## Information Theory



Master of Logic 2014/15

2nd Block Nov/Dec 2014

Some of these slides are copied from or heavily inspired by the University of Illinois at Chicago, ECE 534: Elements of Information Theory course given in Fall 2013 by Natasha Devroye Thank you very much for the kind permission to re-use them here!

## Christian Schaffner





- me
- pure mathematics at ETH Zurich
- PhD from Aarhus, Denmark
- research: quantum cryptography
- c.schaffner@uva.nl
- plays <u>ultimate frisbee</u>

## Philip Schulz



- your teaching assistant
- PhD student @ILLC
- working on machine translation
- P.Schulz@uva.nl

#### Practicalities

- final grade consists of 50-50:
  - 7 weekly homework series, to be graded
  - final exam in week of 31/3/14 4/4/14: move to Tuesday, 16 December, but clashes with:
    - \* Lambda Calculus
    - \* Foundations of Neural and Cognitive Modelling
- details on course homepage:
   <a href="http://homepages.cwi.nl/~schaffne/courses/">http://homepages.cwi.nl/~schaffne/courses/</a>
   <a href="mailto:inftheory/2014/">inftheory/2014/</a>

## Expectations

#### We expect from you

- be on time
- code of honor (do not cheat)
- focus
- ask questions!

Why multitasking is bad for learning: <a href="https://medium.com/@cshirky/why-i-just-asked-my-students-to-put-their-laptops-away-7f5f7c50f368">https://medium.com/@cshirky/why-i-just-asked-my-students-to-put-their-laptops-away-7f5f7c50f368</a>

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#### You can expect from us

- be on time
- make clear what goals are
- listen to you and respond to email requests
- keep website up to date

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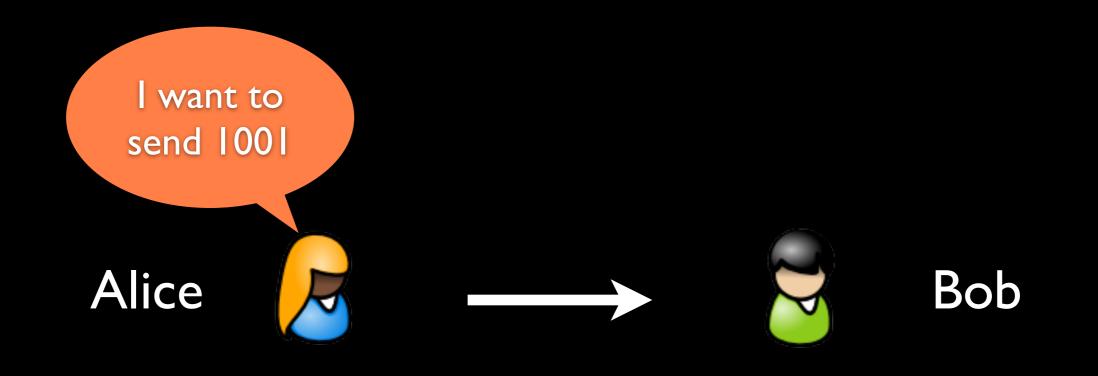
# Questions?

"The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point." - C.E. Shannon, 1948

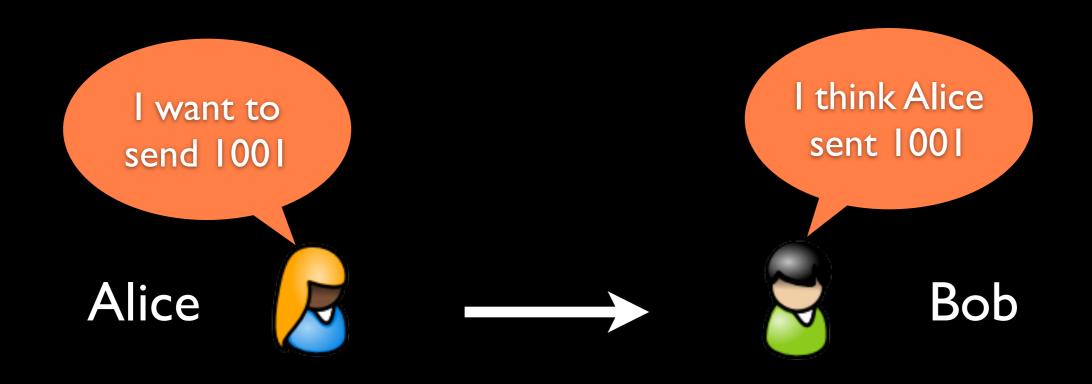
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Alice Bob

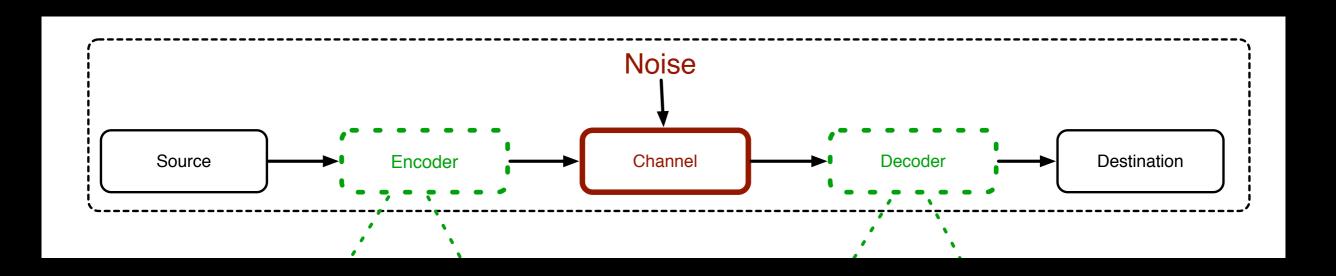
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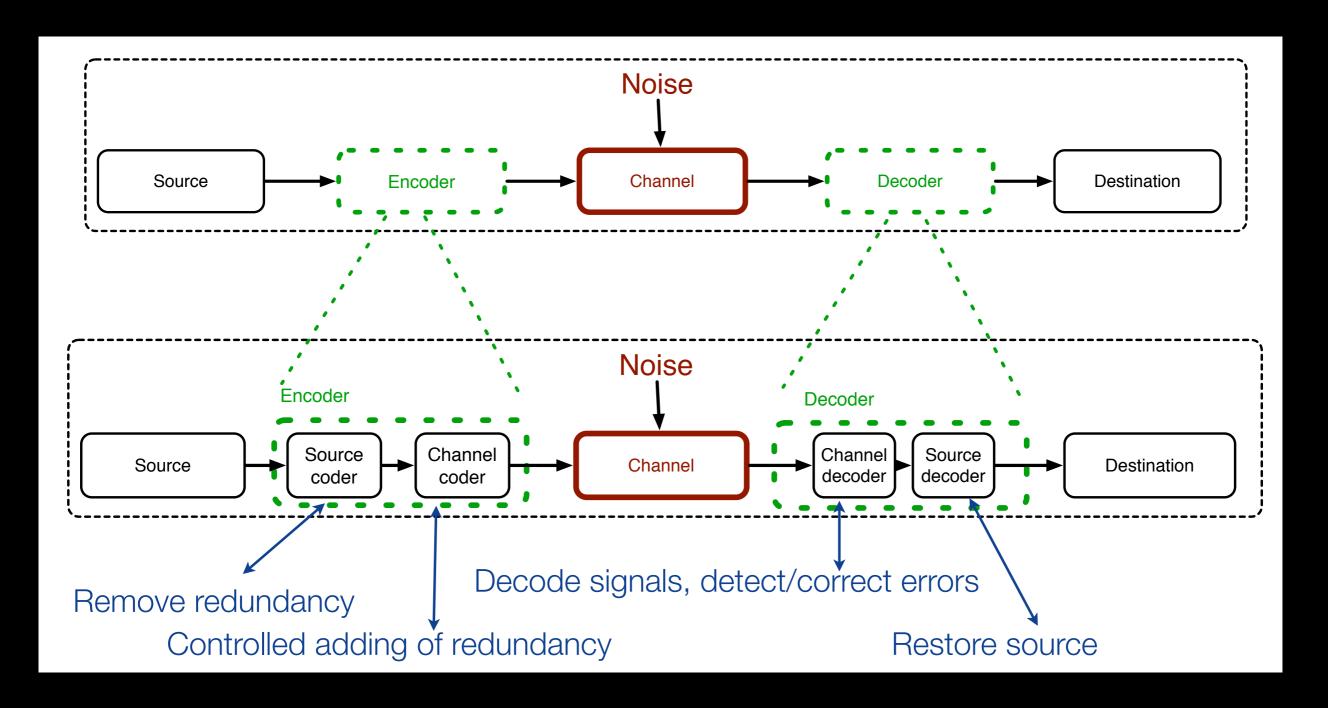
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## Generic communication block diagram



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Smoke signals



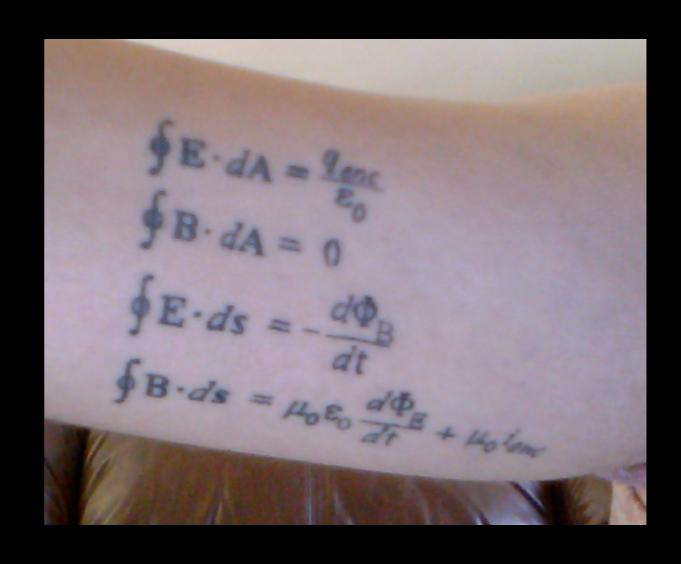
- Smoke signals
- 1861: Maxwell's equations

$$\oint \mathbf{E} \cdot d\mathbf{A} = \frac{q_{enc}}{\varepsilon_0}$$

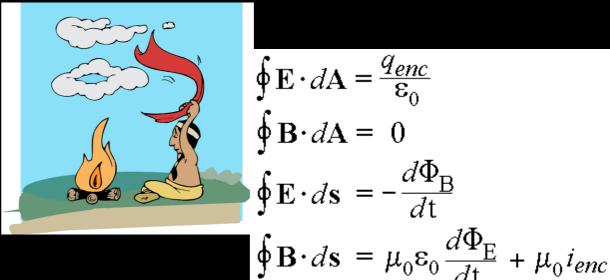
$$\oint \mathbf{B} \cdot d\mathbf{A} = 0$$

$$\oint \mathbf{E} \cdot d\mathbf{s} = -\frac{d\Phi_B}{dt}$$

$$\oint \mathbf{B} \cdot d\mathbf{s} = \mu_0 \varepsilon_0 \frac{d\Phi_E}{dt} + \mu_0 i_{enc}$$

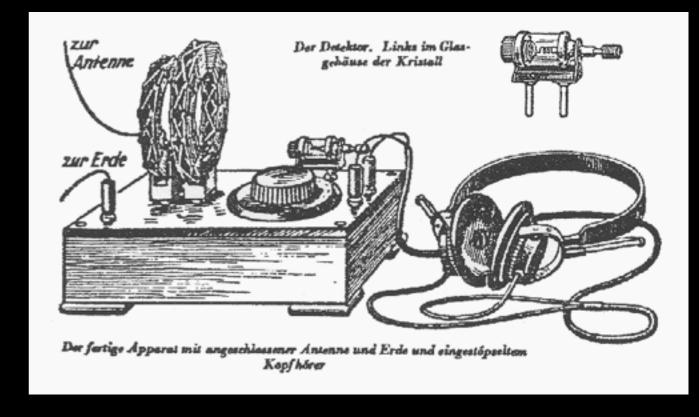


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 1900: <u>Guglielmo Marconi</u> demonstrates wireless telegraph





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- 1900: <u>Marconi</u> demonstrates wireless telegraph
- 1920s: Edwin Howard Armstrong
  - demonstrates FM radio





#### Big Open Questions

- mostly analog
- ad-hoc engineering, tailored to each application
- is there a general methodology for designing communication systems?
- can we communicate reliably in noise?
- how fast can we communicate?



#### Claude Elwood Shannon

1916 - 2001



- Father of Information Theory
- Graduate of MIT 1940:
  - "An Algebra for Theoretical Genetics"
- 1941-1972: Scientist at Bell Labs
- 1958: Professor at MIT:

When he returned to MIT in 1958, he continued to threaten corridor-walkers on his unicycle, sometimes augmenting the hazard by juggling. No one was ever sure whether these activities were part of some new breakthrough or whether he just found them amusing. He worked, for example, on a motorized pogo-stick, which he claimed would mean he could abandon the unicycle so feared by his colleagues ...

- juggling, unicycling, chess
- ultimate machine

- BITS!
- arguably, first to really define and use "bits"
- "He's one of the great men of the century. Without him, none of the things we know today would exist. The whole digital revolution started with him." -Neil Sloane, AT&T Fellow

#### The Bell System Technical Journal

Vol. XXVII

July, 1948

No. 3

A Mathematical Theory of Communication

By C. E. SHANNON



Introduced a new field: Information Theory

What is communication?

What is information?

How much can we compress information?

How fast can we communicate?

## Main Contributions of Inf Theory

#### Source coding

source = randomvariable



ultimate data
 compression limit is the source's entropy H

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#### Channel coding

- channel = conditional distributions
- ultimate transmission rate is the channel capacity C

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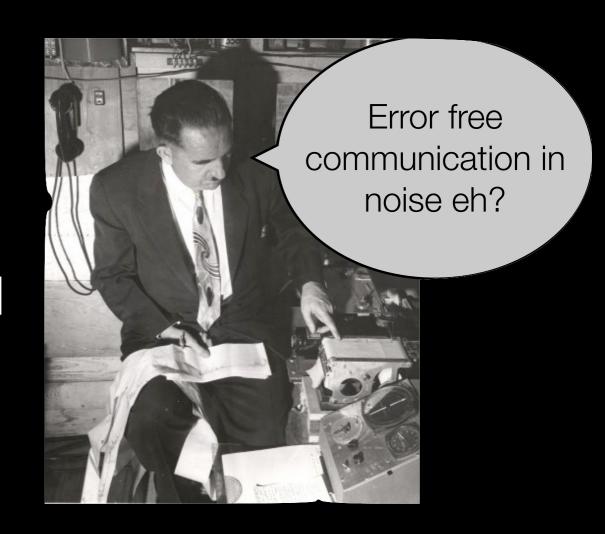
#### Channel coding

- channel = conditional distributions
- ultimate transmission
   rate is the channel
   capacity C

Reliable communication possible  $\iff$  H < C

#### Reactions to This Theory

- Engineers in disbelief
- stuck in analogue world



#### How to approach the predicted limits?

Shannon says: can transmit at rates up to say 4Mbps over a certain channel without error. How to do it?

How to approach the predicted limits?

• 50's: algebraic codes

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- 50's: algebraic codes
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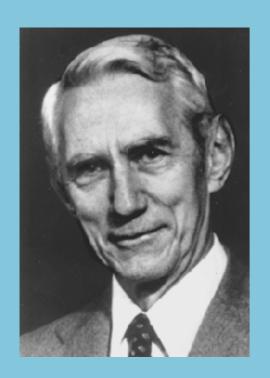
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How to approach the predicted limits?

review article by [Costello Forney 2006]



Claude Shannon — Born on the planet Earth (Sol III) in the year 1916 A.D. Generally regarded as the father of the Information Age, he formulated the notion of channel capacity in 1948 A.D. Within several decades, mathematicians and engineers had devised practical ways to communicate reliably at data rates within 1% of the Shannon limit . . .

Encyclopedia Galactica, 166th ed.

## Applications

- Communication Theory
- Computer Science (e.g. in <u>cryptography</u>)
- Physics (<u>thermodynamics</u>)
- Philosophy of Science (Occam's Razor)
- Economics (investments)
- Biology (genetics, bio-informatics)

## Topics Overview

- Entropy and Mutual Information
- Entropy Diagrams
- Perfectly Secure Encryption
- Data Compression
- Coding Theory
- Channel-Coding Theorem
- Zero-Error Information Theory
- Noisy-Channel Theorem
- Application to Machine Learning

# Questions?

## Example: Letter Frequencies

i	$a_i$	$p_i$		
1	a	0.0575	a	
2	b	0.0128	b	Ы
3	С	0.0263	С	
4	d	0.0285	d	
5	е	0.0913	е	П
6	f	0.0173	f	Ы
7	g	0.0133	g	В
8	h	0.0313	h	П
9	i	0.0599	i	
10	j	0.0006	j	
11	k	0.0084	k	
12	1	0.0335	1	
13	m	0.0235	m	
14	n	0.0596	n	
15	0	0.0689	0	
16	р	0.0192	р	
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	У	0.0164	У	
26	z	0.0007	Z	
27	_	0.1928	_	

Figure 2.1. Probability distribution over the 27 outcomes for a randomly selected letter in an English language document (estimated from *The Frequently Asked Questions Manual for Linux*). The picture shows the probabilities by the areas of white squares.

#### Example: Letter Frequencies

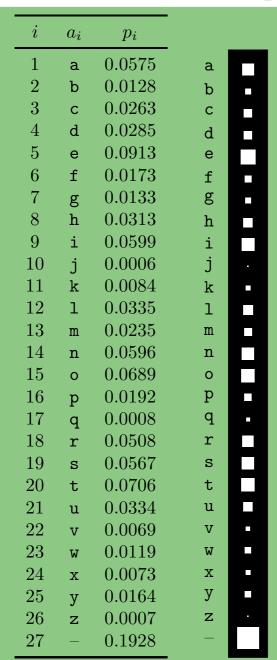
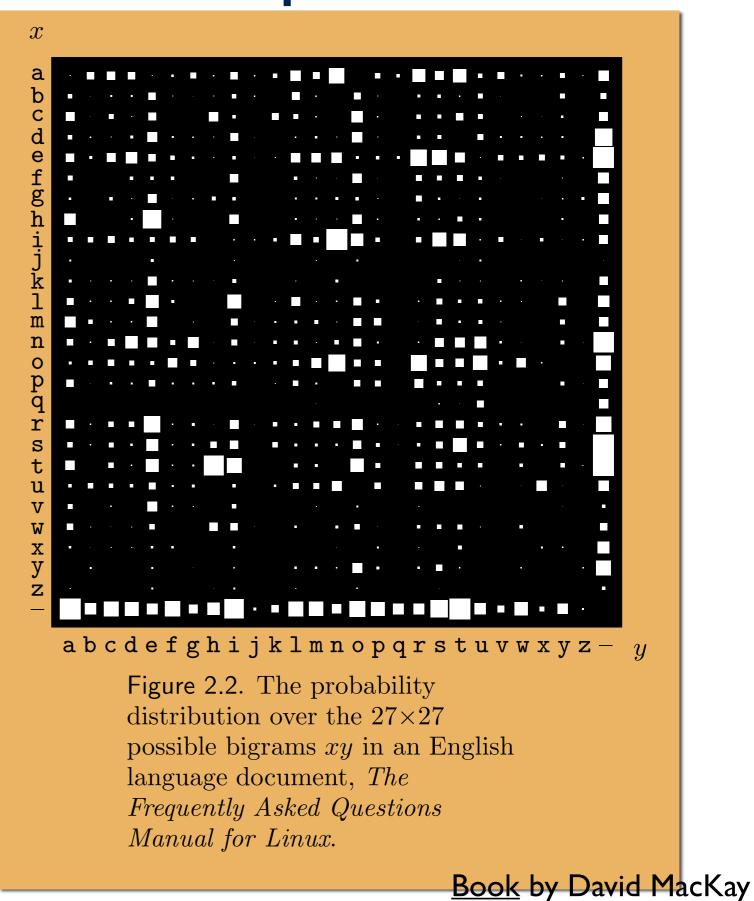


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#### Example: Surprisal Values

#### from http://www.umsl.edu/~fraundorfp/egsurpri.html

situation	probability p = 1/2#bits	surprisal #bits = In <sub>2</sub> [1/p]
one equals one	1	0 bits
wrong guess on a 4-choice question	3/4	In <sub>2</sub> [4/3] ~0.415 bits
correct guess on true-false question	1/2	In <sub>2</sub> [2] =1 bit
correct guess on a 4-choice question	1/4	In <sub>2</sub> [4] =2 bits
seven on a pair of dice	6/6 <sup>2</sup> =1/6	In <sub>2</sub> [6] ~2.58 bits
snake-eyes on a pair of dice	1/6 <sup>2</sup> =1/36	In <sub>2</sub> [36] ~5.17 bits
random character from the 8-bit ASCII set	1/256	In <sub>2</sub> [2 <sup>8</sup> ] =8 bits =1 byte
N heads on a toss of N coins	1/2 <sup>N</sup>	In <sub>2</sub> [2 <sup>N</sup> ] =N bits
harm from a smallpox vaccination	~1/1,000,000	~ln <sub>2</sub> [10 <sup>6</sup> ] ~19.9 bits
win the UK Jackpot lottery	1/13,983,816	~23.6 bits
RGB monitor choice of one pixel's color	1/256 <sup>3</sup> ~5.9×10 <sup>-8</sup>	In <sub>2</sub> [2 <sup>8*3</sup> ] =24 bits
gamma ray burst mass extinction event TODAY!	<1/(10 <sup>9</sup> *365) ~2.7×10 <sup>-12</sup>	hopefully >38 bits
availability to reset 1 gigabyte of random access memory	1/2 <sup>8E9</sup> ~10 <sup>-2.4E9</sup>	8×10 <sup>9</sup> bits ~7.6×10 <sup>-14</sup> J/K
choices for 6×10 <sup>23</sup> Argon atoms in a 24.2L box at 295K	~1/2 <sup>1.61E25</sup> ~10 <sup>-4.8E24</sup>	~1.61×10 <sup>25</sup> bits ~155 J/K
one equals two	0	∞ bits

i	$a_i$	$p_i$	$h(p_i)$	
1	a	.0575	4.1	
2	b	.0128	6.3	
3	С	.0263	5.2	
4	d	.0285	5.1	
5	е	.0913	3.5	
6	f	.0173	5.9	
7	g	.0133	6.2	
8	h	.0313	5.0	
9	i	.0599	4.1	
10	j	.0006	10.7	
11	k	.0084	6.9	
12	1	.0335	4.9	
13	m	.0235	5.4	
14	n	.0596	4.1	
15	0	.0689	3.9	
16	p	.0192	5.7	
17	q	.0008	10.3	
18	r	.0508	4.3	
19	s	.0567	4.1	
20	t	.0706	3.8	
21	u	.0334	4.9	
22	V	.0069	7.2	
23	W	.0119	6.4	
24	х	.0073	7.1	
25	У	.0164	5.9	
26	z	.0007	10.4	
27	-	.1928	2.4	
$\sum_{i} p_i \log_2 \frac{1}{p_i} \qquad 4.1$				

Table 2.9. Shannon information contents of the outcomes a-z.

#### **Book** by David MacKay