



# Automated Translation: How Does It Work?

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Acknowledgements: with adaptations of materials that appeared at MT Marathons, WMTs, etc.



#### **Machine Translation**



#### Agenda:

- Why MT: Volume, Quality and Cost?
- Why is MT hard?
- MT + Human Translators = Quality
- How does modern statistical MT work?
- Its all about Data!
- And the right kind of Data!





- Europe = Multilinguality
- 24 official languages, 24+2 CEF languages
- So much to translate!
- Translation costs!?
- Can MT help?
- What about the Quality?

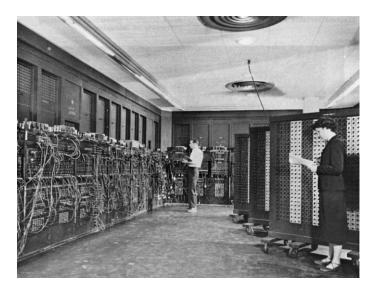


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# Why is MT Hard?



- Human languages are:
  - Elegant
  - Efficient
  - Flexible
  - Complex
- One word/sentence may mean many things
- Many ways of saying the same thing
- Meaning depends on context
- Literal and figurative language (metaphor)
- Language and culture (different ways of conceptualising the same thing)
- Word order
- Morphology

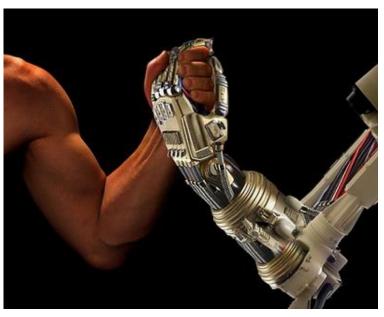


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- Language/translation is complex
- We cannot compute it exactly
- We tried: rule-based MT and LT ...
- What do we do?

European Language

- Machine Learning
  - Learns from data  $\Rightarrow$  data is all important
  - Approximate solution  $\Rightarrow$  not perfect, needs help
    - human professional translators
    - Post-editing
    - Automated Translation ≠ Automatic



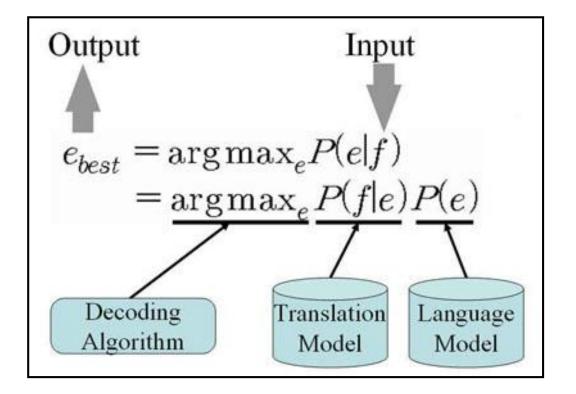






#### How does Modern MT Work?





- No maths today
- Instead:
- The story of Statistical MT in pictures …
- Its all about Data ...



### How does Modern MT Work?



# Statistical MT learns from data Two kinds of data:

- Human translations •
- Text in the target language •
- The more data the better! •
- Also: the right kind of data!

GERMAN	ENGLISH	FRENCH
Einleitung	Introduction	Introduction
I. Von dem Unterschiede der rei- nen und empirischen Erkennt- nis	I. Of the difference between Pure and Empirical Knowledge	I. De la différence de la connais- sance pure et de la connaissance empirique.
Daß alle unsere Erkenntnis mit der Erfahrung anfange, daran ist gar kein Zweifel; denn wo- durch sollte das Erkenntnis- vermögen sonst zur Ausübung erweckt werden, geschähe es nicht durch Gegenstände, die unsere Sinne rühren und teils von selbst Vorstellungen be- wirken, teils unsere Verstan- destätigkeit in Bewegung brin- gen, diese zu vergleichen, sie zu verknüpfen oder zu tren- nen, und so den rohen Stoff sinnlicher Eindrücke zu einer Erkenntnis der Gegenstände zu verarbeiten, die Erfahrung heißt? Der Zeit nach geht also	That all our knowledge begins with experience there can be no doubt. For how is it pos- sible that the faculty of cogni- tion should be awakened into exercise otherwise than by means of objects which affect our senses, and partly of them- selves produce representations, partly rouse our powers of un- derstanding into activity, to compare to connect, or to sep- arate these, and so to convert the raw material of our sensu- ous impressions into a know- ledge of objects, which is called experience? In respect of time, therefore, no know-	Que toute notre connaissance commence avec l'expérience, cela ne soulève aucun doute. En effet, par quoi notre pou- voir de connaître pourrait-il être éveillé et mis en action, si ce n'est par des objets qui frappent nos sens et qui, d'une part, produisent par eux- mêmes des représentations et, d'autre part, mettent en mou- vement notre faculté intellec- tuelle, afin qu'elle compare, lie ou sépare ces représentations, et travaille ainsi la matière brute des impressions sensibles pour en tirer une connaissance des objets, celle qu'on nomme
keine Erkenntnis in uns vor der Erfahrung vorher, und mit dieser fängt alle an.	ledge of ours is antecedent to experience, but begins with it.	l'expérience? Ainsi, chronolo- giquement, aucune connais- sance ne précède en nous l'ex-

périence et c'est avec elle que

toutes commencent.



# What can/do we Learn from Data?



#### Which sentences translate as which: sentence alignment

- Which words translate as which: word alignment + translation probabilities
- What is good target language like: language model

	GERMAN	ENGLISH	FRENCH
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c	der Erfahrung vorher, und mit dieser fängt alle an.	experience, but begins with it.	giquement, aucune connais- sance ne précède en nous l'ex-

périence et c'est avec elle que

toutes commencent.



#### **Sentence Alignment**



TRADUS WinAlign - [C:\\CW_E_H_Ex_wo_me	d.itl: C.\VCW_D_H_Ex_wo_nec	Let (word Documents))					
🔄 Ele Edi Verv Setings Alegement Window Help							
B-M COUNCW_E_H_Ex_wo_mod.rtf 	nn	B-B C0\CW_D_H_Ex_wo_mod.rtf └─B CWatch					
This brings you to the calendar 🔺 where you can begin to enter	BB	Dadurch gelangen Sie zum 🔺 Kalender, in dem alle weiteren					
data.		Eingaben erfolgen.					
After indicating your first cycle date in the dialog box, a calendar appears.	••	Nachden Sie Ihren ersten Zyklus in das Dialogfenster eingegeben haben erscheint eine Kalenderanzicht, die den Monat					
		Three ersten Zyklus wiedergibt.					
It indicates the month of your first cycle date.	•	Der Tag, den Sie als ersten Zyklustag eingegeben haben, wird in der rechten unteren Ecke des Kalenderfensters durch *g1 und *g2 dargestellt.					
The day you indicated as the first cycle day will be marked with a ^gl in the lower right hand corner and a ^g2.	•	Yalls dies nicht der erste Tag Ihres ersten Zyklus ist, wählen Sie bitte "Datei", "Schließen", und "Wicht Speichern".					
If this is not the day you want to indicate as the first cycle day, go to file, Close, and	• •	Yühren Sie dann bitte erneut einen Doppelklick auf das CycleWatch Exblem aus und					
don't save it.		korrigieren das Datum.					
Then double-click on the CycleWatch icon, enter the correct date, and click on OK.	•	On die Eingabe zu beztätigen klicken Sie bitte auf OK.					
You will also see two numbers in each calendar day.	PhPh	Es verden jeveils ovei Sahlen pro Kalendertag dargestellt.					
The succes is the usper right and Ready		NUM					



#### Word Alignment:



					CLASSIC SOUPS Sm.	Lg.
清	嫩	鶔		57.	House Chicken Soup (Chicken, Celery,	
					Potato, Onion, Carrot)1.50	2.75
雞	1	Æ		58.	Chicken Rice Soup 1.85	3.25
雞	3	•		59.	Chicken Noodle Soup1.85	3.25
曆	東	\$	呑	60.	Cantonese Wonton Soup1.50	2.75
*	茄	*		61.	Tomato Clear Egg Drop Soup1.65	2.95
雪	4	5	2	62.	Regular Wonton Soup 1.10	2.10
酸	争	*		63.	₹● Hot & Sour Soup	2.10
푷				64.	Egg Drop Soup1.10	2.10
雲	1	F		65.	Egg Drop Wonton Mix1.10	2.10
료	窟	莱		66.	Tofu Vegetable SoupNA	3.50
雞	Æ	米		67.	Chicken Corn Cream SoupNA	3.50
2	肉王	E *		68.	Crab Meat Corn Cream SoupNA	3.50
海	1	¥		69.	Seafood SoupNA	3.50



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- Word alignment mode knows a lot about Chinese soups
- Doesn't know much else …
- Only knows what it has seen in the training data
- Like people ...
- A common theme ...
- Given word aligned translation data, can we learn a translation dictionary?
- Yes, really easy ...





love the boy.
J'aime le garçon.
love the dog.
J'aime le chien.
They love the dog.
Ils aiment le chien.
They talk to the girl.
Ils parlent à la fille.
They talk to the dog.
Ils parlent au chien.
I talk to the mother.
Je parle à la mère.

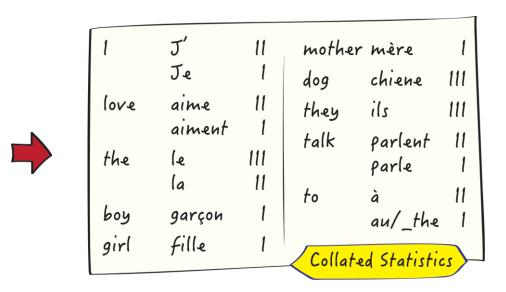
Aligned Data





l love the boy. J'aime le garçon. l love the dog. J'aime le chien. They love the dog. Ils aiment le chien. They talk to the girl. Ils parlent à la fille. They talk to the dog. Ils parlent au chien. I talk to the mother. Je parle à la mère.

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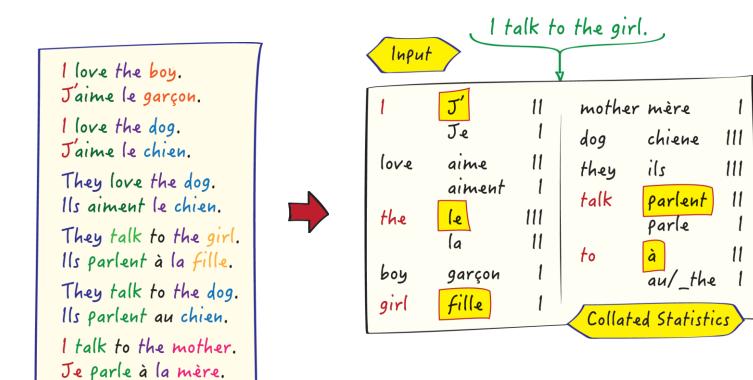




Aligned Data

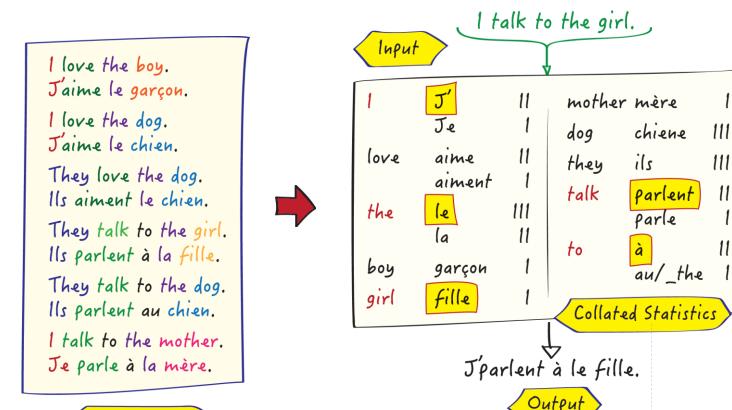
#### **Statistical Machine Translation**















l love the boy. J'aime le garçon. l love the dog. J'aime le chien. They love the dog. Ils aiment le chien. They talk to the girl. Ils parlent à la fille. They talk to the dog. Ils parlent au chien. I talk to the mother. Je parle à la mère.

Aligned Data

1	talk	to	the	girl			
<b>J'</b>	parlen	it	<mark>аи</mark>	<mark>le fille</mark>			
2/3	2/3		2/3	3/5 1/1			
<mark>Je</mark>	parle	<mark>à</mark>	<mark>la</mark>	<mark>fille</mark>			
1/3	1/3	1/3	2/5	1/1			
How to choose?							





l love the boy. J'aime le garçon. l love the dog. J'aime le chien. They love the dog. Ils aiment le chien. They talk to the girl. Ils parlent à la fille. They talk to the dog. Ils parlent au chien. I talk to the mother. Je parle à la mère.

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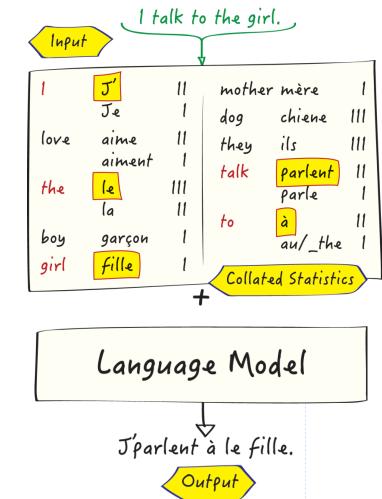
#### The Language Model:

- What is good target language?
- Which words can follow which words and which can't ... the grammar
- Learnt from the data ...
  - Je parle is good ...
  - J' parlent is bad ...
  - *la fille* is good ...
  - le fille is bad ...
- Je parle à la fille >> J' parlent à le fille





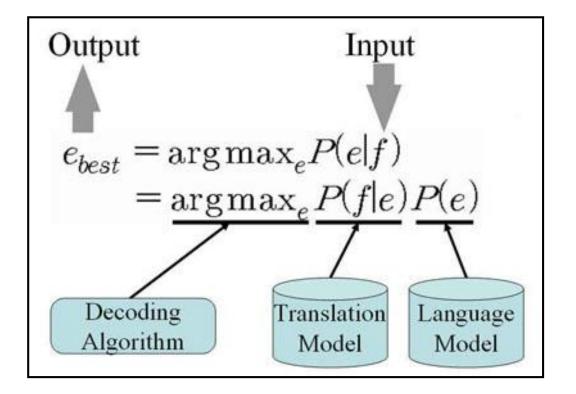
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#### How does Modern MT Work?





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- Instead:
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- Its all about Data ...

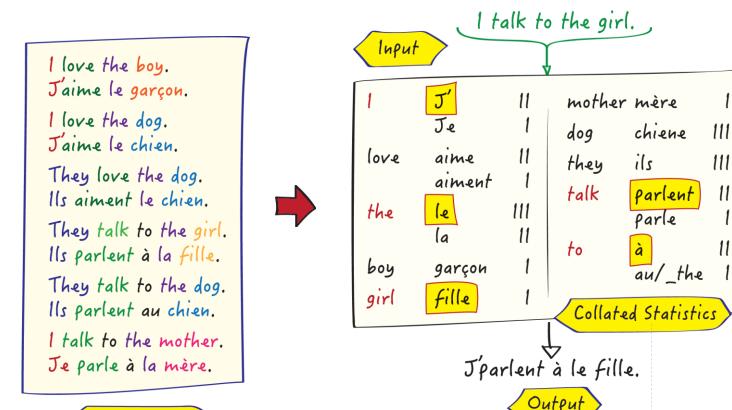




- So far: translating single words
- Loses context: such as agreement (*le fille …*) etc.
- To some extent "repaired" by language model
- A better model:
- Not just translations of single words
- But also phrase translations:
  - the girl : la filleto the girl : a la fille
  - I talk : Je parle





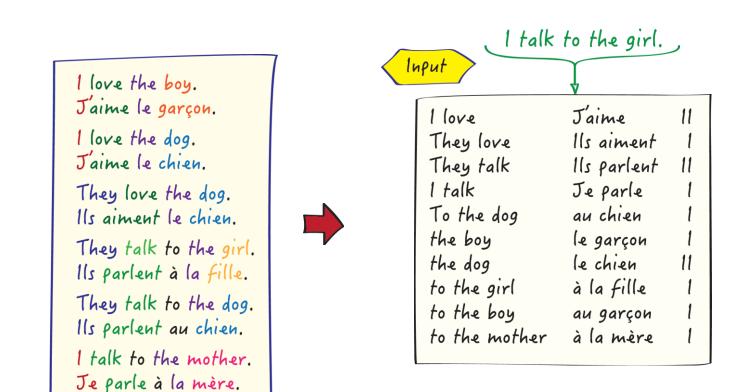




Aligned Data

#### Phrase Based - Statistical Machine Translation

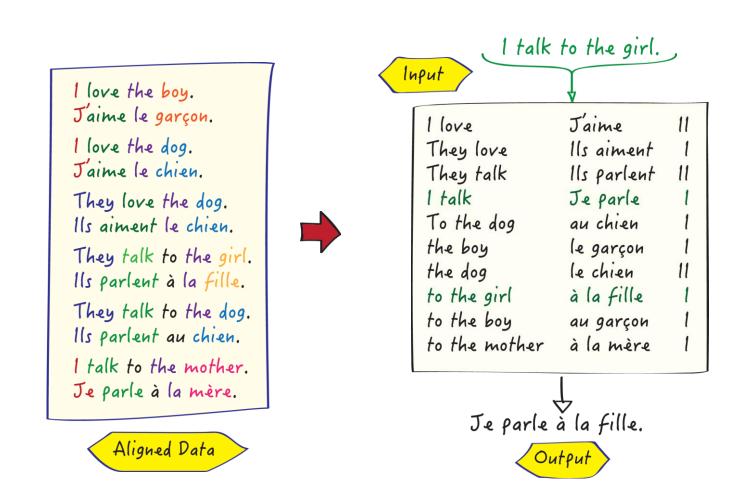






#### Phrase Based - Statistical Machine Translation







# Phrase Based - Statistical Machine Translation



- Much better than word-based SMT!
- Standard technology: Google, Microsoft, Baidu, Global Localisation & Translation Industry
- Moses Open Source PB-SMT
- Most widely used SMT system
- Research funded by EC
- Used by EC DGT's MT@EC





# **Machine Translation and Data**

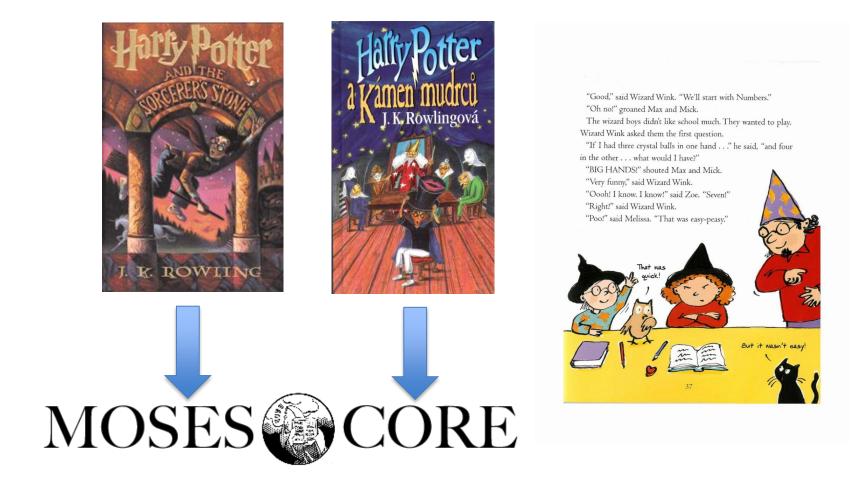


- Statistical Machine Translation is all about data
- SMT learns how to translate from data
- Data
  - translations (bilingual data)
  - Monolingual data (target language text)
  - Dictionaries, terminology, ontologies, named entities
- Like people SMT is good at what it has learned



#### **Machine Translation and Data**

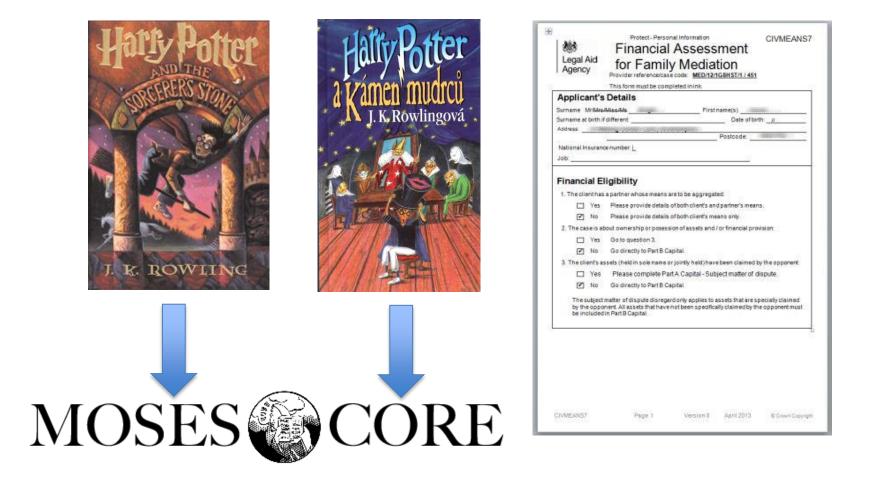






#### **Machine Translation and Data**









- CEF.AT needs the right kind of data
- National governments, public administration, public services, NGOs
- CEF provide services for multilingual engagement with national citizens, EU citizens and other customers of public administration







- Help us make CEF.AT a success
  - Services for Europe's citizens
  - Services for you
  - Support multi-linguality
- Help us find the right kind of data
- Supporting our language is supporting Europe and vice versa