

+0.1 16.2.26

+0.1 11.2.26

+0.1



80% chance of rain in Oxford.
This introduces the probabilistic nature of weather predictions.



ink + think

A better learning method (more effective)

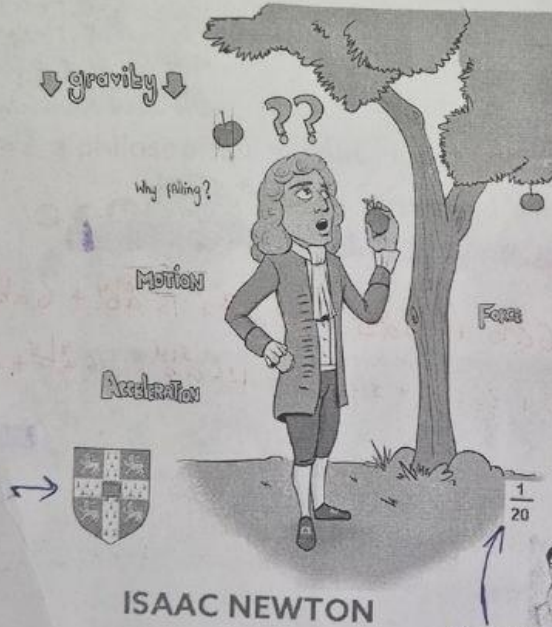


(1) listening
(2) first way of processing
(3) Writing, incl. str. you're not quite sure about

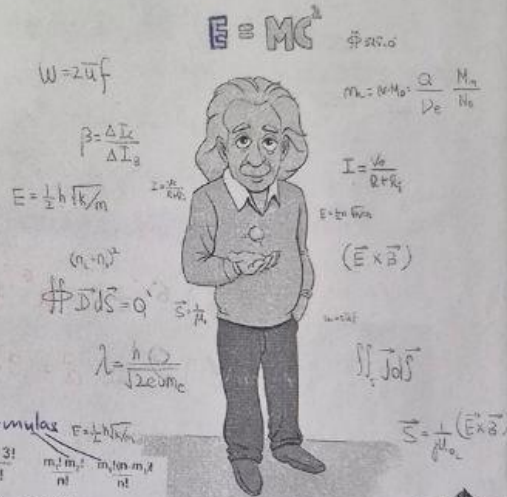
work with concrete
School \downarrow gravity \downarrow **MOTION**

Bad way
shift from concrete to abstract
==formalism==> University $E=MC^2$ Φ $\vec{A} \times \vec{B}$ $\int \vec{J} \cdot \vec{d}\vec{s}$

(CONCRETE) AND (ABSTRACT) THINKING



ISAAC NEWTON



ALBERT EINSTEIN

permutations of vowels & consonants in name GALOIS

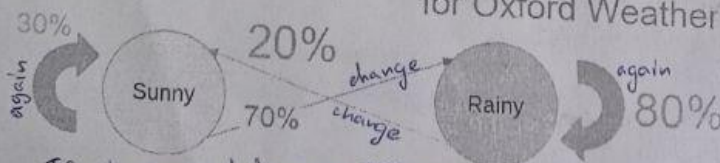
Motivation: 80% chance of rain
Let A_1 be the event of rain on Jan
on any day of this term, $1 \leq j \leq n$
Suppose the events A_j
are independent.

Oxford			
Tue 13th	Wed 14th	Thu 15th	Fri 16th
10° 9°	13° 10°	13° 8°	11° 7°
70%	70%	70%	70%

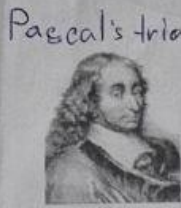
Markoff Chain Probability Model

for Oxford Weather

The probability of tomorrow's weather depends only on today's

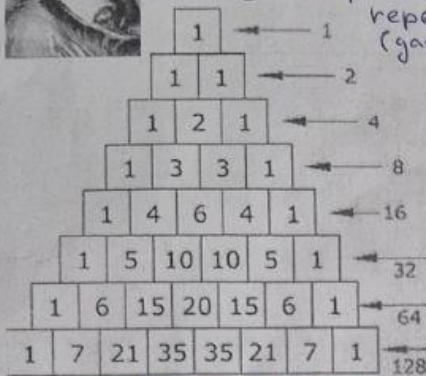


If it's sunny today \rightarrow 70% sunny tomorrow, 20% rain tomorrow;
If it's rain today \rightarrow 80% rain tomorrow, 20% sunny tomorrow.



Pascal's triangle is used to quickly find numbers associated with combinations & probabilities in repeated trials (games, coin)

Pascal's triangle



$$1 \quad 8 \quad 28 \quad 56 \quad 70 \quad 56 \quad 28 \quad 8 \quad 1$$

$$1 \quad 9 \quad 36 \quad 84 \quad 126 \quad 126 \quad 84 \quad 36 \quad 9 \quad 1$$

$$(a+b)^0 = 1$$

$$(a+b)^1 = a + b$$

$$(a+b)^2 = a^2 + 2ab + b^2$$

$$(a+b)^3 = a^3 + 3a^2b + 3ab^2 + b^3$$

$$(a+b)^4 = a^4 + 4a^3b + 6a^2b^2 + 4ab^3 + b^4$$

$$(a+b)^5 = a^5 + 5a^4b + 10a^3b^2 + 10a^2b^3 + 5ab^4 + b^5$$

$$(a+b)^6 = a^6 + 6a^5b + 15a^4b^2 + 20a^3b^3 + 15a^2b^4 + 6ab^5 + b^6$$

$$(a+b)^9 = a^9 + 9a^8b + 36a^7b^2 + 84a^6b^3 + 126a^5b^4 + 126a^4b^5 + 84a^3b^6 + 36a^2b^7 + 9ab^8 + b^9$$

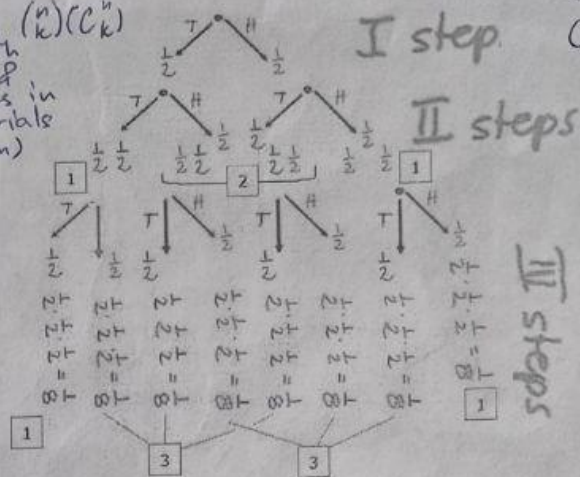
binomial coefficients (how many ways to select k successes from n trials)

$$\binom{n}{k} = \binom{n}{n-k}$$

$$\binom{2}{0} = 1$$

$$\binom{2}{1} = 2$$

$$\binom{2}{2} = 1$$



Newton's Binomial



expansion of $(a+b)^n$ with coefficient C_k^n , which are taken from Pascal's triangle

Handwritten scribble

$$\frac{1}{20} = \frac{\text{Conv. fac. } 3!3!}{\text{Dye. } 6!} = \frac{n_1!(m-n_1)!}{m!}$$

TO, 1

Events & Probabilities

$\omega \in \Omega$ - sample space

For example:

a) $\Omega = \{H, T\}$ - a coin

b) $\Omega = \{(i, j) \mid i, j \in \{1, 2, 3, 4, 5, 6\}\}$ - a dice

A subset of Ω is called an event.

1) coin comes up tail $A = \{T\}$

2) we observe of total of 9

$A = \{(3, 6), (4, 5), (5, 4), (6, 3)\}$ - events

If $\omega \in \Omega$ is the outcome, we say A occur if $\omega \in A$.

Complement of A

$\bar{A} = A^c$ occurs if A does not occur.

Union: $A \cup B$ occurs if A or B occurs

Intersection: $A \cap B$ occurs if both A and B occur.

Set differences: $A \setminus B = A \cap B^c$

Disjoint: A and B disjoint if $A \cap B = \emptyset$

A and B cannot occur together

We assign a probability $P(A)$ of each A

Simplest case:

Ω is finite and all outcomes

Then $P(A) = \frac{|A|}{|\Omega|}$

⊙
↑
Bout

a) $|\Omega| = 2 \quad ? \quad P(A) = \frac{1}{2}$
 $|A| = 1 \quad \downarrow$

b) $|\Omega| = 36 \quad ? \quad P(A) = \frac{4}{36} = \frac{1}{9}$
 $|A| = 4 \quad \downarrow$

Elementary combinatorics

Arrangements of distinguishable objects

Suppose we have n distinguishable objects

How many ways are there to order them?

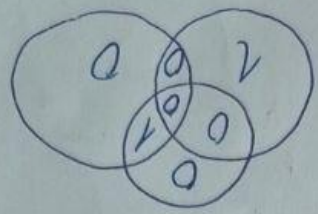
$n = 3 \quad n!$

$3! = 6$

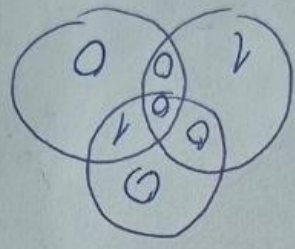
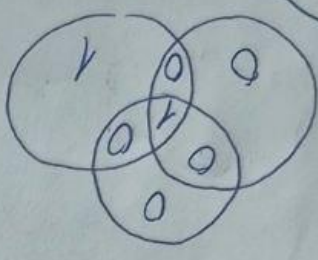
C_m^n - binomial coefficient = $\frac{n!}{m!(n-m)!}$

$C_m^n = \binom{n}{m}$

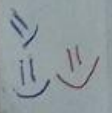
$$e^{j\pi} = -1 \quad \sqrt{-1} \quad \begin{matrix} b \\ \sqrt{a^2+b^2} \\ a \end{matrix} (a+jb)$$



1'0+



↑
Lumospere bot Tare



Boolean vs Aristotelian logic:

Aristotelian - is logic of propositions about classes (syllogisms)

Boolean - is the operations AND, OR and NOT on true/false



Massachusetts Institute of Technology (MIT)

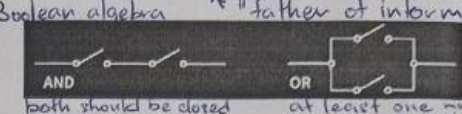
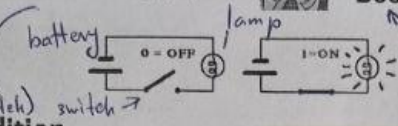


Lecture by Pr. Bob Gallagher

Boole (1815-1864) & Shannon (1916-2001)



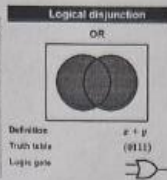
(B. logic as the basis of switch)



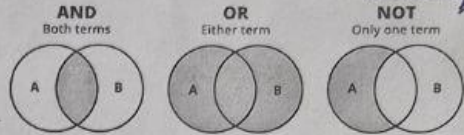
Logical addition (disjunction)

A	B	F=A∨B
0	0	0
0	1	1
1	0	1
1	1	1

A	B	A∧B
True	True	True
True	False	False
False	True	False
False	False	False



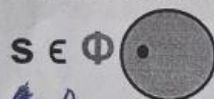
BOOLEAN LOGIC - Venn diagrams



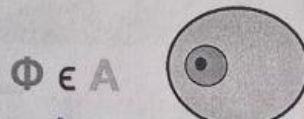
Good logic



Socrates was a philosopher



philosophers are men



Plato



Aristotle



Socrates was a man

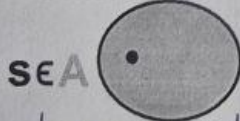


A correct syllogism in Aristotelian logic.

Bad logic



Socrates was a man



Socrates

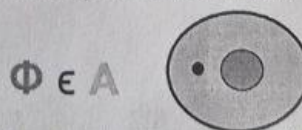


Plato



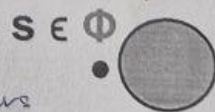
Aristotle

philosophers are men



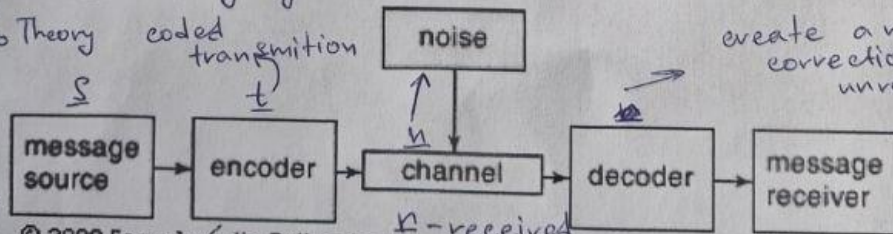
Socrates

Socrates was (error) a philosopher



Incorrect syllogism: not all men are philosophers

This is what classical Info Theory coded transmission is about...



create a reliable correction over and unreliable (noisy) channel.

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(Shannon communication model)

+ error correction

infers S and r

Resume of Lecture by Pr. Bob Gallager from MIT

Massachusetts Institute of Technology (MIT)

But logic understood since Aristotle time

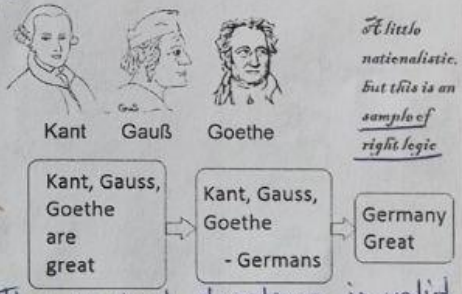
George Boole (1815-1864) developed Boolean logic
 The principles of logical thinking have been understood (and occasionally used) since the Hellenic era.
 Boole's contribution was to show how to systemize these principles and express them in equations (called Boolean logic or Boolean algebra).
Claude Shannon (1916-2001) showed how to use Boolean algebra as the basis for switching technology. This contribution systemized logical thinking for computer and communication systems, both for the design and programming of the systems and their applications.
 Logic continues to be abused in politics, religion, and most non-scientific areas.



Boole created the mathematical of logic and Shannon connected logic with electrical engineering

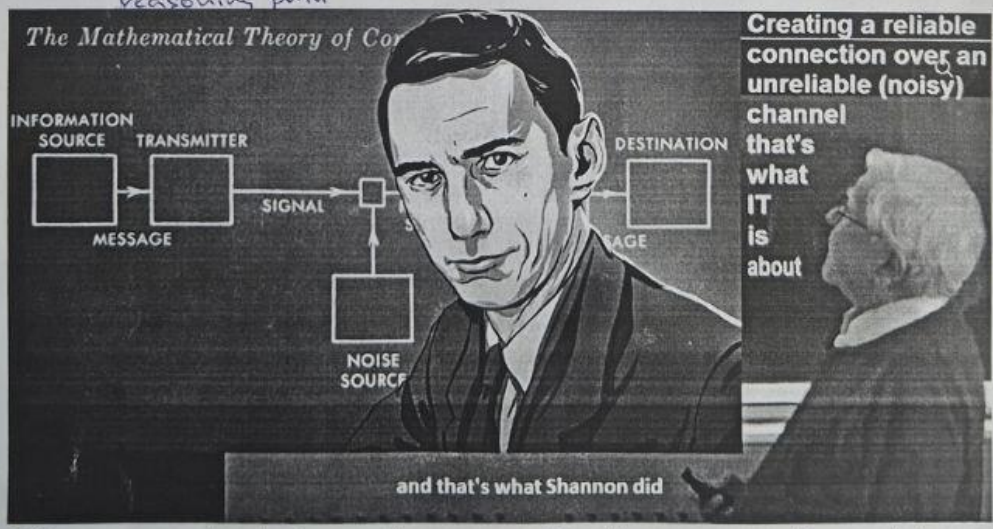
Created Information theory, his major work

People often use logical forms incorrectly (draw false conclusions) and try to abuse the normal one.



Bad logic (abuse of logic)
 An individual incorrectly generalized personal attributes on group identity.

The logical structure is valid and follows a clear, correct reasoning path



Main Idea

Hedging in the fog => Information theory.

A diagram of a communication system

Data => facts
 Info => organized data

and that's what Shannon did

If you encode information correctly, then even through a noisy channel you can transmit data almost without errors

Solution - Phys System

+0.1

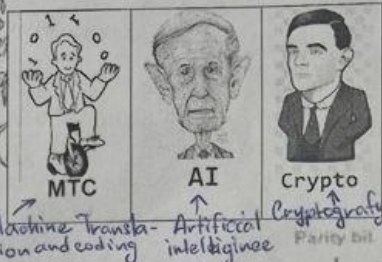
+0.1

$$9000 \pm 30 \quad w(1)$$

+0.1

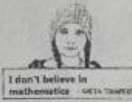
Do tip if I make error

Sir Dr. D. MacKay,
University of Cambridge
(22 April 1967 - 14 April 2016)

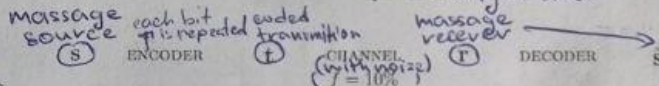
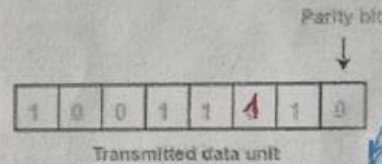
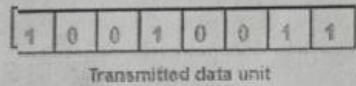
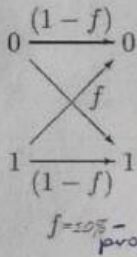


Parity bit = extra bit that makes the num of 1s even (odd) to detect single errors

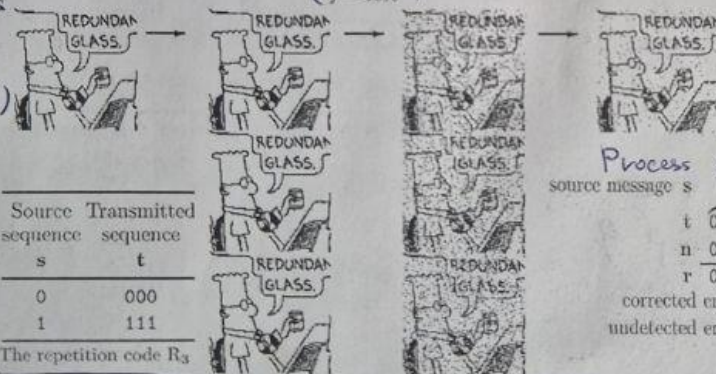
"I believe in clean energy, but I also believe in mathematics"



Redundant data helps ensure accuracy and reliability in the presence of errors.



more channels better (more copies) -> less data loss



Source sequence s	Transmitted sequence t
0	000
1	111

The repetition code R_3

Process of error correction:

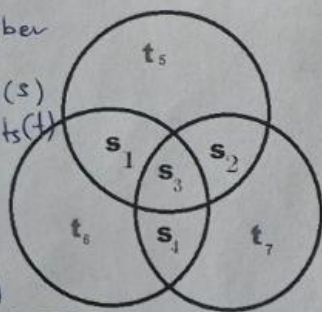
source message s	0	0	1	0	1	1	0
t	000	000	111	000	111	111	000
n	000	001	000	000	101	000	000
r	000	001	111	000	010	111	000

corrected errors *
undetected errors ⊙

7.4. Hamming code.

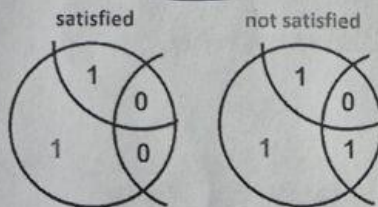
$$\frac{4}{\Sigma} \rightarrow \frac{7}{t}$$

7 - total number of bits (t, s)
4 - data bits (s)
3 - parity bits (t)



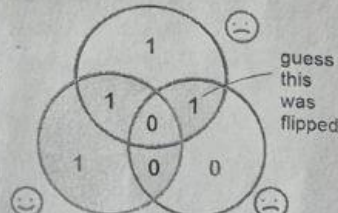
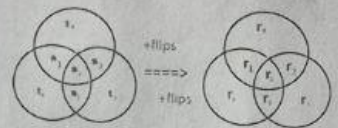
Ex: $n = 2^m - 1, m = 3$
(m - parity bits)
 $n = 2^3 - 1 = 2^3 - 1 = 7$

represents different sets of parity bits. The intersections - the data bit was protected by parity bit.



$1+1+0+0=0$ $1+1+0+1=1$

(1) $6^2 = n p q$
10000 10%
90%

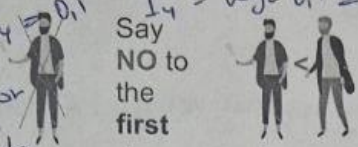


how the overlaps helps in pointing the errors.

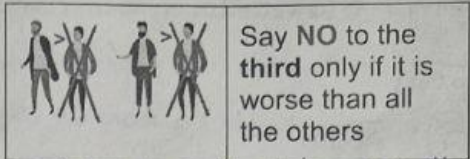
$p_1 = 0,6 \quad I_1 = \log_2 \frac{1}{0,6} = 0,736966$
 $p_2 = 0,2 \quad I_2 = \log_2 \frac{1}{0,2} = 2,32193$
 $p_3 = 0,1 \quad I_3 = \log_2 \frac{1}{0,1} = 3,32193$
 $p_4 = 0,1 \quad I_4 = \log_2 \frac{1}{0,1} = 3,32193$

$0,6 \cdot 0,736966 + 0,2 \cdot 2,32193 + 0,1 \cdot 3,32193 + 0,1 \cdot 3,32193 = 1,5709516$

An example of a strategy for selecting the best candidate through sequential comparisons.



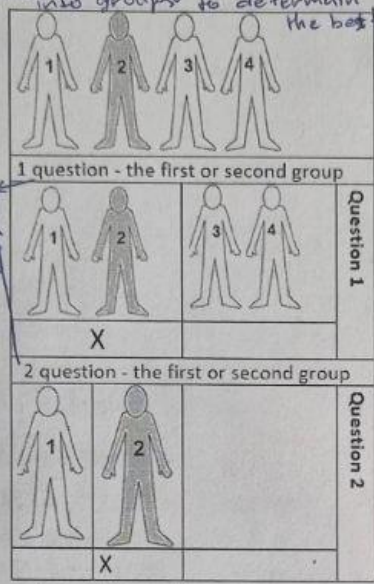
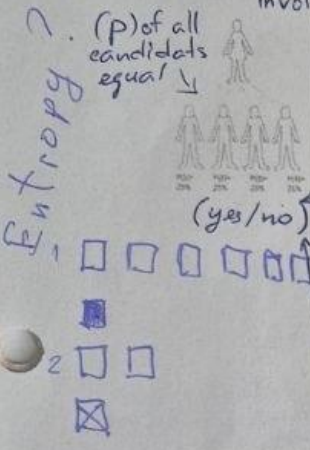
Say YES to the second if it is better than the first



Say NO to the third only if it is worse than all the others

The decision-making process involves comparing these candidates into groups to determine the best.

explains how to determine the average num of questions needed to identify a specific person from a set of candidates based on their (p) different



$1 \cdot 0,5 +$	$2 \cdot 0,25 +$	$3 \cdot 0,125 +$	$3 \cdot 0,125$

Question 1. Is this Zuckerberg?		$1 \cdot 0,5$ 50%
Question 2. Is this Sergey Brin?		$2 \cdot 0,25$ 25%
Question 3. Is this Stefan from BMW?		$3 \cdot 0,125$ 12,5%
So Prince Saud		$3 \cdot 0,125$ 12,5%
Average number of questions = 1,75		

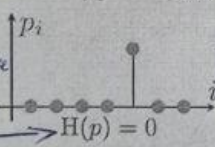
$1 \cdot 0,6$
 $2 \cdot 0,2 = 0,4$
 $3 \cdot 0,1 = 0,3$
 $3 \cdot 0,1 = 0,3$

Average number of questions = $2 \cdot 0,25 + 2 \cdot 0,25 + 2 \cdot 0,25 + 2 \cdot 0,25 = 2$

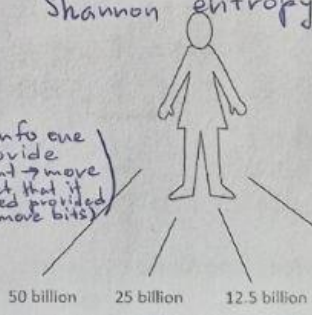
Shannon's entropy formula $H(p) = -\sum_{i=1}^n p(x_i) \log_2 \frac{1}{p(x_i)}$

Shannon entropy

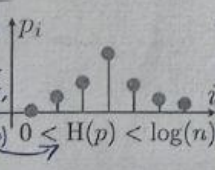
Zero entropy: there is almost no uncertainty, we know in advance that the same outcome will always occur.



Quantifying information (how much info one outcome provide (rarer event → more info the fact that it occurred provided (more bits)))



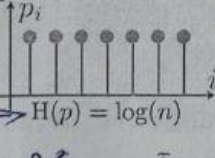
Several outcomes have probabilities but not equal. There is uncertainty, but it's not max, so the entropy is between 0 and $\log_2(n)$.



$I(x_i) = \log_2 \left(\frac{1}{p_i} \right)$



All outcomes are equally likely → the max uncertainty (entropy): we can't guess the outcome at all.



number of bits required to encode choice (the average num of bits needed to report the result (entropy of source))

Mark Zuckerberg	Sergey Brin	Stefan Quandt	Prince Al Saud
$P(1) = 50\%$	$P(2) = 25\%$	$P(3) = 12,5\%$	$P(4) = 12,5\%$

$I(1) = \log_2 \left(\frac{1}{0,5} \right) = 1 \text{ bit}$
 $I(2) = \log_2 \left(\frac{1}{0,25} \right) = 2 \text{ bit}$
 $I(3) = \log_2 \left(\frac{1}{0,125} \right) = 3 \text{ bit}$
 $I(4) = \log_2 \left(\frac{1}{0,125} \right) = 3 \text{ bit}$
 $H(x) = \sum_{i=1}^4 p(x_i) I(x_i) = 0,5 \cdot 1 + 0,25 \cdot 2 + 0,125 \cdot 3 + 0,125 \cdot 3 = 1,75 \text{ bits}$

$p_1 = 0,6 \quad I_1 = \log_2 \frac{1}{0,6} = 0,321928$
 $p_2 = 0,2 \quad I_2 = \log_2 \frac{1}{0,2} = 2,32193$
 $p_3 = 0,1 \quad I_3 = \log_2 \frac{1}{0,1} = 3,32193$
 $p_4 = 0,1 \quad I_4 = \log_2 \frac{1}{0,1} = 3,32193$

+ 0,3

$0,6 \cdot 0,321928 + 0,2 \cdot 2,32193 + 0,1 \cdot 3,32193 = 0,9219284 \approx 0,92$

$H(x) = \sum_{i=1}^n p(x_i) \log_2 \frac{1}{p(x_i)}$ - среднее кол-во битов на 1 исход.

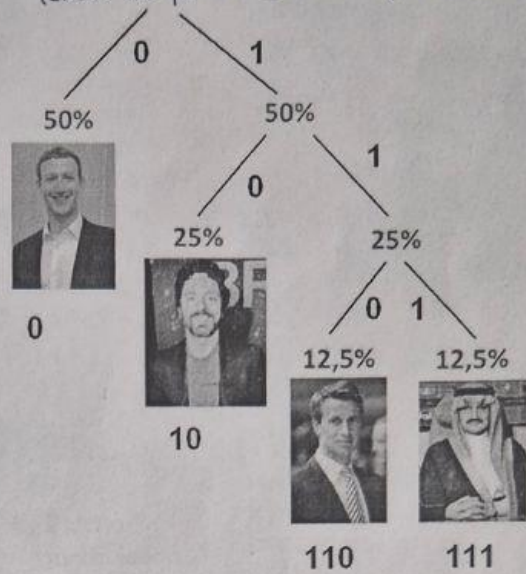
$p(x_i)$ - вероятность того, что произойдет исход x_i

$\frac{1}{p(x_i)}$ - "обратная вероятность"

$\log_2 \frac{1}{p(x_i)} = I(x) =$ кол-во битов в одном конкретном исходе (в битах \log_2)

Huffman Tree:

Yes/No question tree: we give a short code (0) to a frequent person, and longer ones (10, 110, 111) to use fewer bits on average for rare people. (Show the process of creating a Huffman coding tree)



we calculate the probabilities of individual letters in the message

↑
First-order approximation
(symbols independent but with frequencies of Belarusian txt).

Мама мыла ра

М - 3 — 30%	1-3 М
а - 4 — 40%	4-7 а
ы - 1 — 10%	8-ы
л - 1 — 10%	9-л
р - 1 — 10%	10-р
10	

лла мама р

Мама мыла ра

Ма - 2 22%	1-2 ма
ам - 2 22%	3-4 ам
мы - 1 11%	5 мы
ыл - 1 11%	6 ыл
ла - 1 11%	7 ла
ар - 1 11%	8 ар
ра - 1 11%	9 ра

9
Ni-position in the text

0.	4	6	7	3	1	9	1	6	7	3	5
ам	ыл	ла	ам	ма	ра	ма	ыл	ла	ам	мы	
мылла			рама								

2nd-order approx.

Second-order approximation (digram (2-symbols) structure as in Belarusian)

letter pair frequencies for more efficient coding.

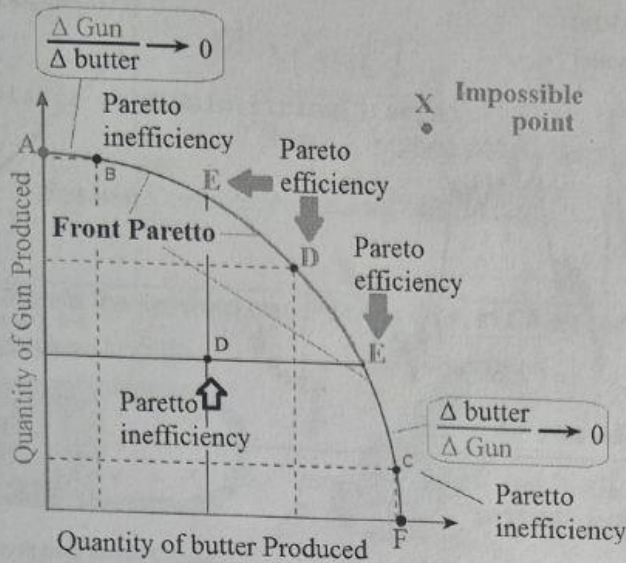


mamma mia
 m - 40% 0-25% 0: a - - a - - ma
 A - 30% 26-50% 1st: - imaiiamin - -
 u - 20% 51-75%
 i - 10% 76-100%

1st 2nd 1-12%
 0-40% ma - 12,5% mimiammaa - mimammia
 41-70% am - 25% 13-38%
 71-90% mm - 12,5% 39-51%
 91-100% mi - 25% 52% - 75%
 ia - 12,5%
 a - - 25% . . .

Pareto curve: above the frontier, one resource cannot be improved without worsening another; inside, resources are used inefficiently.

Pareto efficiency ~~is~~ - is a state where resources are allocated in the most efficient manner.



by Vilfredo Pareto
1848-1923

The orange sector E-D-E is the most Pareto efficient - since an increase in one indicator leads to a decrease in another.

Individual choice to confess worsens the overall outcome

Prisoners' dilemma

		prisoner B	
		confess	remain silent
prisoner A	confess	5 years 5 years	0 year 20 years
	remain silent	20 years 0 year	1 year 1 year

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Nash equilibrium - a concept of game theory where no player can benefit by changing their strategy while the other players keep their unchanged.

Game Theory Nash Equilibrium



** => Nash equilibrium

		H ₂ (x)	
		Recognition;	Non-recognition;
H ₁ (x)	Player 1	1	2
	Recognition;	1, -5* -5	-20
Non-recognition;	2	-20	-1, -1

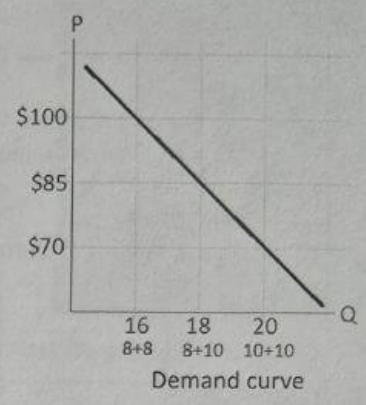
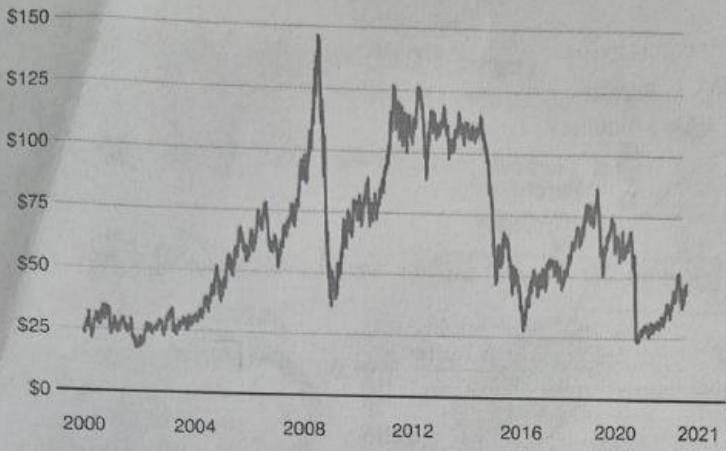
-1-1
Pareto Optimality



- Every mafioso has 2 options: remain silent or confess.
- 1) If both remain silent, they are considered "man of respect" and receive a more lenient punishment.
 - 2) If one confesses and the other remains silent, the one who confesses receives a prison sentence with freedom and is labeled as "penito". Who remains silent - receives the full punishment 20y.

Oil price hits 18-year low

Brent crude, US dollars per barrel



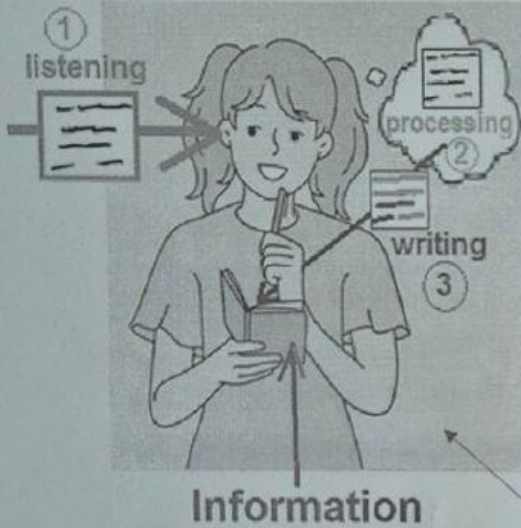
Pareto

		Barrel	
		1.	2.
1.	8 · 10 ⁶	$8 \cdot 10^6$ day \$800 millions per day \$100 \$800 millions per day	$10 \cdot 10^6$ day \$850 millions per day \$85 \$680
	10 · 10 ⁶	\$680 millions per day \$85 \$850 millions per day	\$700 millions per day \$70 \$700 millions per day
2.	10 · 10 ⁶		



Information Theory

Hearding



The founder of logic (Boolean logic)



ARISTOTLE

My lectures are published and not published; they will be intelligible to those who heard them, and to none beside [Letter to Alexander the Great].



Little information is like a hedgehog in the fog: high uncertainty.

HEDGEHOG IN THE FOG



~~entropy~~ Shannon - the average number of bits needed to describe a message.

$$\sum_x P(x) \cdot \log\left(\frac{1}{P(x)}\right)$$

to w