CS 188: Artificial Intelligence

Language

Pieter Abbeel – UC Berkeley Slides from Dan Klein

What is NLP?





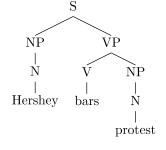
- Fundamental goal: analyze and process human language, broadly, robustly, accurately...
- End systems that we want to build:
 - Ambitious: speech recognition, machine translation, information extraction, dialog interfaces, question answering...
 - Modest: spelling correction, text categorization...

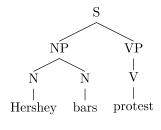
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Problem: Ambiguities

- Headlines:
 - Enraged Cow Injures Farmer With Ax
 - Hospitals Are Sued by 7 Foot Doctors
 - Ban on Nude Dancing on Governor's Desk
 - Iraqi Head Seeks Arms
 - Local HS Dropouts Cut in Half
 - Juvenile Court to Try Shooting Defendant
 - Stolen Painting Found by Tree
 - Kids Make Nutritious Snacks
- Why are these funny?

Parsing as Search



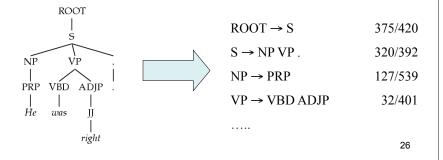


Hershey bars protest

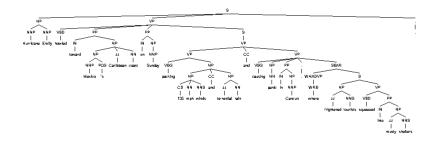
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Grammar: PCFGs

- Natural language grammars are very ambiguous!
- PCFGs are a formal probabilistic model of trees
 - Each "rule" has a conditional probability (like an HMM)
 - Tree's probability is the product of all rules used
- Parsing: Given a sentence, find the best tree search!



Syntactic Analysis



Hurricane Emily howled toward Mexico's Caribbean coast on Sunday packing 135 mph winds and torrential rain and causing panic in Cancun, where frightened tourists squeezed into musty shelters.

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Machine Translation





- Translate text from one language to another
- Recombines fragments of example translations
- Challenges:
 - What fragments? [learning to translate]
 - How to make efficient? [fast translation search]



面

摘心

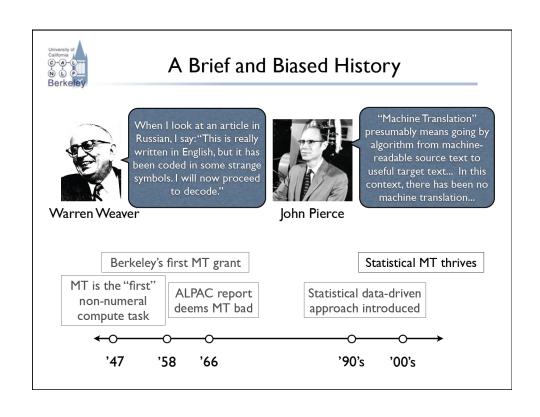
The Problem with Dictionary Look-ups

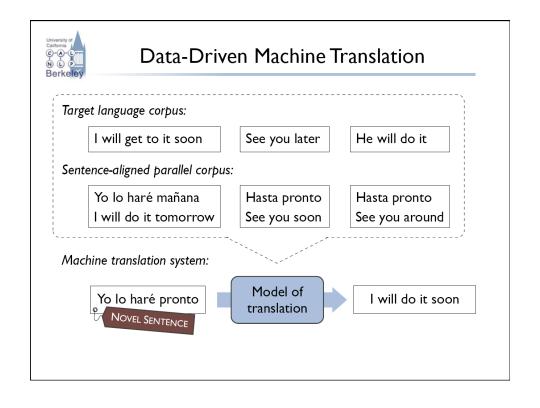
语 页	/top/roof/
顶端	/summit/peak/ top /apex/
顶头	/coming directly towards one/top/end/
盖	/lid/ top /cover/canopy/build/Gai/
盖帽	/surpass/ top /
极	/extremely/pole/utmost/top/collect/receive/
尖峰	/peak/top/

/top/topping/

Example from Douglas Hofstadter

/fade/side/surface/aspect/top/face/flour/



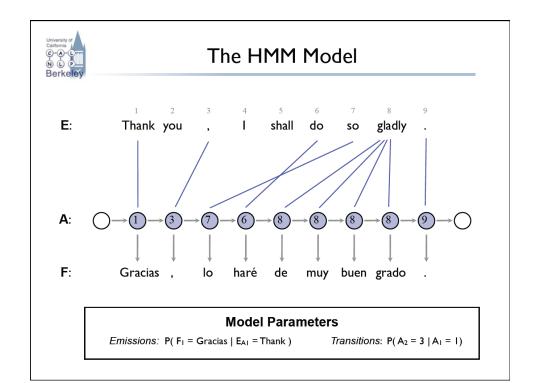


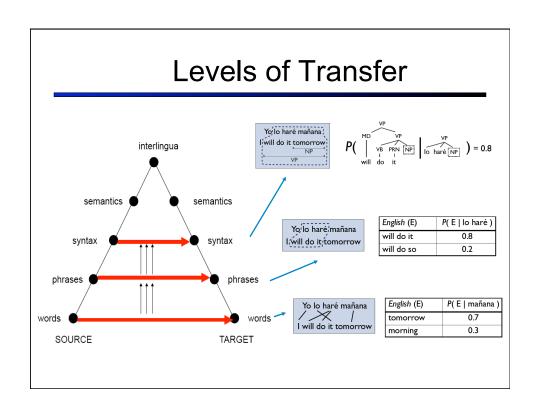


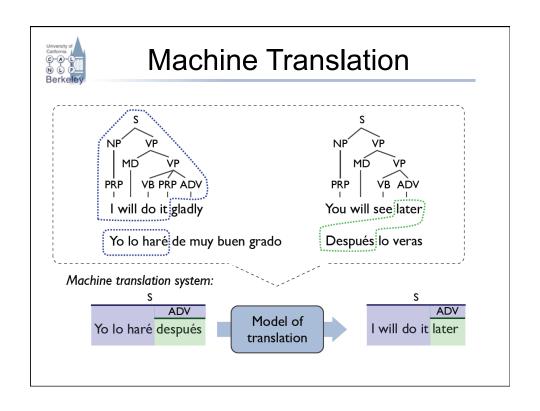
Learning to Translate

					CLASSIC SOUPS	Sm.	Lg.
力	燉	雞	10	57.	House Chicken Soup (Chicken, Celery,		
					Potato, Onion, Carrot)	1.50	2.75
雞	Ê	Ŕ	20	58.	Chicken Rice Soup		3.25
雞	奏	<u> </u>	\$	59.	Chicken Noodle Soup	1.85	3.25
廣	東	亚	吞	60.	Cantonese(Wonton)Soup		2.75
番	茄	\$	9	61.	Tomato Clear Egg Drop Soup	1.65	2.95
雪	2	Ś	\$	62.	Regular (Wonton) Soup		2.10
酸	乳	東	*	63. ₹	Hot & Sour Soup		2.10
₹	7	£		64.	Egg Drop Soup		2.10
李	7	-	*	65.	Egg Drop(Wonton)Mix	1.10	2.10
豆豆		茱	-	66.	Tofu Vegetable Soup	NA	3.50
雞	玉	米	当	67.	Chicken Corn Cream Soup		3.50
A24	肉王	:米	書	68.	Crab Meat Corn Cream Soup		3.50
海	4	¥	*	69.	Seafood Soup		3.50

Example from Adam Lopez









A Statistical Translation Model

Synchronous Derivation

Yo lo haré después

S ADV I will do it later

Synchronous Grammar Rules

 $S \rightarrow \langle Yo lo haré ADV ; I will do it ADV \rangle$ ADV $\rightarrow \langle después ; later \rangle$

A Statistical Model

Translation model components factor over applied rules

How well are these rules supported by the data?

Language model factors over n-grams

How well is this output sentence supported by the data?

