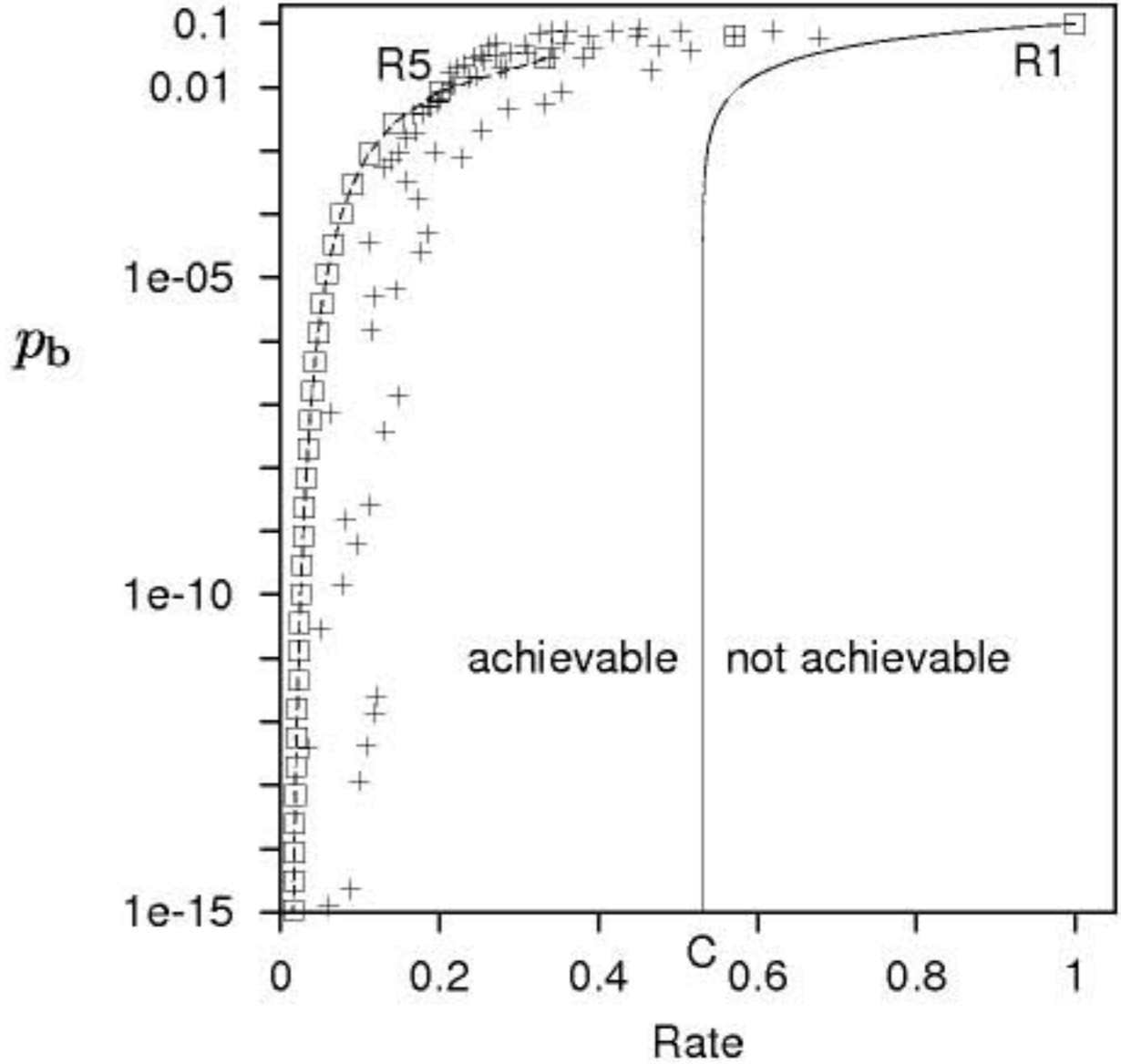
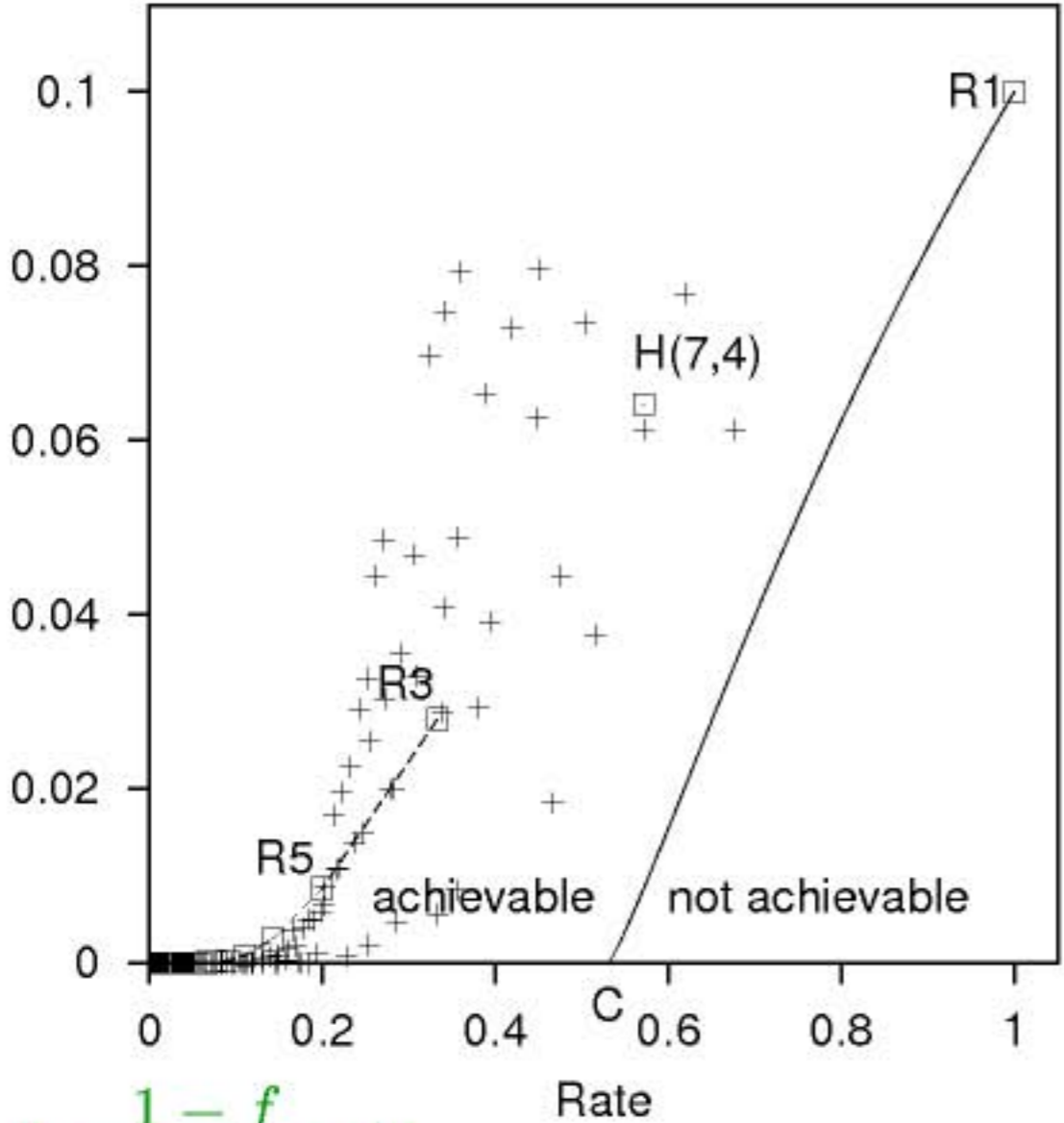
The binary symmetric channel



● Repetition codes

● (7,4) Hamming code

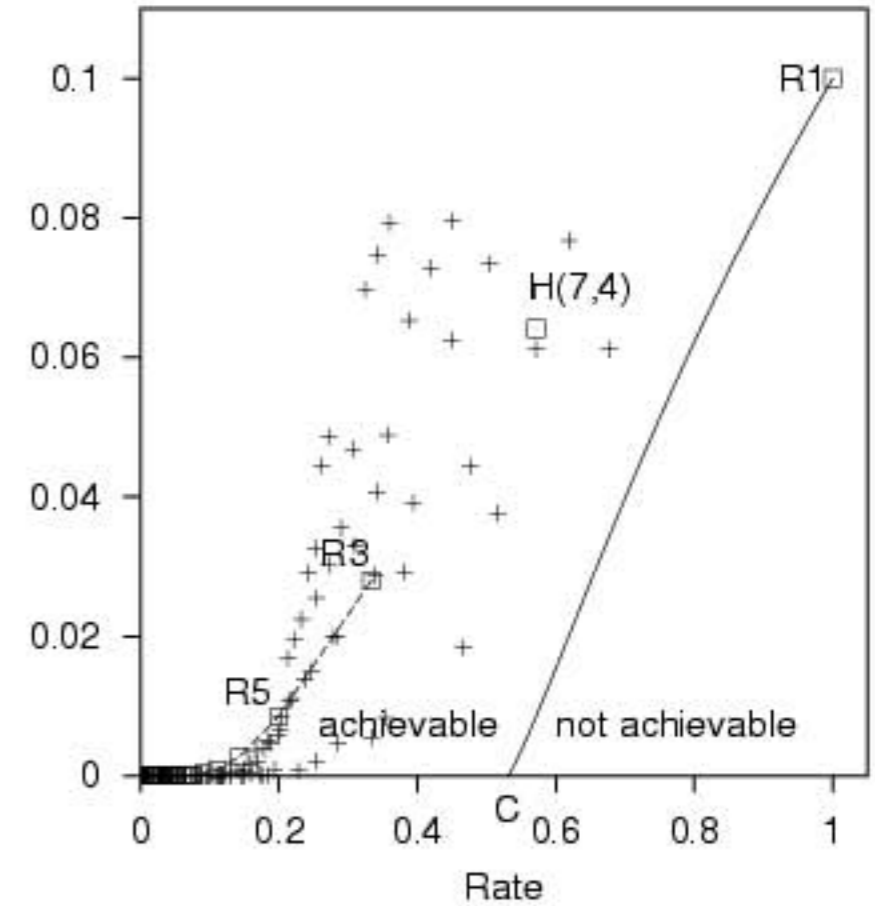
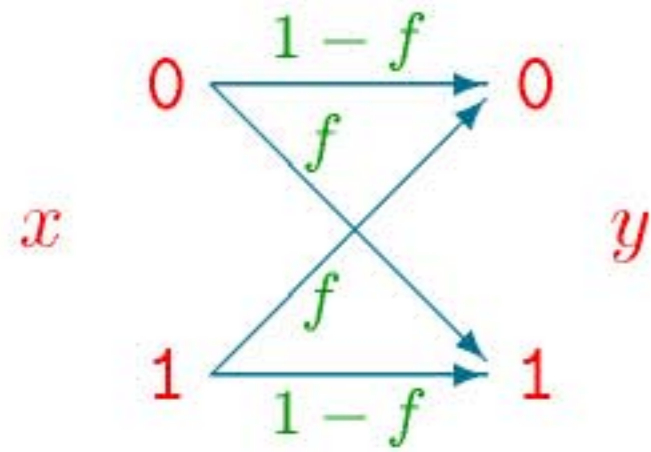
Shannon's noisy-channel coding theorem



$$C_{\text{BSC}}(f) = 1 - H_2(f)$$

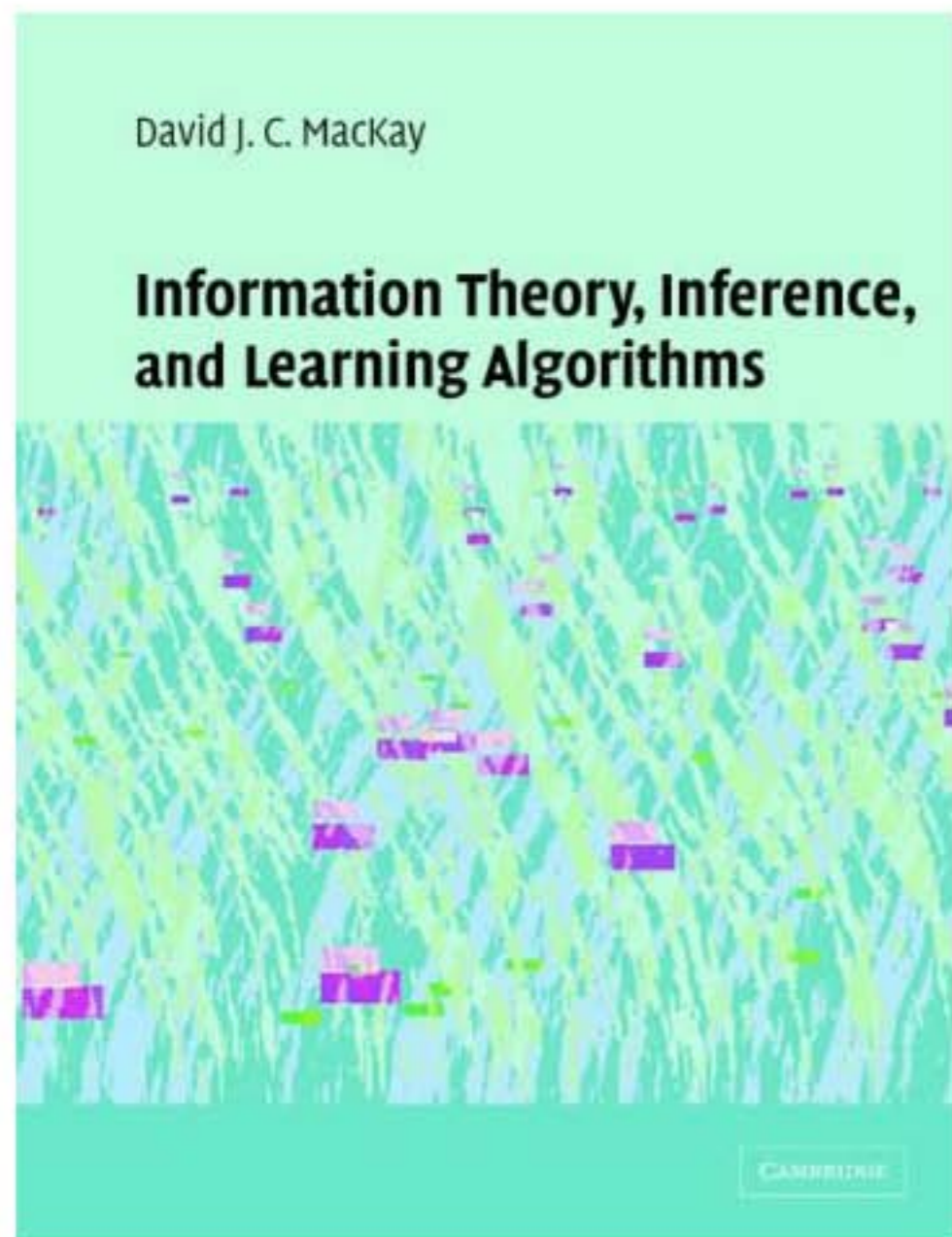
$$H_2(f) = f \log_2 \frac{1}{f} + (1 - f) \log_2 \frac{1}{1 - f}$$

Shannon's noisy channel coding theorem



For any channel:
Reliable (virtually error-free) communication is possible
at rates up to C

Outline



- Source coding (Data compression)
 - key ideas
 - optimal symbol codes
 - arithmetic coding
- Noisy-channel coding
 - the theorem
 - state-of-the art error-correcting codes
- Lecture notes - Chapters 4, 5, 6, 47
 - Cambridge University Press
 - 640 pages, 35 pounds
 - Also available **free online**

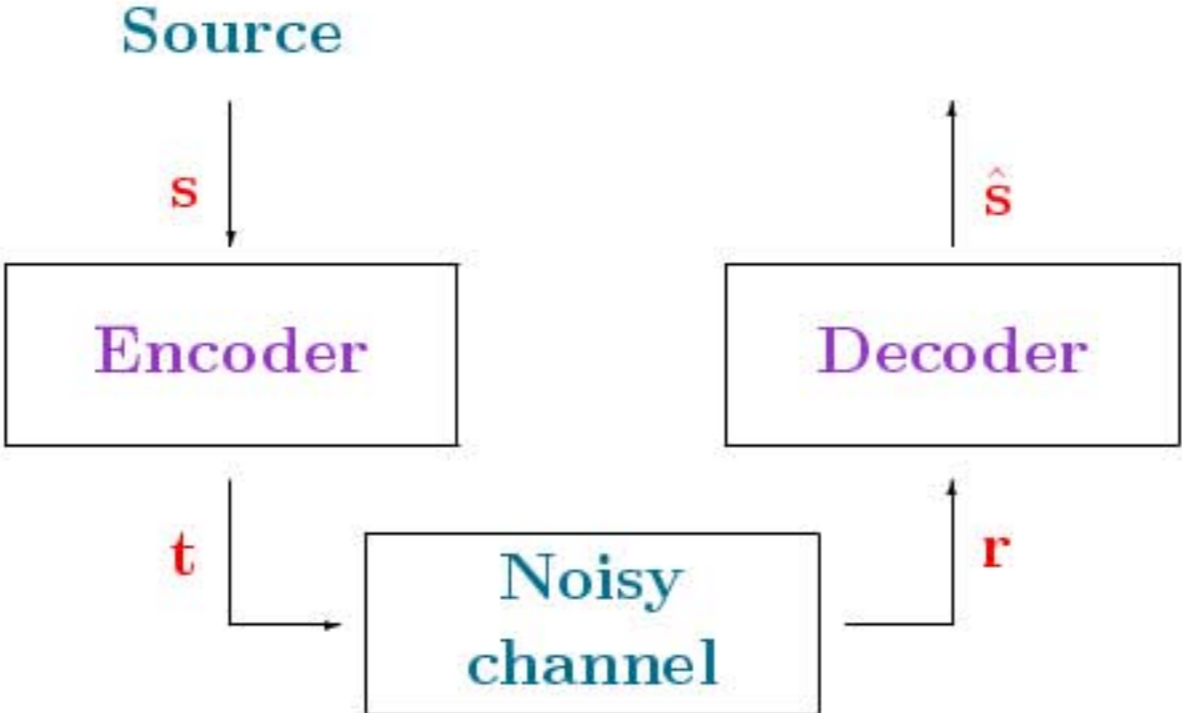
www.inference.phy.cam.ac.uk/mackay/itila/

Emma Woodh*use, hands*me, clever* and rich,*with a
comfortab*le home an* happy di*position,*seemed to*unite som*
of the b*st bless*ngs of e*istence;*and had *ived nea*ly
twenty *ne year* in the*world w*th very*little *o distr*ss
or vex*her. *he was*the yo*ngest *f the *wo dau*hters *f a
most *ffect*onate* indu*gent *ather* and *ad, i* cons*quenc*
of h*r si*ter'* mar*riage* bee* mis*ress*of h*s ho*se f*om a
ver* ea*ly *eri*d. *er *oth*r h*d d*ed *oo *ong*ago*for*her
to*ha*e *or* t*an*an*in*is*in*t *em*mb*an*e *f *er*ca*es*es*
a*d*h*r*p*a*e*h*d*b*e *u*p*i*d*b* *n*e*c*l*e*t*w*m*n*a*
g**e**e**, **h**h** **l**n**i**l**s**r**o**a**o**e**i*
a**c**n**S**e**y**s**d**s**a**r**e**n**
W***o***s***i***l***a***g***n***t***a***e***v***

Emma Woodhouse, handsome, clever, and rich, with a comfortable home and happy disposition, seemed to unite some of the best blessings of existence; and had lived nearly twenty one years in the world with very little to distress or vex her. She was the youngest of the two daughters of a most affectionate, indulgent father; and had, in consequence of her sister's marriage, been mistress of his house from a very early period. Her mother had died too long ago for her to have more than an indistinct remembrance of her caresses; and her place had been supplied by an excellent woman as governess, who had fallen little short of a mother in affection. Sixteen years had Miss Taylor been in Mr Woodhouse's family, less as a governess than a friend, very

Emma Woodh*use, hands*me, clever* and rich,*with a
comfortab*e home an* happy di*position,*seemed to*unite som*
of the b*st bless*ngs of e*istence;*and had *ived nea*ly
twenty *ne year* in the*world w*th very*little *o distr*ss
or vex*her. *he was*the yo*ngest *f the *wo dau*hters *f a
most *ffect*onate* indu*gent *ather* and *ad, i* cons*quenc*
of h*r si*ter'* mar*iage* bee* mis*ress*of h*s ho*se f*om a
ver* ea*ly *eri*d. *er *oth*r h*d d*ed *oo *ong*ago*for*her
to*ha*e *or* t*an*an*in*is*in*t *em*mb*an*e *f *er*ca*es*es*
a*d*h*r*p*a*e*h*d*b*e* *u*p*i*d*b* *n*e*c*l*e*t*w*m*n*a*
g**e**e**, **h**h** **l**n**i**l**s**r**o**a**o**e**i*
a**c**n**S**e**y**s**d**s**a**r**e**n**
W***o***s***i***l***a***g***n***t***a***e***v***

Channel coding



A simple redundant source - a bent coin

```
000000000000000000001001001000000000000000000000000010000010000101000000
0001010000000000000000000000000000000000000000000001000000000010000000000000100000010
000010000000000000000000100000000000000100000100000010000000000000000000000000
00001000000010000000000000100000000000000000000000000000000000000000000000000000
00000000100010000000100000001001000000000000000000000000000000000000000000000000
0001001000001010100000001100000100000000000000000000000000000000000000000000000000
0000000000000000000000000000000000000000000000000001010000000100000000000000000000100010100001
10000000000100000000000000000000000000000000000000000000000000000000000000000000000000000
010000000000000000100000000000000000000000000000000000000000000000000000000000000000000000000
10100010100000000000000000000000000000000000000000000000000000000000000000000000000000000000000
000000000000001000000100000100000100100000000000000000000000000000000000000000000000000000000
0000000000000000000000000000000000000000000000000000011000000001010000000000000000000000000000000
0001100101010000000000000000000000000000000000000000000000000000000000000000000000000000000000000
10000000100000000101100000000100000000000000000000000000000000000000000000000000000000000000000
1000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000
0000100010000000100000000000000000000000000000000000000000000000000000000000000000000000000000000
1101000001001000000000000000000000000000000000000000000000000000000000000000000000000000000000000
10000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000
0000000000001000000000000000000000000000000000000000000000000000000000000000000000000000000000000
0000000000001000000000000000000000000000000000000000000000000000000000000000000000000000000000000
1010001001000100000000000000000000000000000000000000000000000000000000000000000000000000000000000
0100100000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000
0000000010001000000000000000000000000000000000000000000000000000000000000000000000000000000000000
0000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000
0000010000000000000000000000000000000000000000000000000000000000000000000000000000000000000000000
```

$$p_1 = 0.1$$

How to compress a redundant file?

```
00000000000000000000000100100100000000000000000000000000000100000100001010000000
000101000000000000000000000000000000000000000000000000000000000000000000000000000110
000010000000000000000000000001000000000000000000000000000000000000000000000000000000
000010000000010000000000000001000000000000000000000000000000000000000000000000000000
0000000010001000000001000000010010000000000000000000000000000000000000000000000000000
0001001000001010100000001100000100000000000000000000000000000000000000000000000000000
0000000000000000000000000000000000000000000000000000000000000000000000000000000000001
1000000000010000000000000000000000000000000000000000000000000000000000000000000000000
0100000000000000000000000000000000000000000000000000000000000000000000000000000000000
1010000101000000000000000000000000000000000000000000000000000000000000000000000000000
```

e.g., $N = 1000$ tosses of a bent coin with $p_1 = 0.1$

- How to measure information content?
 - How much compression should we expect is possible?

How to measure information content?

Claims: 1. The *Shannon information content* of an outcome

$$h(x = a_i) = \log_2 \frac{1}{P(x = a_i)}$$

is a sensible measure of information content.

2. The *entropy*

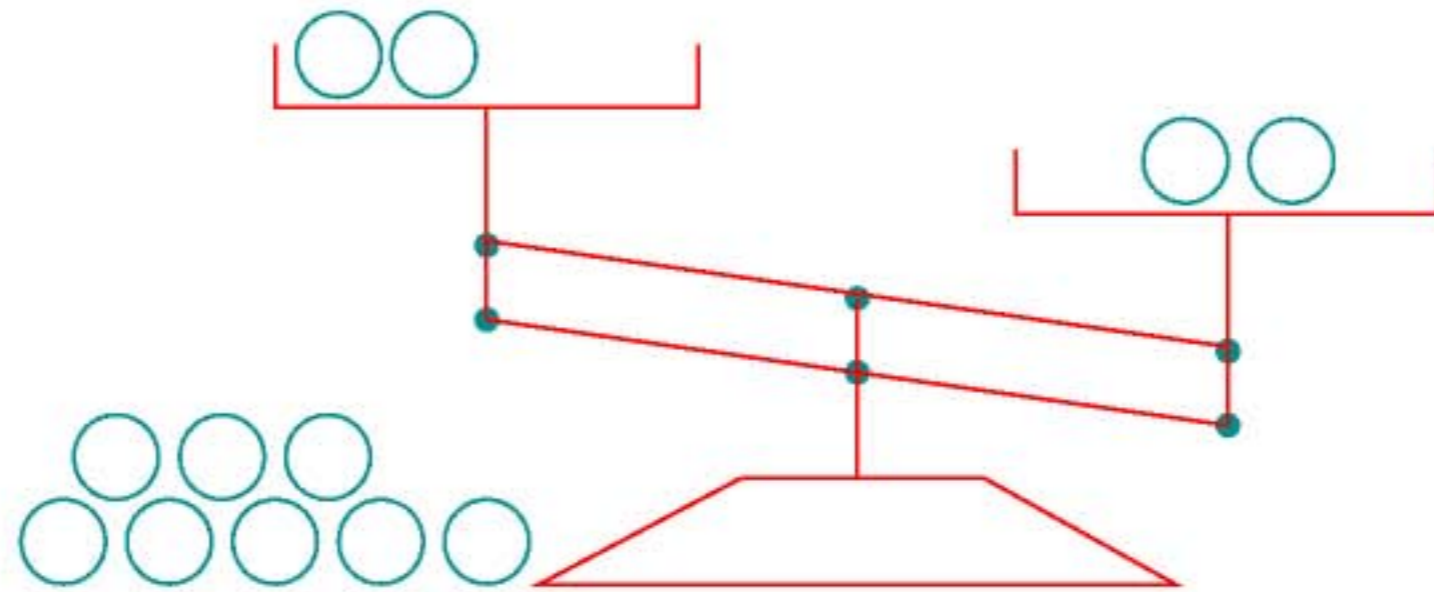
$$H(X) = \sum_x P(x) \log_2 \frac{1}{P(x)}$$

is a sensible measure of expected information content.

(sketch $h(p)$)

Point out additivity of h .

The weighing problem

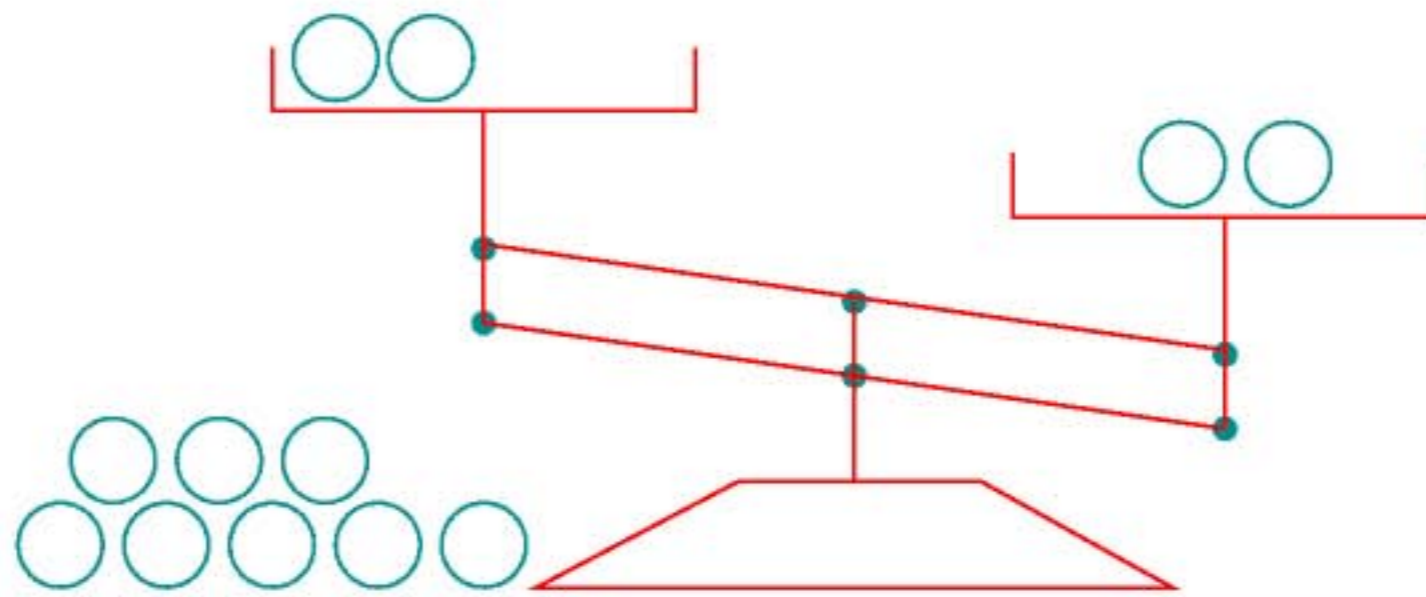


You are given 12 balls, all equal in weight except for one that is either heavier or lighter.

Design a strategy to determine
which is the odd ball

and whether it is heavier or lighter,

in as few uses of the balance as possible.



My strategy will find the odd ball **and** whether it is heavier or lighter in at most **13** uses of the balance.

12

11

10

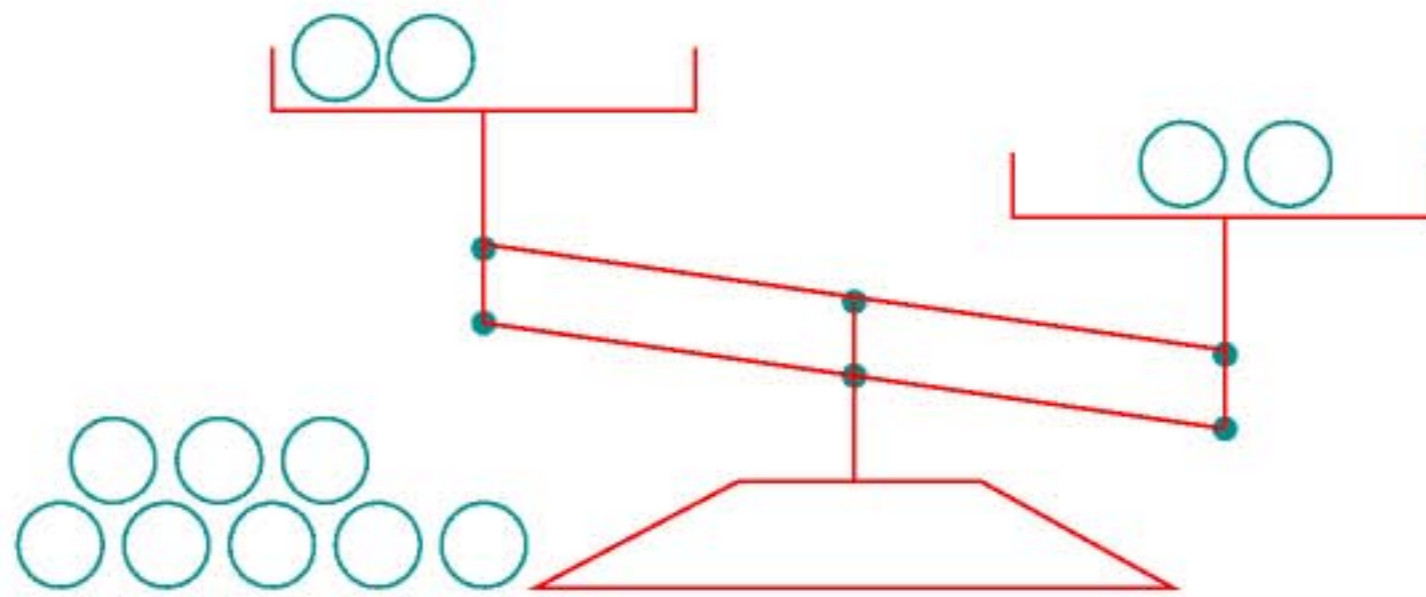
9

8

7

6

5



My strategy's first weighing is

6 v 6

5 v 5

4 v 4

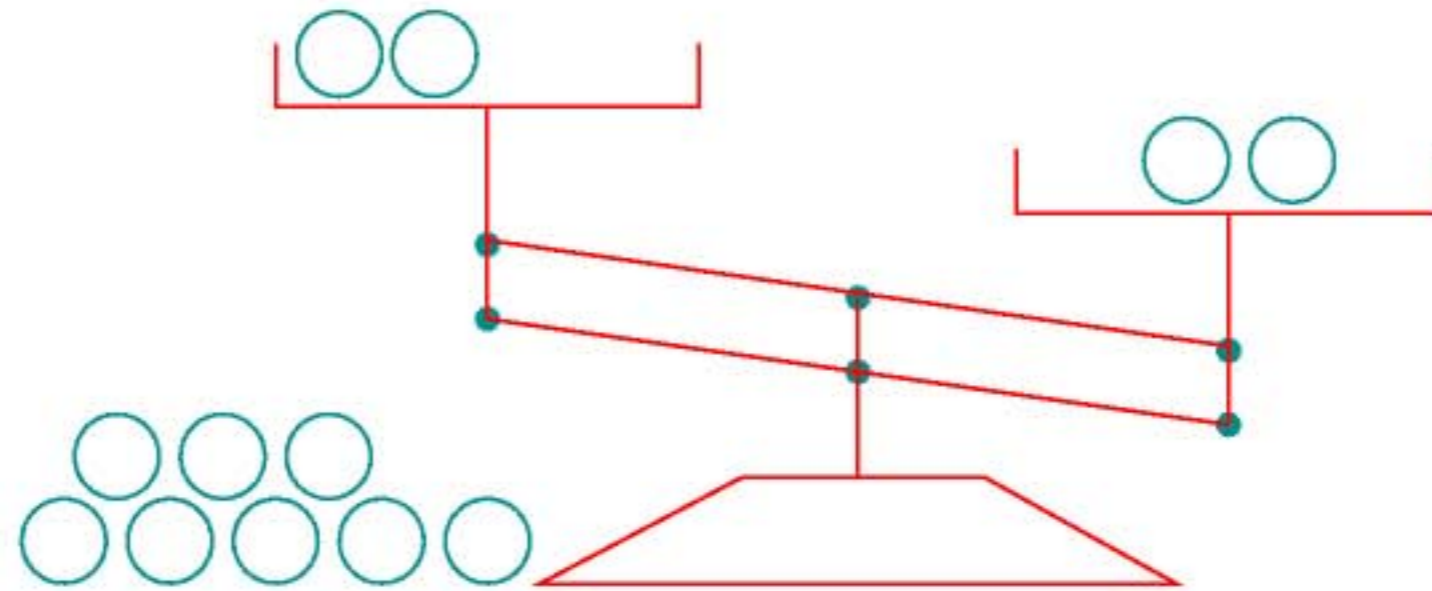
3 v 3

2 v 2

1 v 1

Testing 'Shannon information content' claims

● Weighing problem



● Shannon says

Most 'informative' experiment is the one with maximum entropy

How to measure information content?

Claims: 1. The *Shannon information content* of an outcome

$$h(x = a_i) = \log_2 \frac{1}{P(x = a_i)}$$

is a sensible measure of information content.

2. The *entropy*

$$H(X) = \sum_x P(x) \log_2 \frac{1}{P(x)}$$

is a sensible measure of expected information content.

3. *Source coding theorem* –

N outcomes from a source X can be compressed into roughly $NH(X)$ bits.

Symbol codes

i	a_i	p_i		
1	a	0.0575	a	■
2	b	0.0128	b	■
3	c	0.0263	c	■
4	d	0.0285	d	■
5	e	0.0913	e	■
6	f	0.0173	f	■
7	g	0.0133	g	■
8	h	0.0313	h	■
9	i	0.0599	i	■
10	j	0.0006	j	·
11	k	0.0084	k	■
12	l	0.0335	l	■
13	m	0.0235	m	■
14	n	0.0596	n	■
15	o	0.0689	o	■
16	p	0.0192	p	■
17	q	0.0008	q	·
18	r	0.0508	r	■
19	s	0.0567	s	■
20	t	0.0706	t	■
21	u	0.0334	u	■
22	v	0.0069	v	·
23	w	0.0119	w	■
24	x	0.0073	x	·
25	y	0.0164	y	■
26	z	0.0007	z	·
27	–	0.1928	–	■

The symbol-code supermarket

0	00	000	0000
			0001
		001	0010
			0011
	01	010	0100
			0101
		011	0110
			0111
1	10	100	1000
			1001
		101	1010
			1011
	11	110	1100
			1101
		111	1110
			1111

i	a_i	p_i
1	a	0.0575
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5	e	0.0913
6	f	0.0173
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8	h	0.0313
9	i	0.0599
10	j	0.0006
11	k	0.0084
12	l	0.0335
13	m	0.0235
14	n	0.0596
15	o	0.0689
16	p	0.0192
17	q	0.0008
18	r	0.0508
19	s	0.0567
20	t	0.0706
21	u	0.0334
22	v	0.0069
23	w	0.0119
24	x	0.0073
25	y	0.0164
26	z	0.0007
27	-	0.1928

a	■
b	■
c	■
d	■
e	■
f	■
g	■
h	■
i	■
j	■
k	■
l	■
m	■
n	■
o	■
p	■
q	■
r	■
s	■
t	■
u	■
v	■
w	■
x	■
y	■
z	■
-	■

a	0000
b	001000
c	00101
d	10000
e	1100
f	111000
g	001001
h	10001
i	1001
j	1101000000
k	1010000
l	11101
m	110101
n	0001
o	1011
p	111001
q	110100001
r	11011
s	0011
t	1111
u	10101
v	11010001
w	1101001
x	1010001
y	101001
z	1101000001
-	01

Symbol codes

The ideal codelengths l_i^* are the information contents

$$l_i^* = \log \frac{1}{p_i}$$

The optimal symbol code's expected length L satisfies

$$H(X) \leq L < H(X) + 1$$

The optimal symbol code is generated by the Huffman algorithm

- Does that wrap up compression?
 - What's wrong with optimal symbol codes?
- Arithmetic coding

The Guessing Game

Headline composed of

{ A, B, C, D, E, F, G, H, I, J, K, L, M, ..., Z, - }

Realistic compression

Optimal symbol code for

a	0.001	00000
b	0.001	00001
c	0.990	1
d	0.001	00010
e	0.001	00011
f	0.001	0100
g	0.001	0101
h	0.001	0110
i	0.001	0111
j	0.001	0010
k	0.001	0011

expected length	1.034
entropy	0.11401
length / entropy	9

Arithmetic coding

uniform bounded at_mouse at_cross

Colors: 28 Truncate: 6 At: < > 6 Slide: < > -2

Reset magnifn: 0 width: 1.0 centre: 0.5 0.0 1.0

english1 english2 jack Alphabet 3 adaptive 0.45 0.45 0.15 Alphabet 2 adaptive 0.5 0.5

Active Pushiness 1 ps Clean VariableL DeDisRel Reinstate Active Pushiness 1 ps

: ba 2 : 01 2

Quit Hide Show OneCanvas Dump Mousepad Help

Other uses for arithmetic coding

- Efficient writing

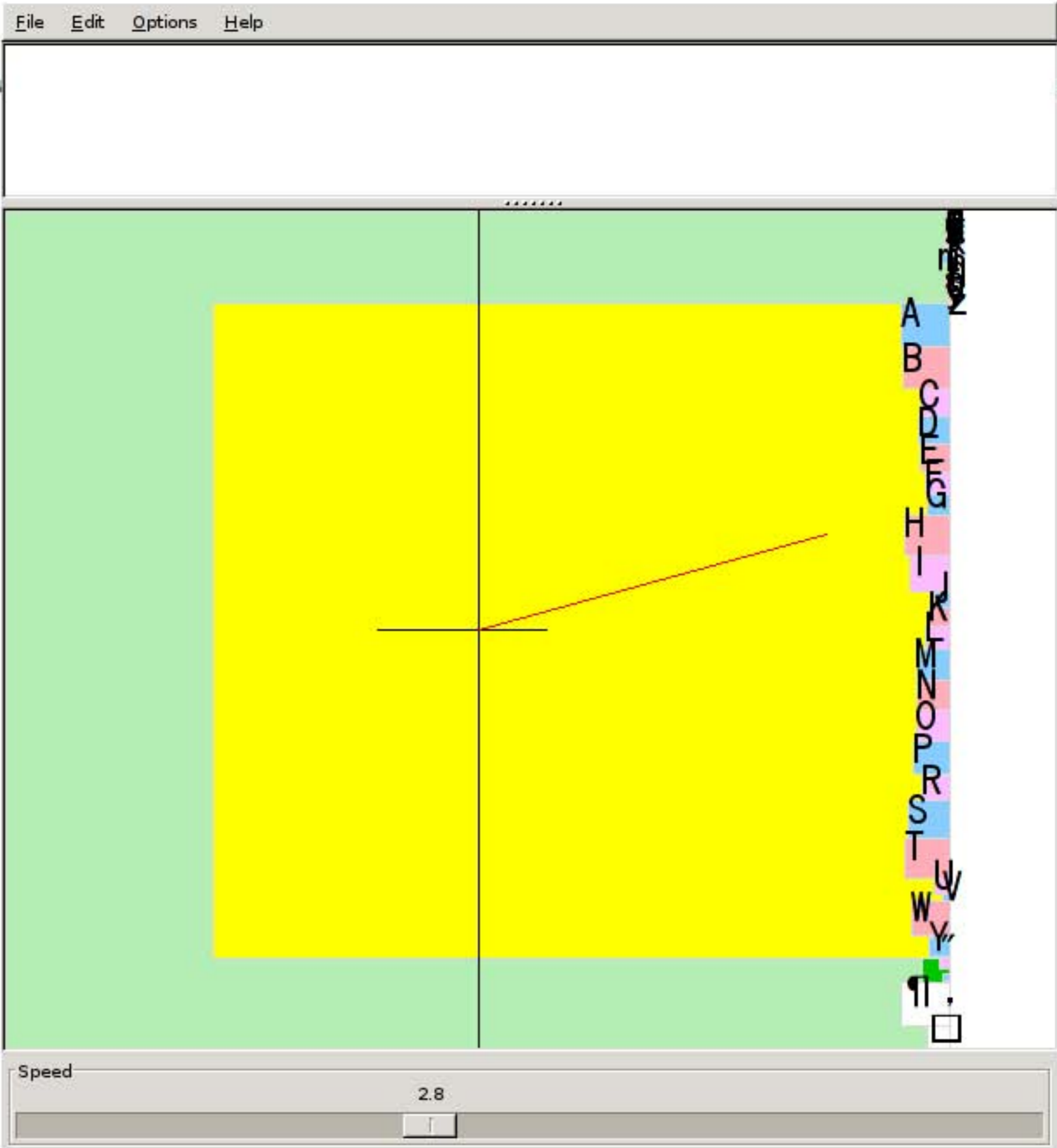
Compression:

Text \longrightarrow **Bit string**
(preferably short)

Writing:

Text \longleftarrow **Gesture**
(preferably brief)

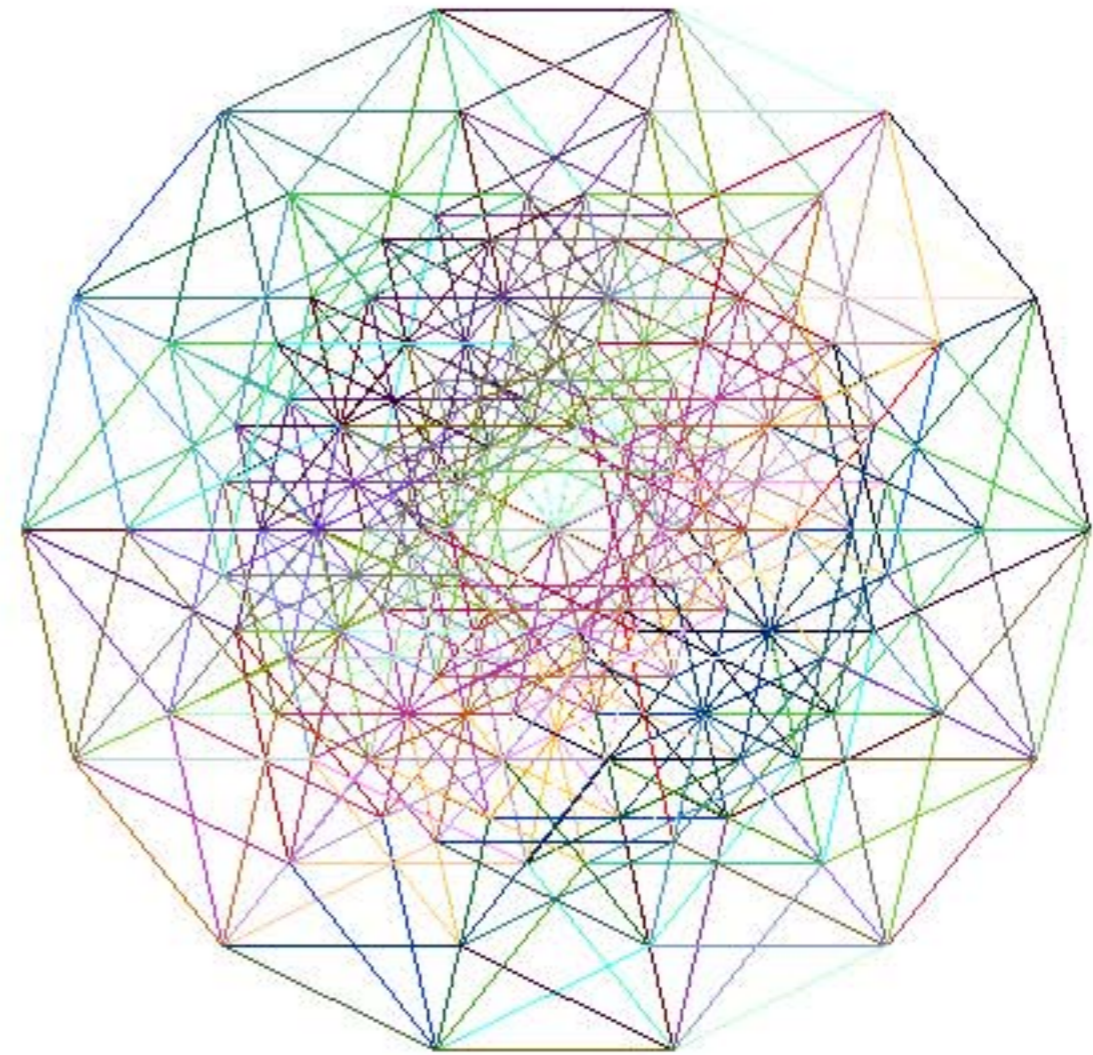
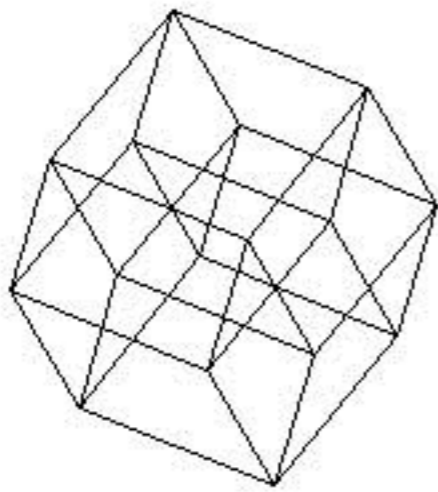
Dasher



Block codes

(7,4) Hamming Code

<u>s</u>	<u>t</u>	<u>s</u>	<u>t</u>	<u>s</u>	<u>t</u>	<u>s</u>	<u>t</u>
0000	0000 000	0100	0100 110	1000	1000 101	1100	1100 011
0001	0001 011	0101	0101 101	1001	1001 110	1101	1101 000
0010	0010 111	0110	0110 001	1010	1010 010	1110	1110 100
0011	0011 100	0111	0111 010	1011	1011 001	1111	1111 111



(7,4) Hamming Code

$$\mathbf{H} = \begin{bmatrix} 1 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{array}{l} \uparrow \\ M = 3 \\ \downarrow \end{array}$$

$\leftarrow N = 7 \rightarrow$

Valid transmissions \mathbf{t} satisfy

$$\mathbf{H} \mathbf{t} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \pmod{2}$$

(7,4) Hamming Code

$$\mathbf{H} = \begin{bmatrix} 1 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{array}{l} \uparrow \\ M = 3 \\ \downarrow \end{array}$$

$\leftarrow N = 7 \rightarrow$

Valid transmissions \mathbf{t} satisfy

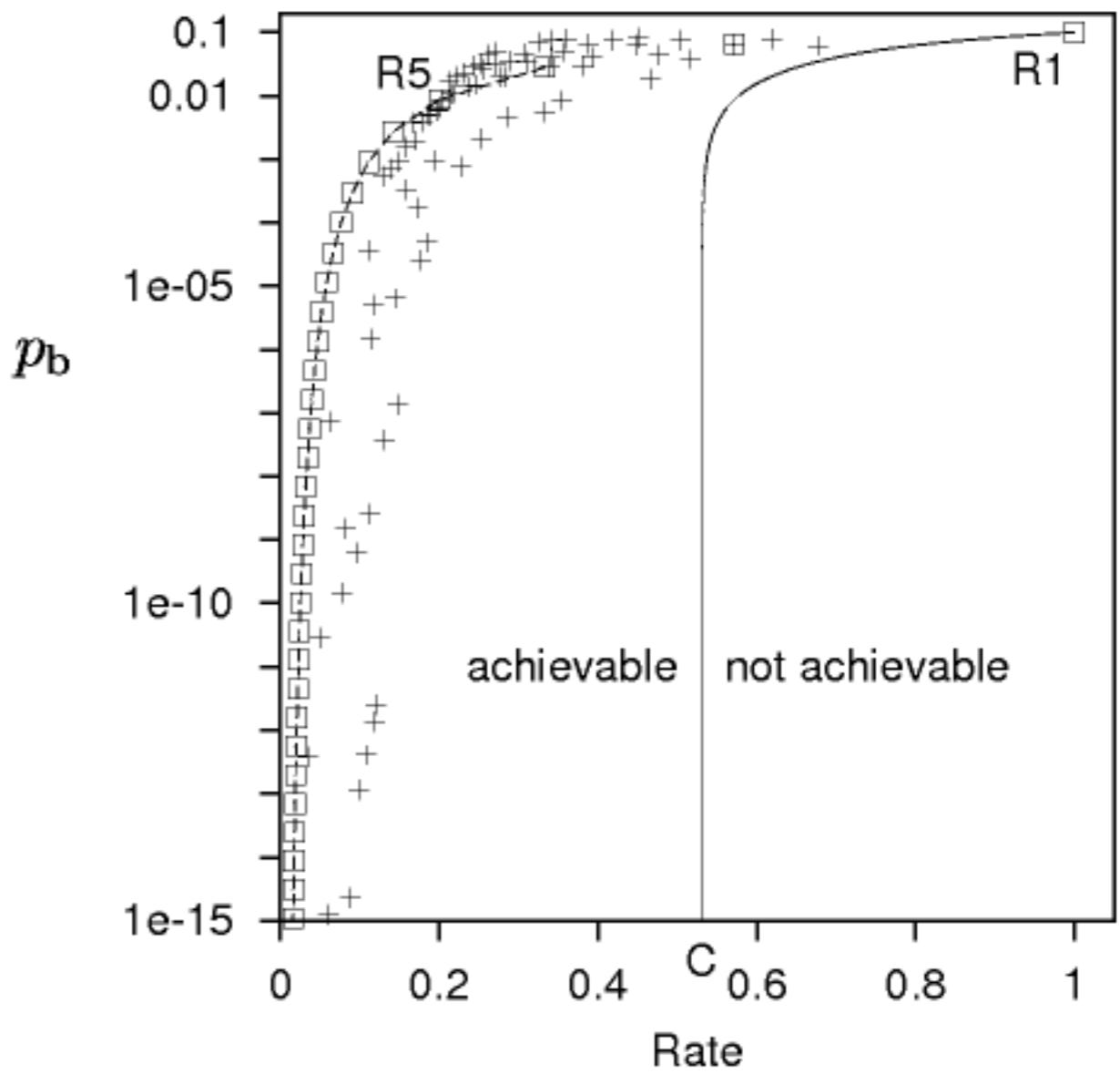
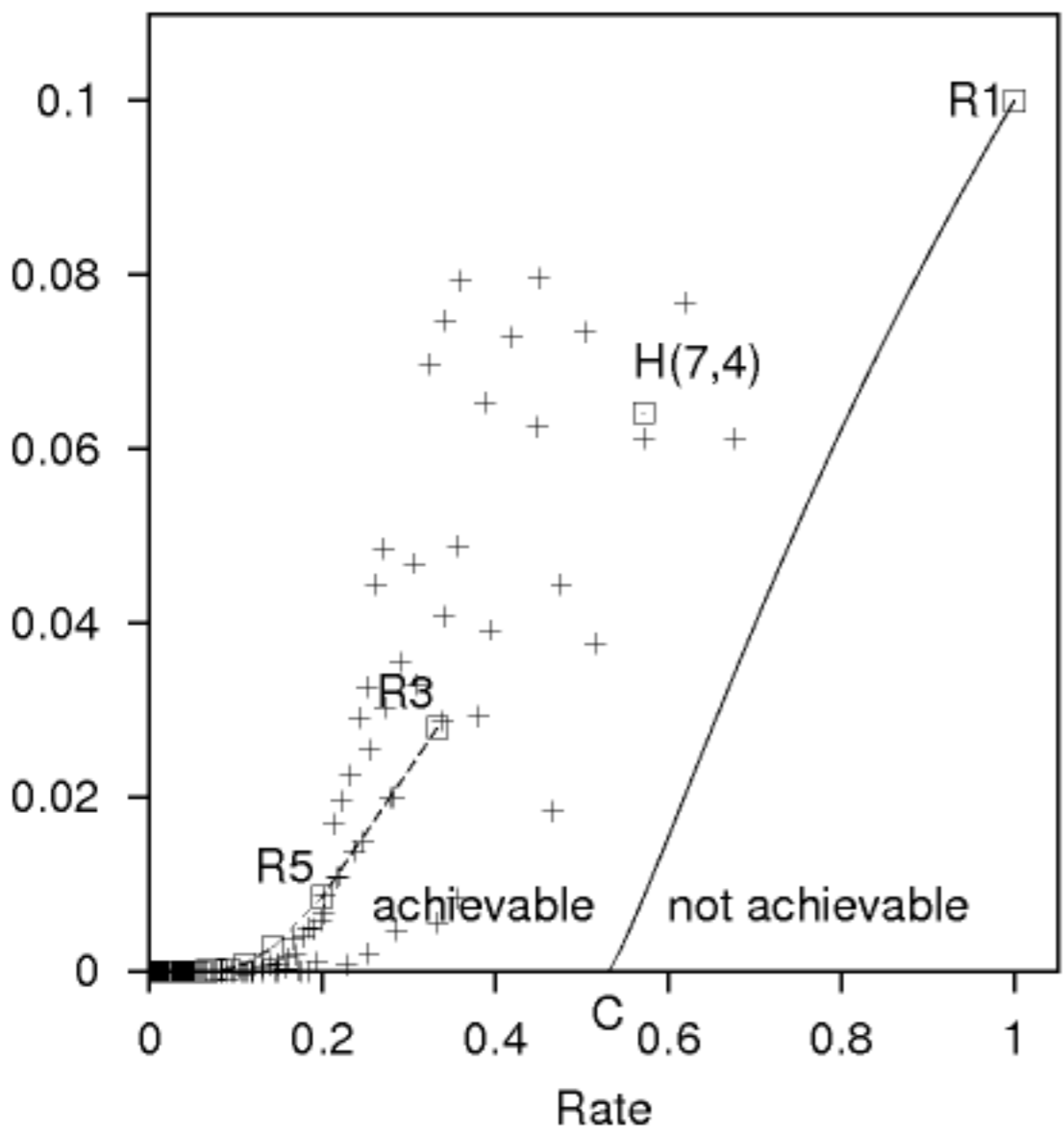
$$\mathbf{H} \mathbf{t} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \pmod{2}$$

Received signal $\mathbf{r} = \mathbf{t} + \mathbf{n}$

Syndrome $\mathbf{z} = \mathbf{H} \mathbf{r} = \mathbf{H} \mathbf{n}$.

Syndrome decoder $\mathbf{z} \longrightarrow \hat{\mathbf{n}}$.

Shannon's noisy-channel coding theorem



$$C_{\text{BSC}}(f) = 1 - H_2(f)$$

$$H_2(f) = f \log_2 \frac{1}{f} + (1 - f) \log_2 \frac{1}{1 - f}$$

How to prove good codes exist

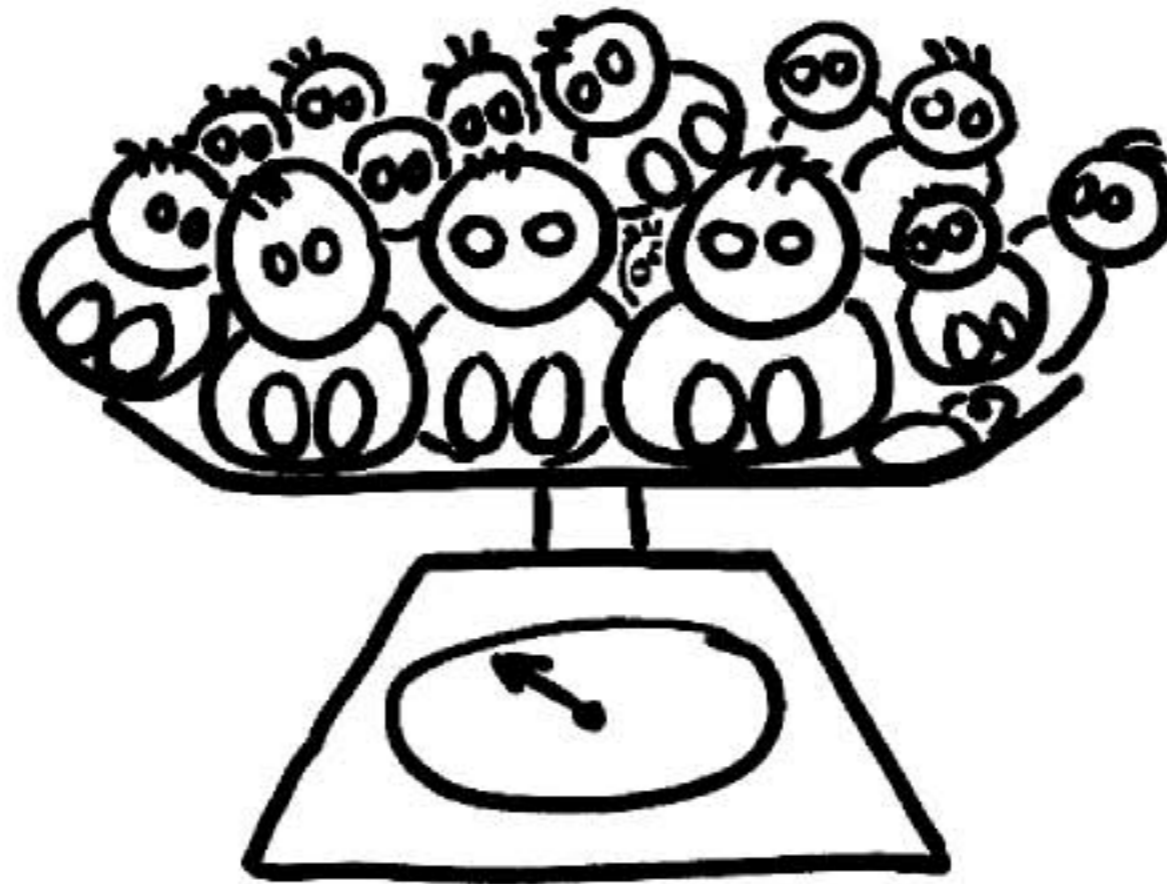
● Constructive proof

Given required $R < C$, and $\epsilon > 0$,

$$\mathbf{H} = \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 1 & 0 & & & \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 & \dots & \dots & \dots \\ 1 & 0 & 1 & 1 & 1 & 0 & 0 & & & \\ \vdots & & & \vdots & & \vdots & & \ddots & & \\ \vdots & & & \vdots & & \vdots & & & \ddots & \\ \vdots & & & \vdots & & \vdots & & & & \ddots \end{bmatrix}$$

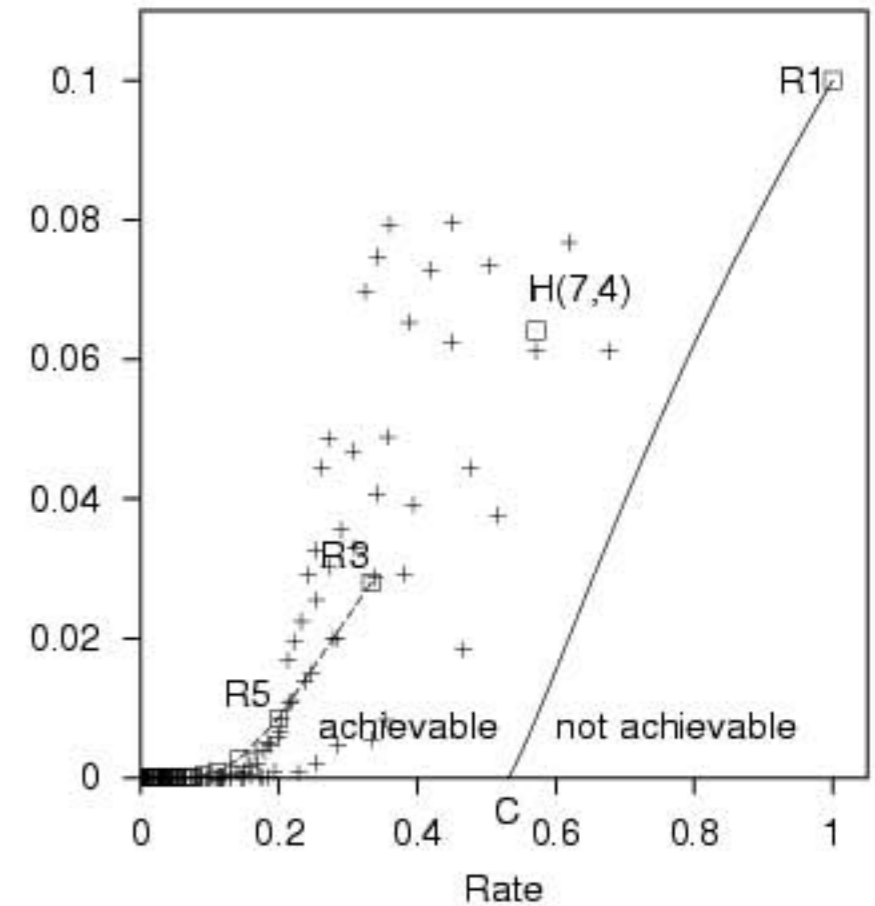
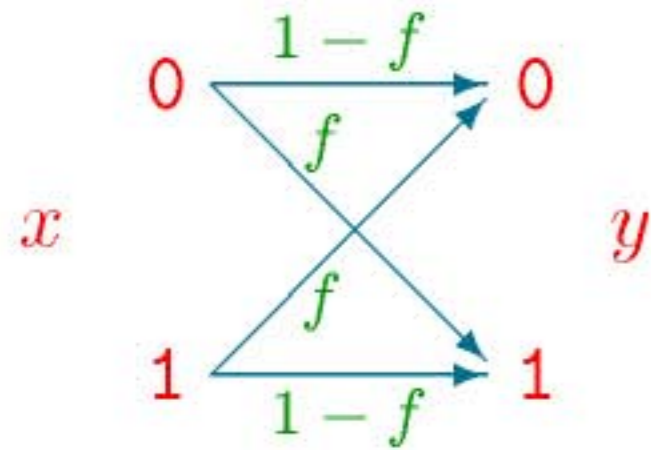
● Non-constructive proof

Shannon's way of proving malnutrition



If **average** weight
of all babies is $< \epsilon$,
there must be **(at least!)**
one baby with weight $< \epsilon$.

Shannon's noisy channel coding theorem

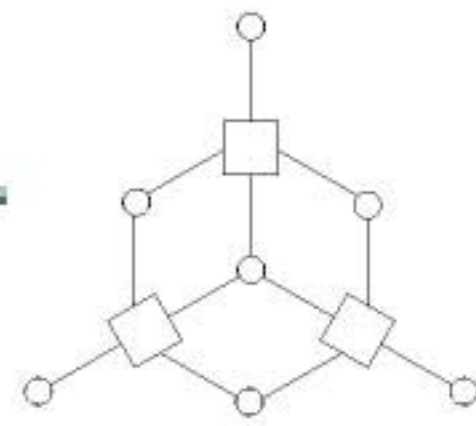


For any channel:
Reliable (virtually error-free) communication is possible
at rates up to C

Information theory Shannon, 1948

Coding theory Hamming, 1948; Reed-Solomon; Forney (Convolutional & concatenated codes)

Idea



- Decoding problems, such as

$$P_1(\mathbf{x}) = \frac{1}{Z_1} e^{\beta [x_1 x_2 x_3 x_5 + x_2 x_3 x_4 x_6 + x_1 x_3 x_4 x_7] + \sum_{n=1}^N b_n x_n}$$

look a bit like Boltzmann machines + Hopfield networks.... so

- Solve the decoding problem

$$\max_{\mathbf{x}} P_1(\mathbf{x})$$

using variational methods?

Electronics Letters, March 1995

Free energy minimisation algorithm for decoding and cryptanalysis

D.J.C. MacKay

Indexing terms: Decoding, Cryptography

An algorithm is derived for inferring a binary vector \mathbf{s} given noisy observations of $\mathbf{A}\mathbf{s}$ modulo 2, where \mathbf{A} is a binary matrix. The binary vector is replaced by a vector of probabilities, optimised by free energy minimisation. Experiments on the inference of the state of a linear feedback shift register indicate that this algorithm supersedes the Meier and Staffelbach polynomial algorithm.

Decoding error-correcting codes

- For which codes are **approximate message-passing methods**
effective?

Electronics Letters, August 1996

Near Shannon limit performance of low density parity check codes

D.J.C. MacKay and R.M. Neal

[Low Density Parity Check Codes: Gallager 1962]

Indexing terms: Probabilistic decoding, Error correction codes

The authors report the empirical performance of Gallager's low density parity check codes on Gaussian channels. They show that performance substantially better than that of standard convolutional and concatenated codes can be achieved; indeed the performance is almost as close to the Shannon limit as that of turbo codes

(7,4) Hamming Code - recap

$$\mathbf{H} = \begin{bmatrix} 1 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \uparrow \\ M \\ \downarrow \end{matrix}$$

$\leftarrow N=7 \rightarrow$

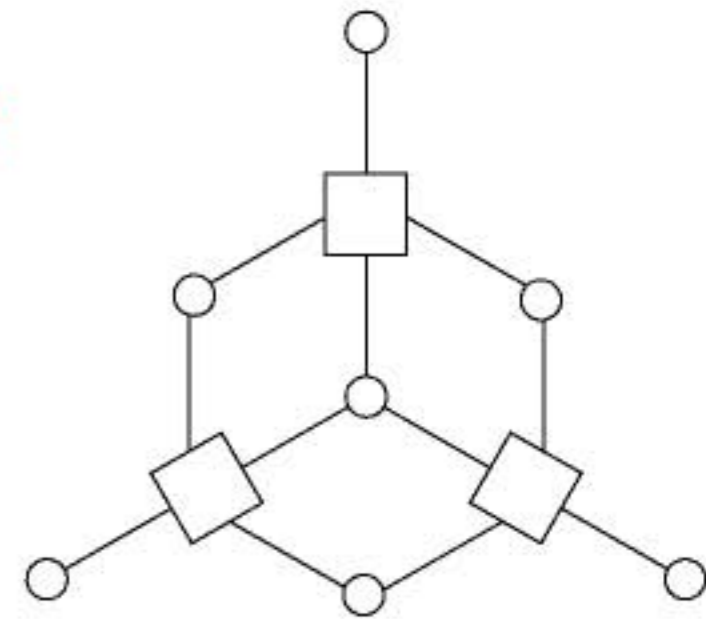
Valid transmissions \mathbf{t} satisfy

$$\mathbf{H} \mathbf{t} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \pmod{2}$$

Received signal $\mathbf{r} = \mathbf{t} + \mathbf{n}$

Syndrome $\mathbf{z} = \mathbf{H} \mathbf{r} = \mathbf{H} \mathbf{n}$.

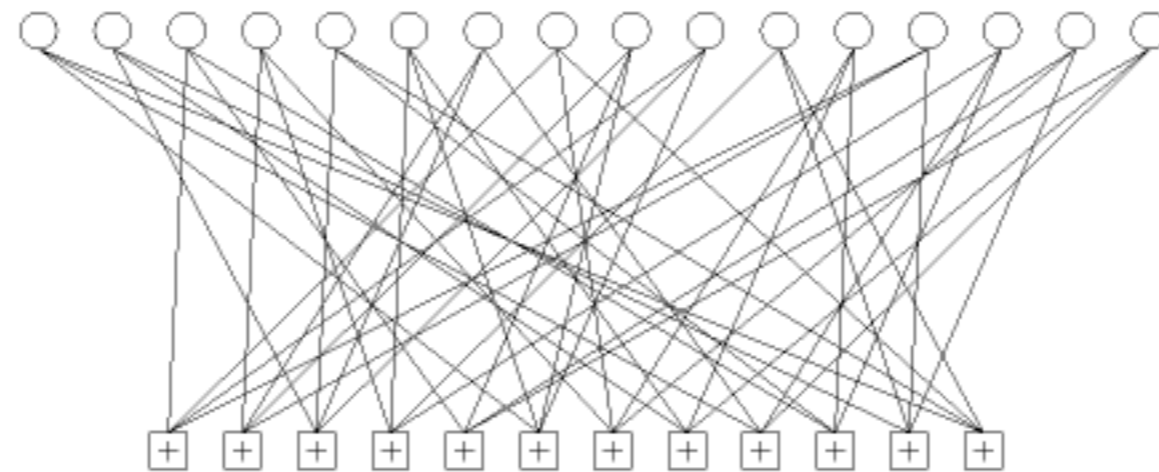
Syndrome decoder $\mathbf{z} \rightarrow \hat{\mathbf{n}}$.



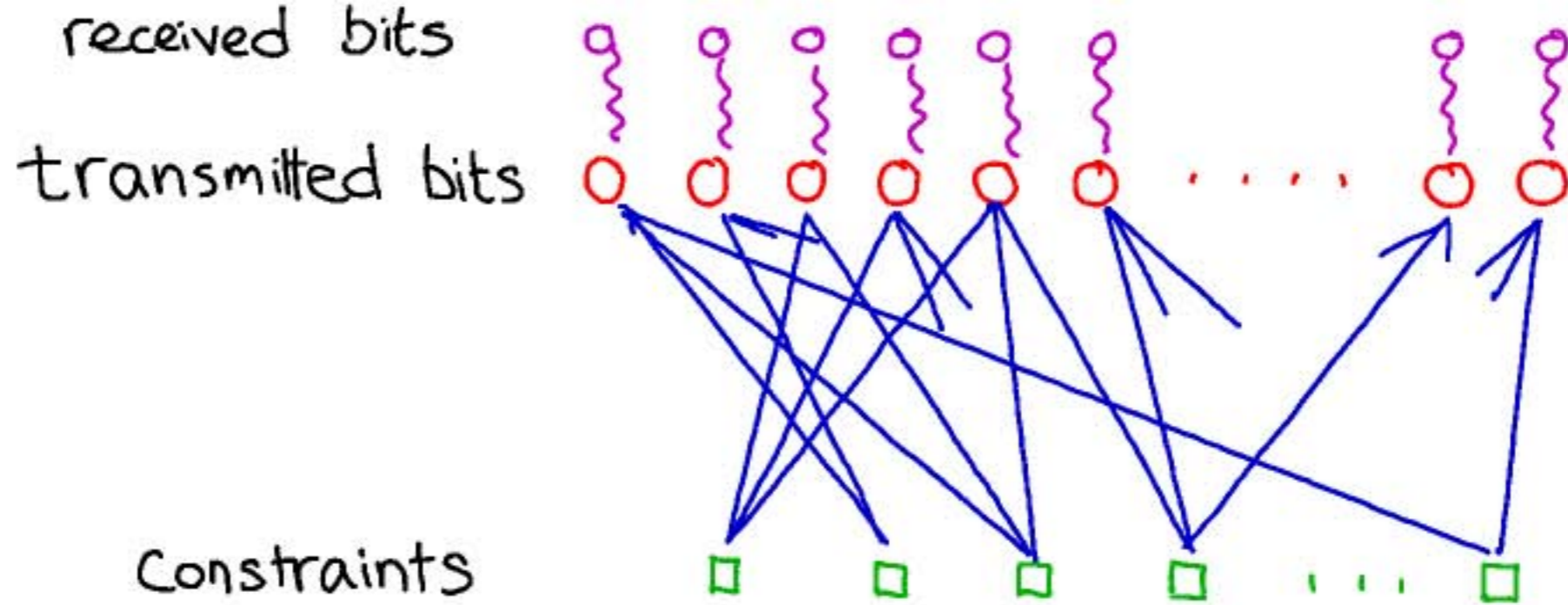
factor graph

Low-density parity-check code

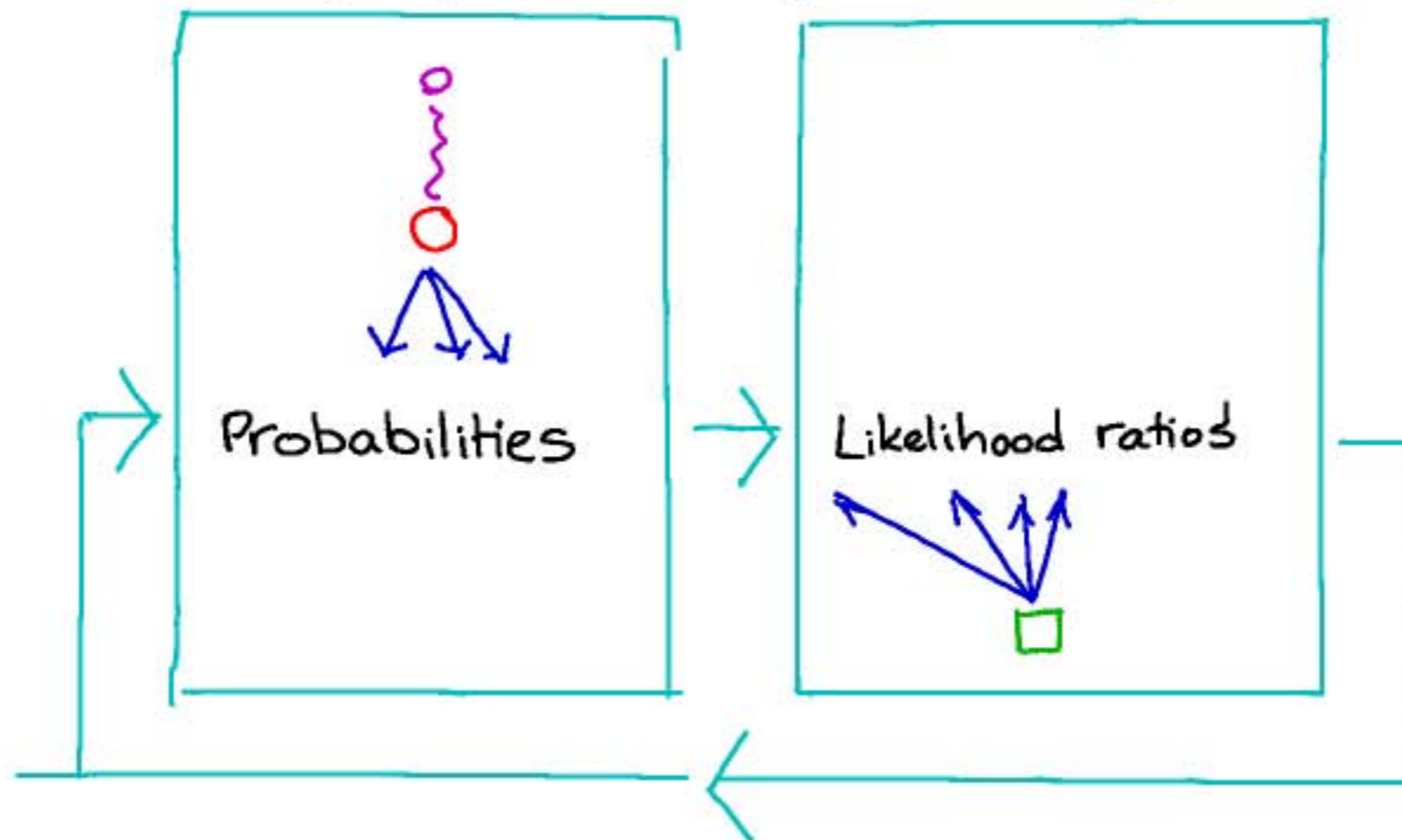
$$\mathbf{H} = \begin{array}{c} \begin{array}{cccccccccccc} & & & & 1 & & & & & 1 & 1 & & & 1 & \\ & & & & & & & & & & 1 & & & & 1 \\ & & 1 & & & 1 & & & 1 & & & 1 & & & 1 \\ & & & 1 & & & 1 & & & & & & & 1 & & 1 \\ & & & & 1 & & & & 1 & & & & & & & \\ 1 & & & & & & 1 & & & 1 & 1 & & & & & \\ & 1 & & & & & & 1 & & & & 1 & & & & \\ & & 1 & & & 1 & & & & & & & & & 1 & \\ 1 & & & & 1 & & & & & & 1 & & & & & \\ & 1 & & & & & & & & & & 1 & & & & \\ & & 1 & & & & & & & & & 1 & & & & \\ & & & 1 & & & & & & & & & & & & \\ & 1 & & & & & & & & & 1 & & & & & \end{array} \end{array}$$



Gallager 1962; MacKay & Neal 1995



Decoding by the sum-product algorithm



Low Density Parity Check Code

We demonstrate a large code that encodes $K = 10000$ source bits into $N = 20000$ transmitted bits.

Each parity bit depends on about 5000 source bits.

The encoder is derived from a very sparse 10000×20000 matrix \mathbf{H} with three 1s per column.

TRANSMITTED:

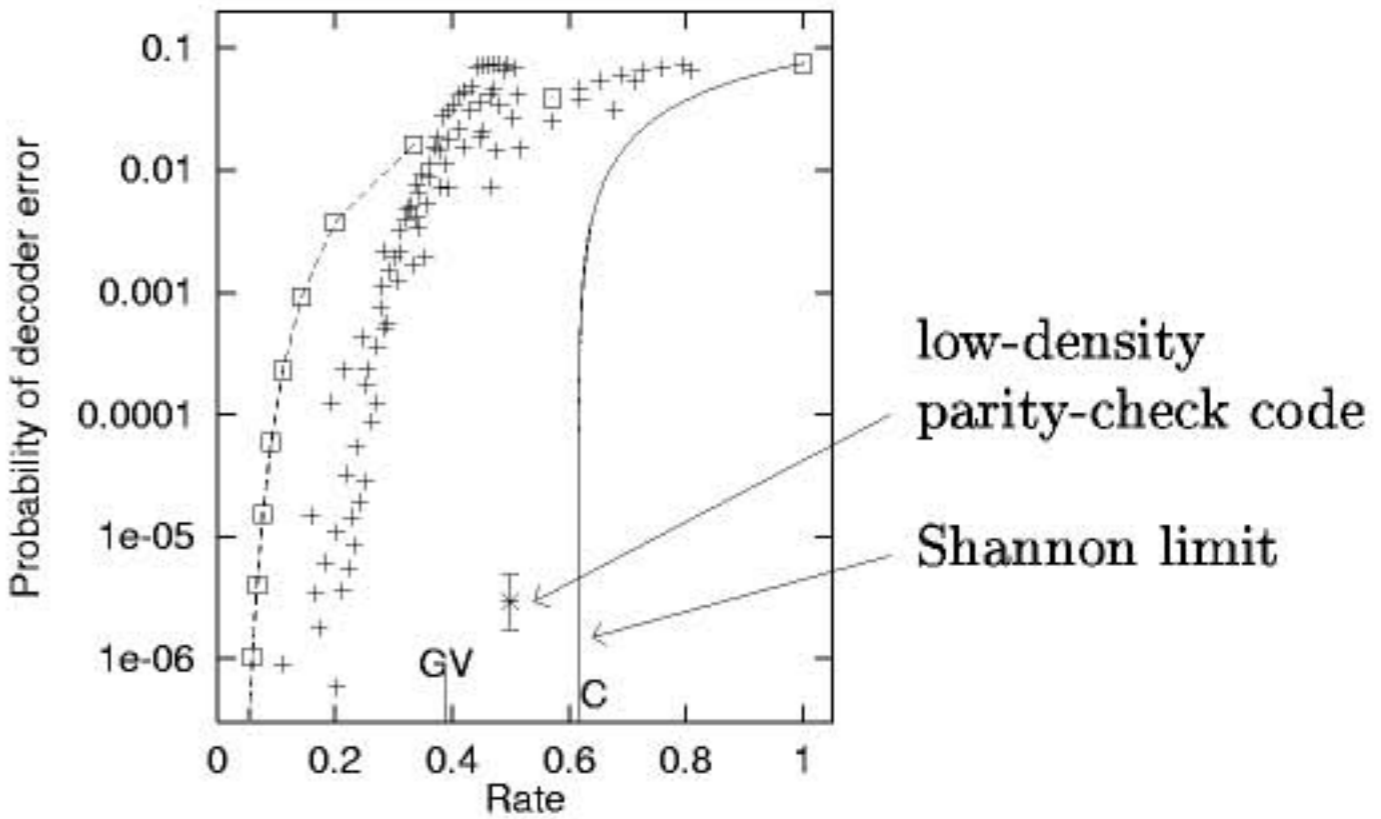


$\mathbf{H} =$

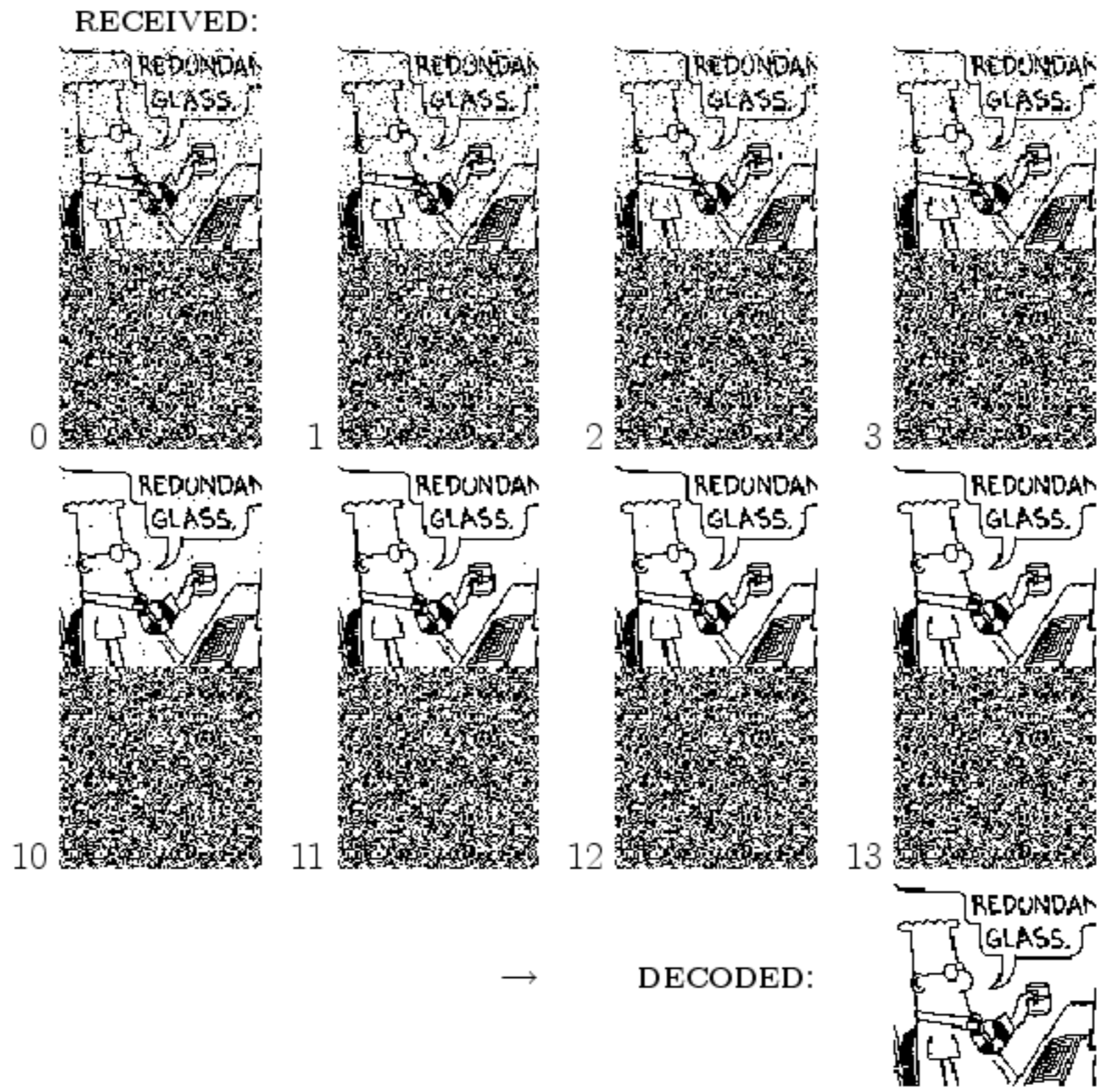


Low Density Parity Check Code (f = 7.5%)

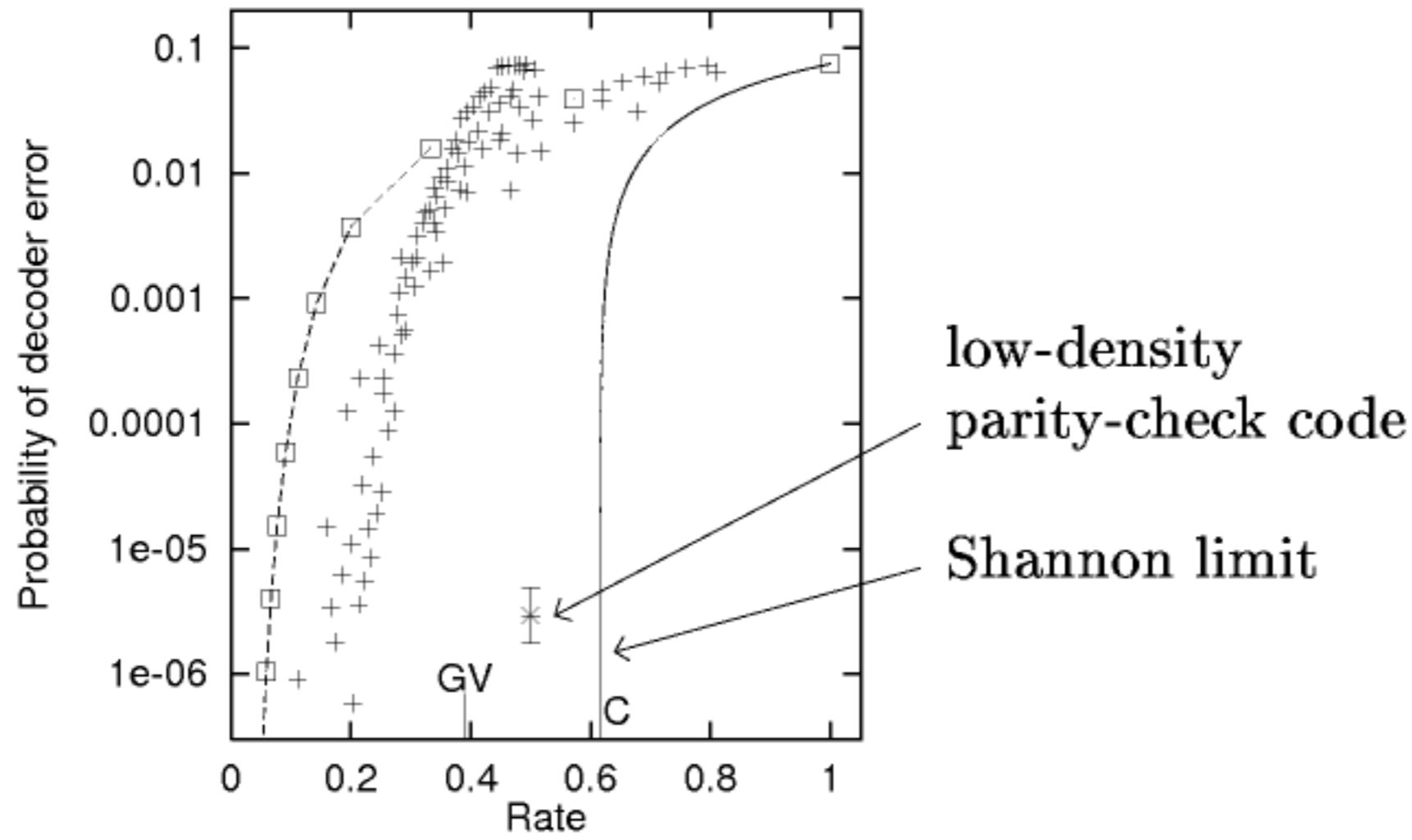
Iterative probabilistic decoding



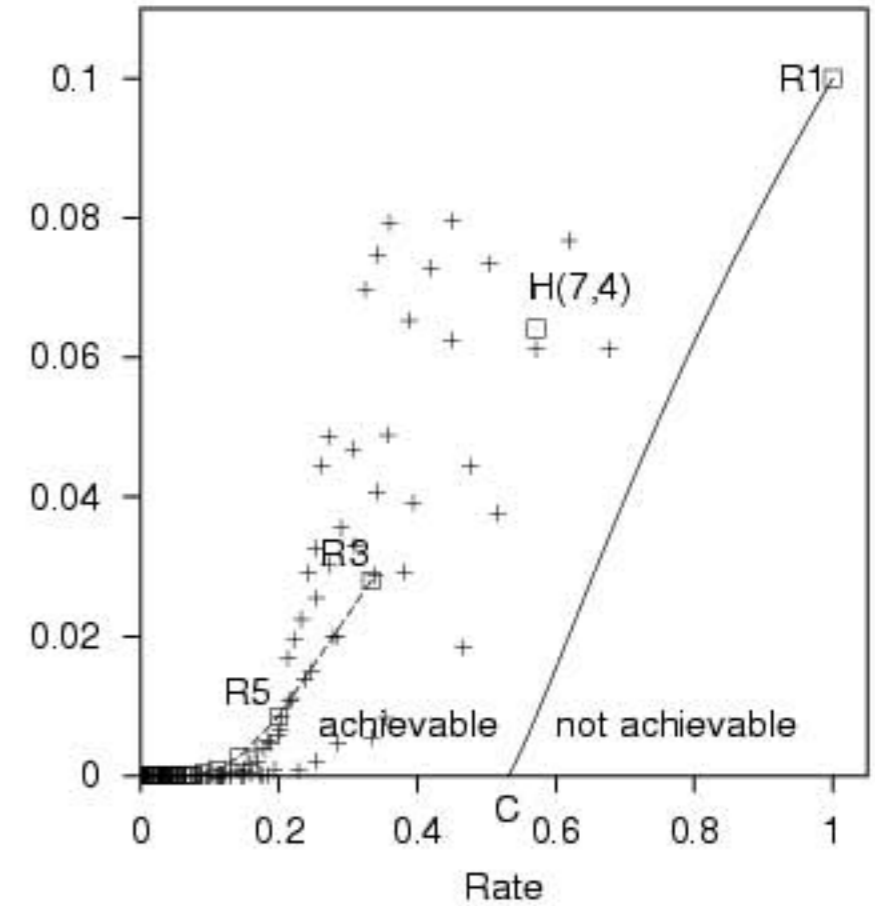
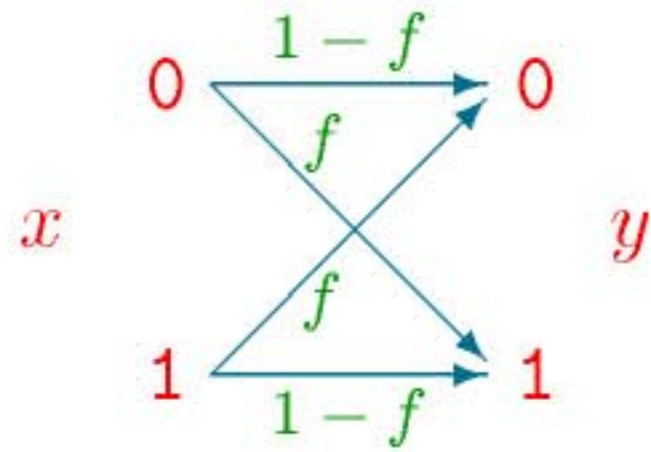
After the transmission is sent over a channel with noise level $f = 7.5\%$:



BSC: $f=7.5\%$

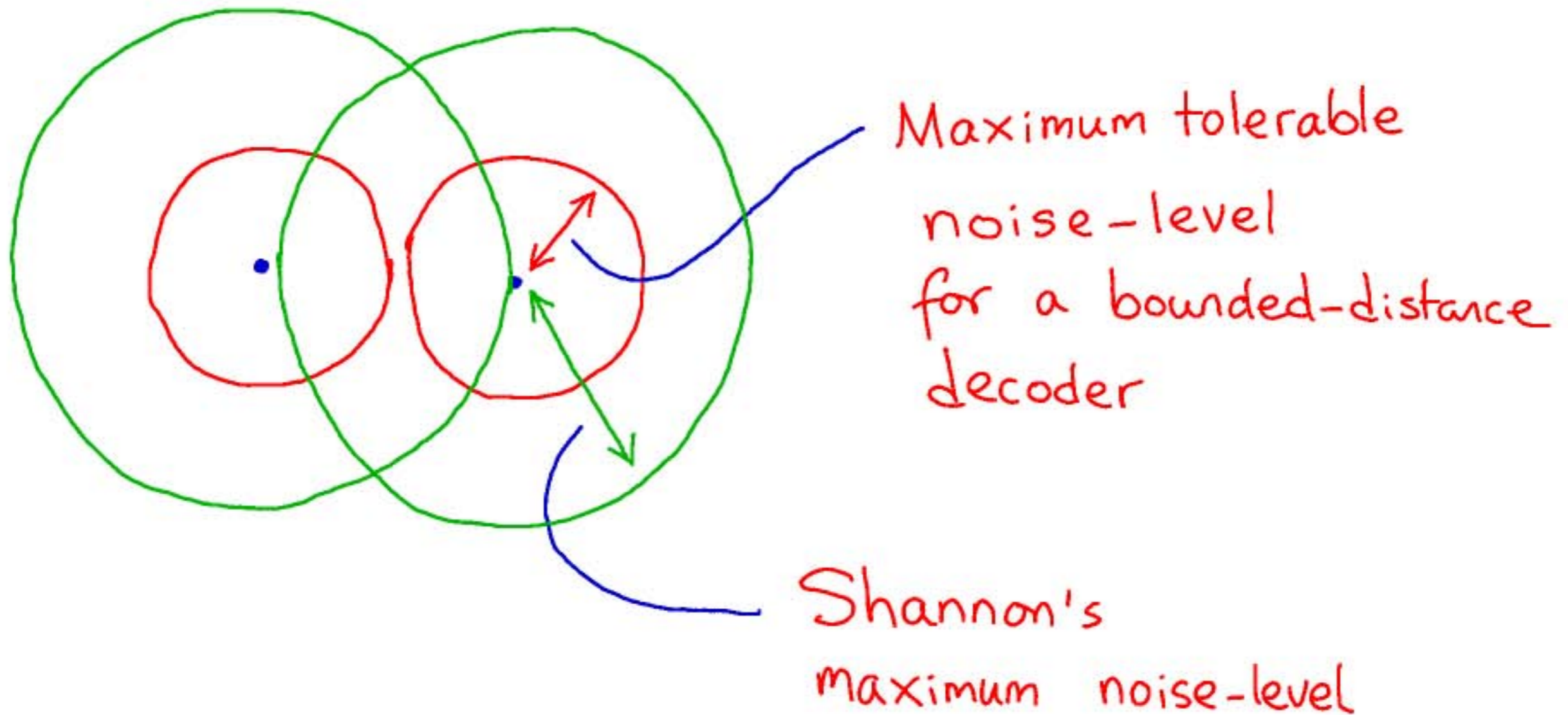


Rate, error probability, complexity



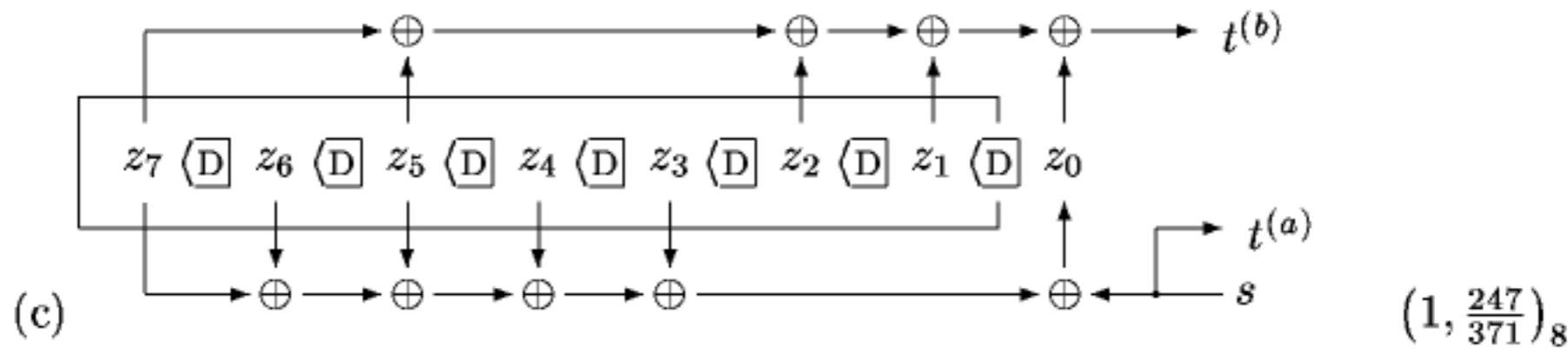
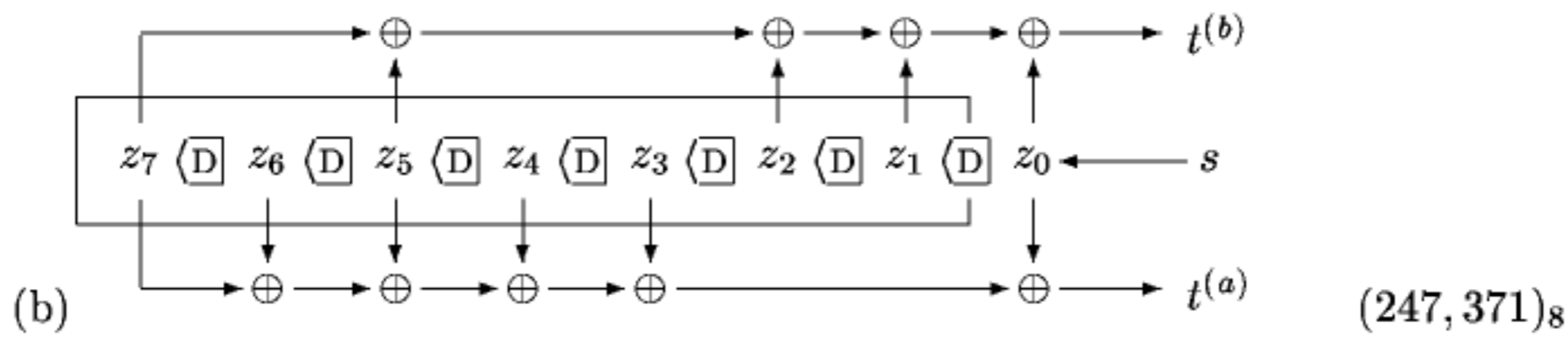
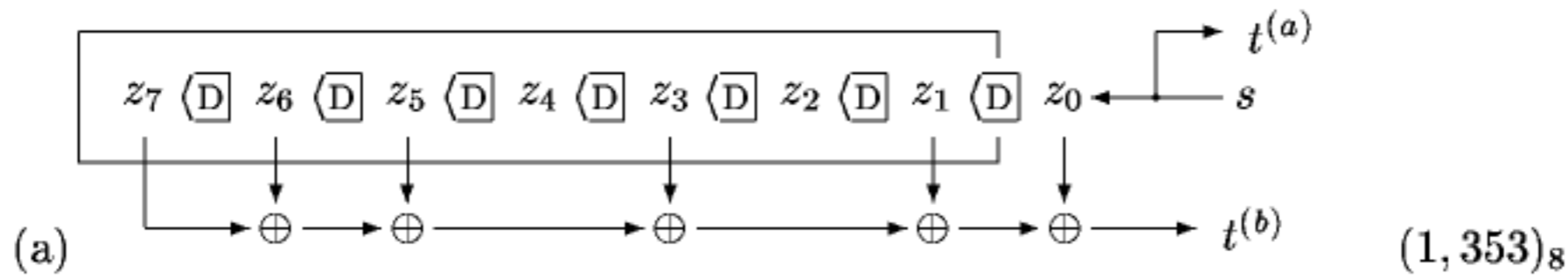
For any channel:
Reliable (virtually error-free) communication is possible
at rates up to C

Two codewords

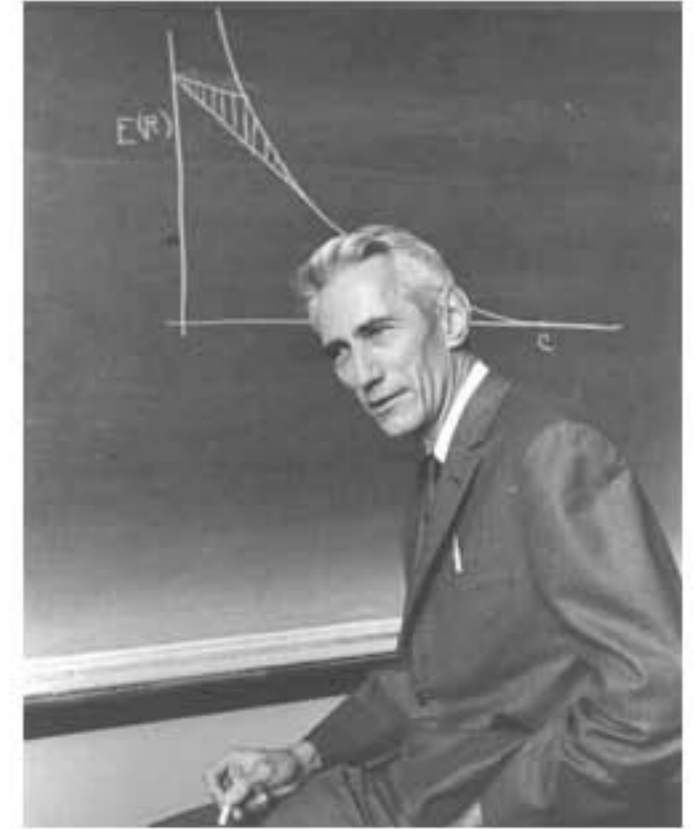


Convolutional codes

Octal name



Feedback



`Feedback? Pah! Who needs feedback?
Just use a random code!'

Sphere packing view

Count inputs and outputs \rightarrow get a bound on what's achievable.

Given a transmission of length N ,
the output will probably have Nf bits flipped,
so it will be in a typical set of size

$$\binom{N}{Nf} \simeq 2^{NH_2(f)}$$

So if we have 2^K alternative inputs, and almost all these typical outputs are distinct, we must have

TOTAL NUMBER OF TYPICAL OUTPUTS

$$\underbrace{2^K \times 2^{NH_2(f)}}_{\text{TOTAL NUMBER OF TYPICAL OUTPUTS}}$$

\leq

TOTAL SIZE OF OUTPUT SPACE

$$\underbrace{2^N}_{\text{TOTAL SIZE OF OUTPUT SPACE}}$$

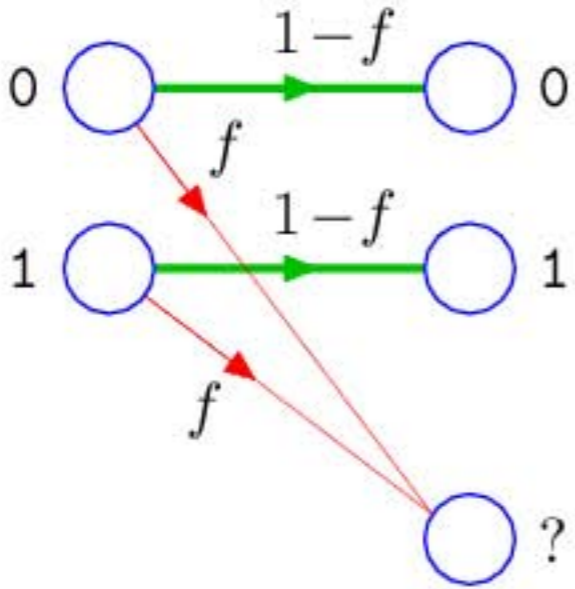
i.e.,

$$K + NH_2(f) \leq N$$

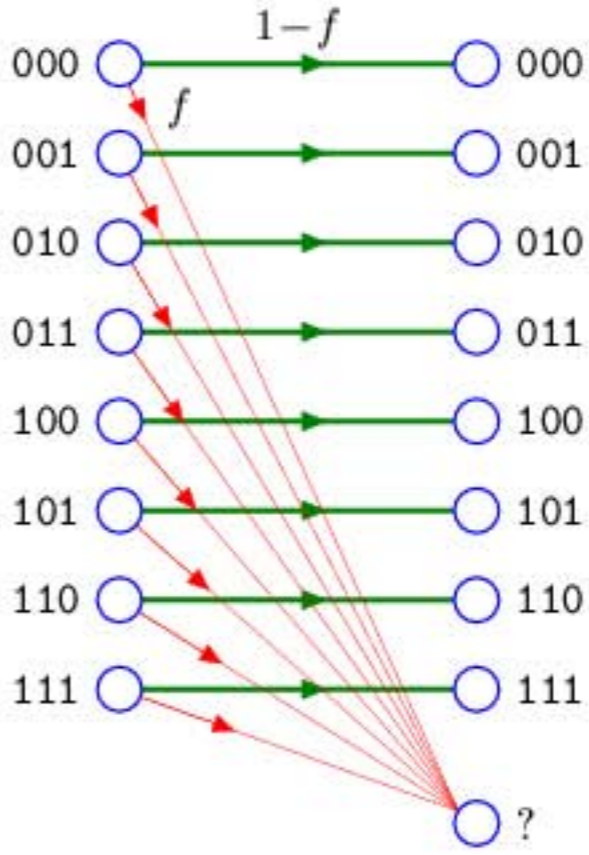
i.e.,

$$\frac{K}{N} \leq 1 - H_2(f)$$

Channels with erasures



Binary erasure channel



8-ary erasure channel

Alice



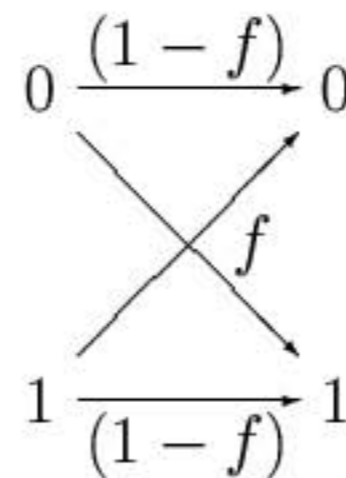
Bob



Packet-erasing channel



There are other noisy channels...



Insertions
and
Deletions
→

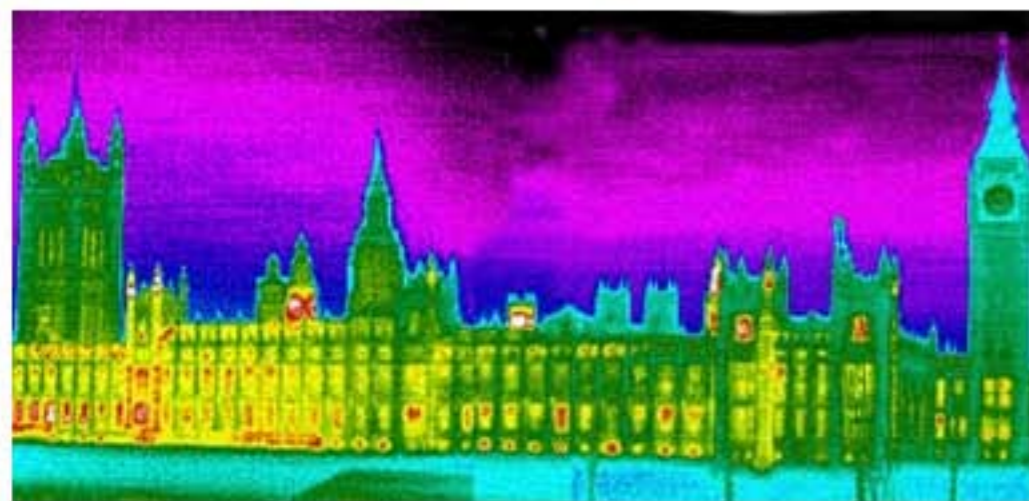


synchronization errors

From [The Sunday Times](#)

August 30, 2009

Bestselling guru David MacKay to lead climate fight



A thermal imaging camera reveal large amounts of hot air escaping from the Houses of Parliament

[Dominic O'Connell](#)

RECOMMEND?

A CAMBRIDGE academic who has suggested importing solar energy from the Sahara and using Scottish lakes as giant batteries is to be named the government's scientific adviser on climate change.

Write free software and free books, and who knows where you will end up

THE  ON SUNDAY
INDEPENDENT

BUSINESS



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Nuclear energy firms welcome policy adviser

Government to appoint Professor David MacKay to help with its ambitious climate change strategy

By Mark Leftly

Sunday, 30 August 2009

SHARE | PRINT | EMAIL | TEXT SIZE

Ed Miliband, the Energy and Climate Change Secretary, is set to appoint Cambridge University Professor David MacKay as his chief scientific officer this week.

The move will be a boon to the British energy sector: industry leaders from Royal Dutch Shell, EDF Energy and QinetiQ have all praised Professor MacKay's hugely successful book, *Sustainable Energy: Without the Hot Air*. Companies looking to get involved in the Government's nuclear roll-out programme will be particularly hopeful that his appointment will quash some of the political arguments against the plans.

Another book

Sustainable Energy –
without the hot air

David JC MacKay

