# Natural Language Processing

Lecture 05

Dirk Hovy

dirk.hovy@unibocconi.it





# Today's Goals

- Know when (and when not) to use Regular expressions
- Learn about forms of TF-IDF and its possibilities
- Understand how information can be quantified with entropy
- Understand PMI and see why it can help find collocations



# Flexible Matches: Regular Expressions

# The promise...

WHENEVER I LEARN A
NEW SKILL I CONCOCT
ELABORATE FANTASY
SCENARIOS WHERE IT
LETS ME SAVE THE DAY.

OH NO! THE KILLER MUST HAVE POLLOWED HER ON VACATION!



BUT TO FIND THEM WE'D HAVE TO SEARCH THROUGH 200 MB OF EMAILS LOOKING FOR SOMETHING FORMATTED LIKE AN ADDRESS!



IT'S HOPELESS!



















# Is it an (Email) Address?

- notMyFault@webmail.com
- smithie123@gmx
- Free stuff@unibocconi.it
- mark\_my\_words@hotmail;com
- truthOrDare@webmail.in
- look@me@twitter.com
- how2GetAnts@aol.dfdsfgfdsgfd

NAME

@

DOMAIN

. CODE



# Simple Matching

sequence	Matches				
e	any single occurrence of e				
at	<pre>at, rat, mat, sat, cat, attack,   attention, later</pre>				

# Quantifiers

	Means	Example	Matches
*	0 or more	cooo*l	cool, coool
+	1 or more	hello+	hello, helloo, hellooooooo
?	0 or 1	fr?og	fog, frog

# Special Characters

	Means	Example	Matches		
	any single character	.el	eel, Nel, gel		
\n	newline character (line break)	\n+	One or more line breaks		
\t	a tab stop	\t+	One or more tabs		
\d	a single digit [0-9]	B\d	во, в1,, в9		
\D	a non-digit	\D.t	' t, But, eat		
\w	any alphanumberic character	\w\w\w	Top, WOO, ash, bee,		
\W	non-alphanumberic character				
\s	a whitespace character				
\\$	a non-whitespace character				
\	"Escapes" special characters to match them	.+ \.com	abc.com, united.com		
٨	the beginning of the input string	^ • • •	First word in line		
<b>\$</b>	the end of the input string	^\n\$	Empty line		

# Classes

	Means	Example	Matches		
[abc]	Match any of a, b, c	[bcrms]at	bat, cat, rat, mat, sat		
[^abc]	Match anything BUT a, b, c	te[^ ]+s	tens, tests, teens, texts, terrors		
[a-z]	Match any lowercase character	[a-z][a-z]t	act, ant, not, wit		
[A-Z]	Match any uppercase character	[A-Z]	Ahab, Brit, In a,, York		
[0-9]	Match any digit	DIN A[0-9]	DIN A0, DIN A1,, DIN A9		



# Groups

	Means	Example	Matches		
(abc)	Match abc	.(ar).	hard, cart, fare,		
(ab c)	Match ab OR c	(ab c)ate	abate, cate		

# Matching Addresses

NAME @ DOMAIN . CODE

```
^[A-Za-z0-9_\.-]+@[A-Za-z0-9_\.-]+\.[A-Za-z0-9_][A-Za-z0-9_]+$
```



http://xkcd.com/1171/

A (W w) ord of [Ww] arning

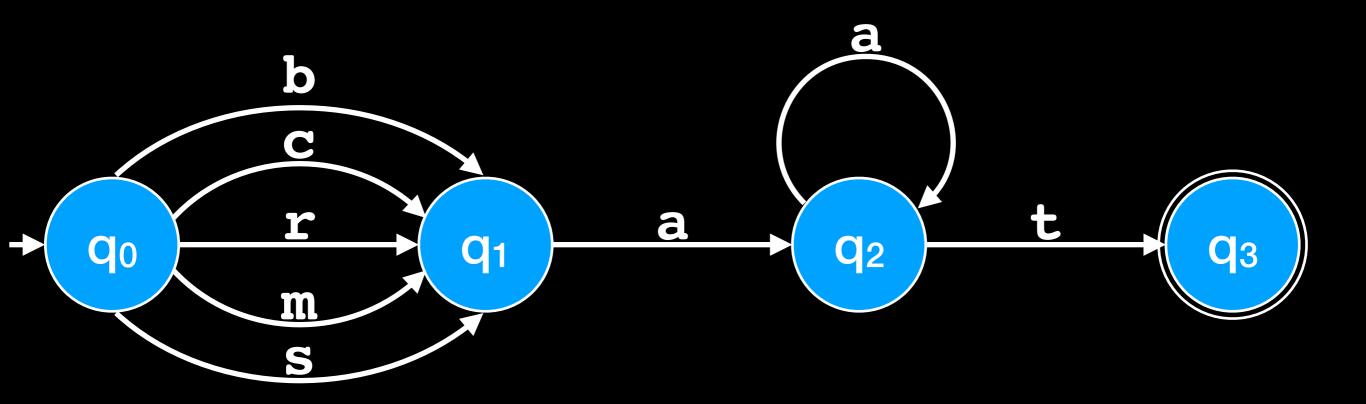


[\t])+\Z|(?=[\["()<>@,;:\\".\[\]]))\\[([^\[\]\\.)\*\](?:(?:\r\n)?[\t])\*)(?:\.(?:(?:\r\n)?[\t])\*(?:[^()<>@,;:\\".\[\]\000-\031]+(?:(?:\r\n)?[\t])+\\Z|(?=[\["()<>@,;:\\".\[\]]))\\[([^\[\]\r\\]\\.)\*\](?:(?:\r\n)?[\t])\*)\(?:[^()<>@,;:\\".\[\]\000-\031]+(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\*\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\\(]([^\[\]\r\n)]\\(?:(?:\r\n)?[\t])\\(?:(?:\r\n)?[\t])\\(]([^\[\]\r\n)]\\(?:(?:\r\n)?[\t])\

日ク<sup>て」) \* ) ) \* ) \* ; (S\*)</sup>

Boccon

# RegEx as Automata

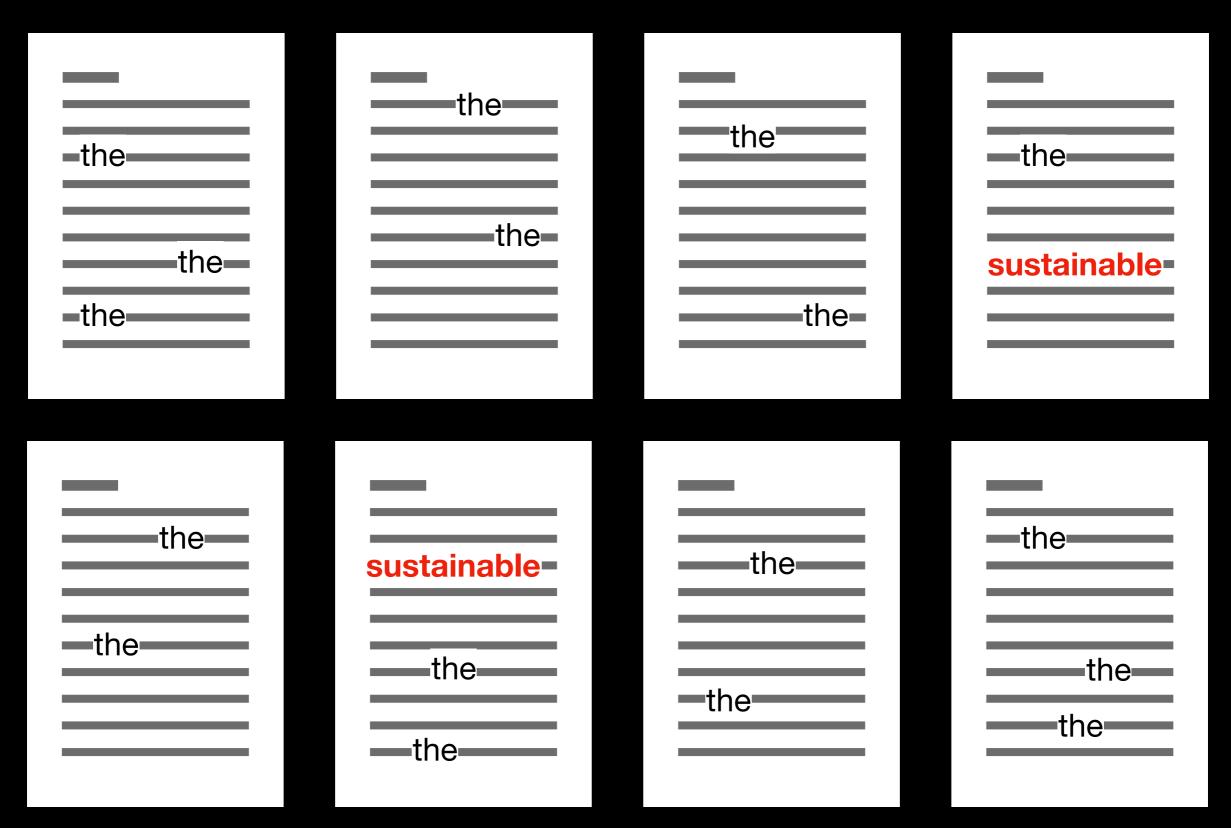


[bcrms]a+t



# Finding Important Words: TF-IDF

#### Some Words are Just More Interesting...





# Karen Spärck Jones

#### 1935-2007

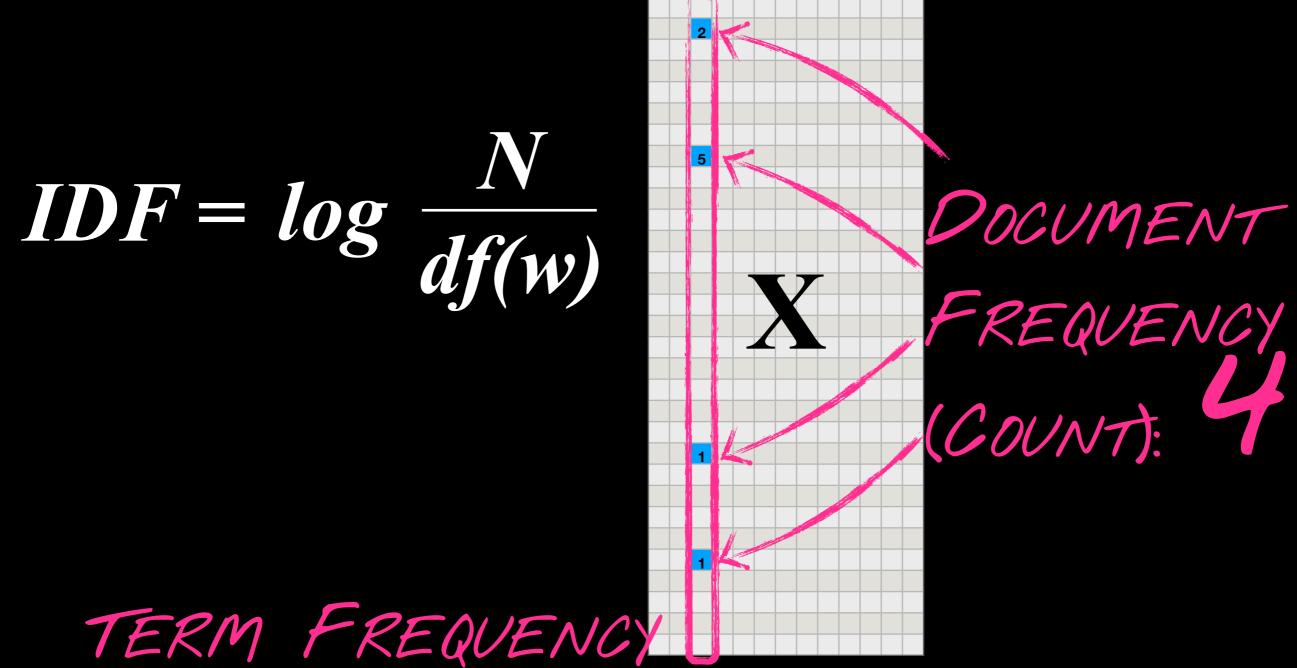
- Became a teacher before starting CS career at Cambridge
- Laid the foundation for modern NLP, Google Search, text classification
- Campaigned for more women in CS
- Namesake of prestigious CS prize



Problems with Term Frequency BORING STUFF 40000 TOLD YA SO ... 30000 20000 JUST RIGHT ??? 10000 OBSCURE STUFF

## Document and Term Frequency





(SUM): 9 TF

# Putting it Together

HOW OFTEN WE SAW THE WORD

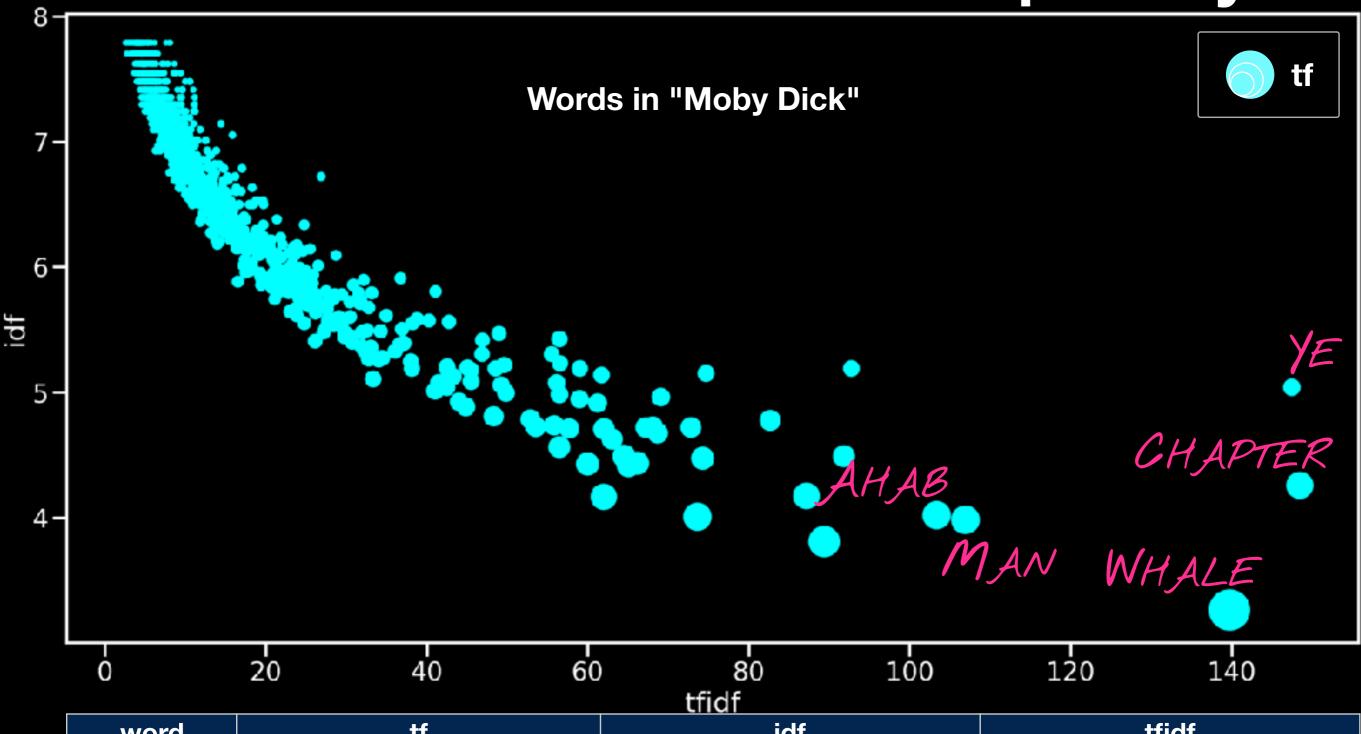
TFIDF(w) = 
$$TF(w) \cdot log \frac{N}{df(w)}$$

ADJUSTED BY

HOW MANY

DOCUMENTS

## Document and Term Frequency



word	tf	idf	tfidf	
ye	467	4.257380	148.497079	
chapter	171	5.039475	147.504638	
whale	1150	3.262357	139.755743	
man	525	3.982412	106.932953	
ahab	511	4.019453	103.357774	

20

# The Probability of Words: Entropy



# What's More Likely?

We finish each others'

SENTENCES

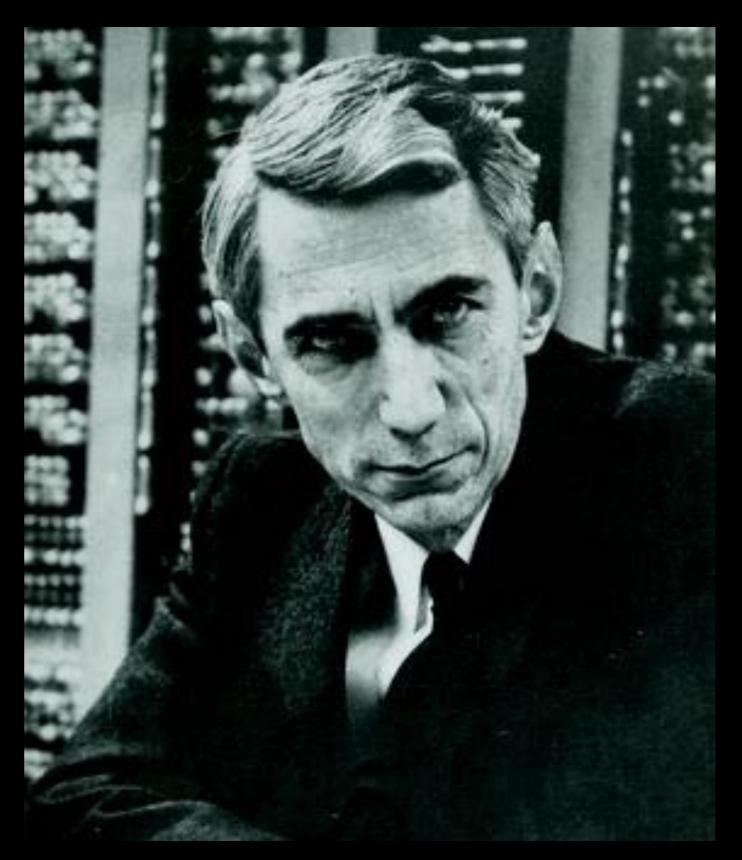
LEFTOVERS



# Claude Shannon

#### 1916-2001

- His master's thesis founded a new field: digital circuits
- Invented entropy to quantify language – and a flamethrowing trumpet
- Enabled NLP, cryptography, modern computers...
- Died of Alzheimer's, oblivious to his own inventions' impact





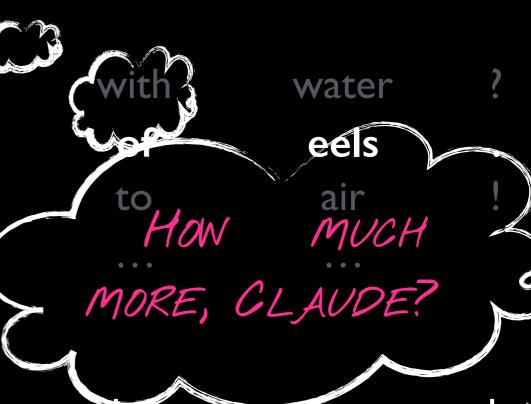
# Shannon Game



WHAT'S THE NEXT WORD?

The house friend Then dog If car When water My hovercraft He pants You God word







entropy

$$H(X) = -\sum_{x} p(x) \log p(x)$$

**Information** 

p(x)

# Probability of a Word

"It must be recognized that the notion 'probability of a sentence' is an entirely useless one, under any known interpretation of this term."

**Noam Chomsky** 

- Choose a word w
- Open a page at random and point at a word: Is it w?

HOW OFTEN WE

HAVE SEEN W

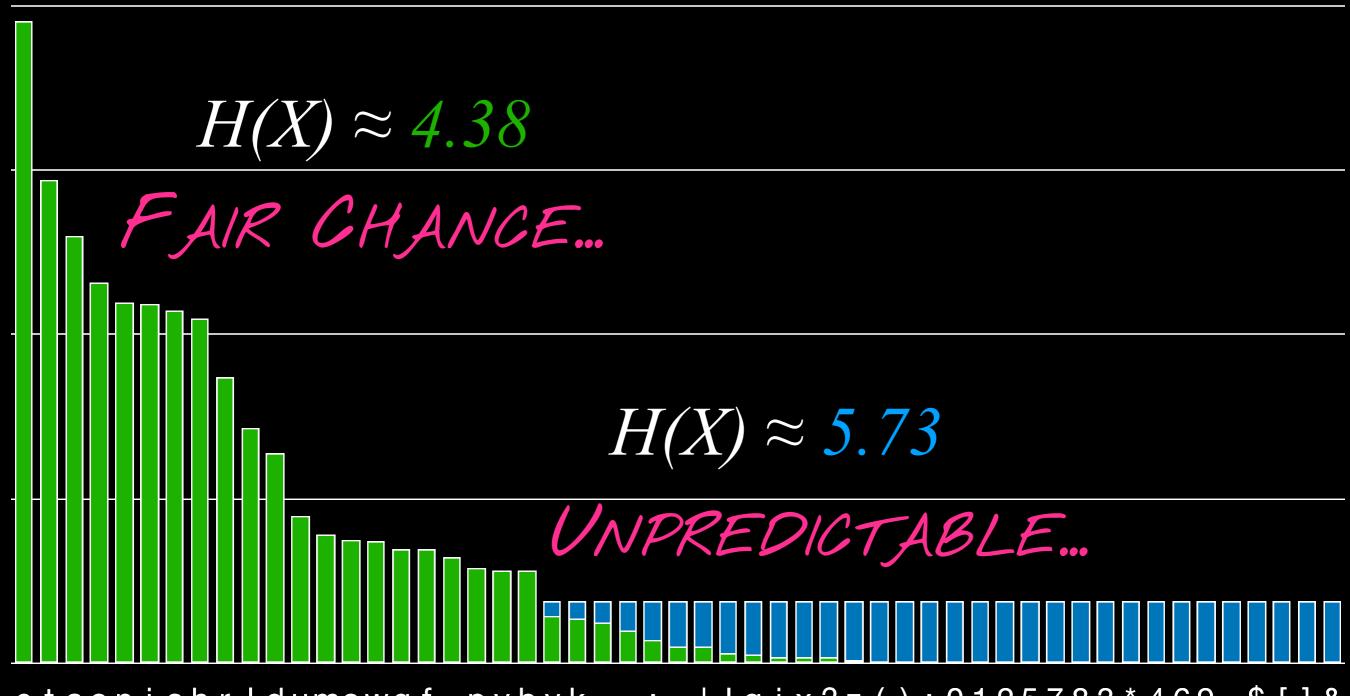
$$C(w)$$

$$P(w) = \frac{\sum c(v)}{\sum w}$$

...ALL WORDS

# Entropy in Use

WHATS THE NEXT LETTER?



etaonishrldumcwgf,pybvk,-; '!qjx?z():0125783\*469\_\$[]&

# Telling Neighbors: Pointwise Mutual Information



### Some are not like the Others



# Mutual Informativity

HOW WELL CAN WE GUESS THE BLANK?

social \_\_\_\_

and

media

the

# Pointwise Mutual Information CHANCE OF SEEING THEM TOGETHER

 $PMI(x, y) = log \underbrace{P(x, y)}_{P(x)P(y)}$   $\underbrace{F(x, y)}_{P(x)P(y)}$   $\underbrace{\underbrace{F(x, y)}_{P(x)P(y)}}_{SEEING EITHER}$ 

X	У	c(x)	c(y)	c(xy)	P(x)	P(y)	P(x, y)	PMI(x; y)
moby	dick	83	83	82	0.0003	0.0003	0.0003	3.48
captain	ahab	327	511	61	0.0013	0.0020	0.0002	1.97
white	whale	280	1150	106	0.0011	0.0045	0.0004	1.93
under	the	119	14175	45	0.0005	0.0553	0.0002	0.83
is	а	1690	4636	110	0.0066	0.0181	0.0004	0.56

c(X) = 256,149c(XY) = 256,148

Bocconi

# Wrapping up

# Take home points

- Regular expressions allow us to search for flexible patterns
- Entropy allows us to quantify how surprising/ predictable something is
- **TF-IDF** finds "bursty" words: medium frequency overall, but concentrated in few documents
- PMI tells us how likely one word is to occur with/ without another to find collocations

