Adam Lopez University of Edinburgh → Johns Hopkins University

### 虽然北风呼啸,但天空依然十分清澈。

### 虽然北风呼啸,但天空依然十分清澈。



### The Tower of Babel

Pieter Brueghel the Elder (1563)



Georges Artsrouni's "mechanical brain", patented 1933 (France)



## ENIAC (1946)



When I look at an article in Russian, I say: "This is really written in English, but it has been coded in some strange symbols. I will now proceed to decode."

Warren Weaver (1949)



## 4/28/2006 03:40:00 PM Franz Och

Because we want to provide everyone with access to all the world's information, including information written in every language, one of the exciting projects at Google Research is machine translation... Now you can see the results for yourself. We recently launched an online version of our system for Arabic-English and English-Arabic. Try it out!



### Language Tools

#### Translated search

Type a search phrase in your language. Google will find results in other languages and translate them for you to read.

Search for:		Translate and Search
	Search pages written in:       My langua         ⊙ Automatically selected languages       Englis         ⊙ Specific languages       Englis	ige: h▼
Example:	<ol> <li>Search for <u>Bern tourist information</u>.</li> <li>We translate your query into French and German, and find French and German 3. Finally, we translate the French and German results back into your language.</li> </ol>	results.
Translat	e text	
Bienvenu	e à Le Mans	

French

English
 English

+)(



### Language Tools

#### Translated search

Type a search phrase in your language. Google will find results in other languages and translate them for you to read.

Search for:	121	Translate and Search
English	Degge written in:	210.00
Estonian	pages written in:	iguage:
Filipino	omatically selected languages	nglish 🔻
Finnish	cific languages	
French		
Galician	ch for Bern tourist information.	
German	anslate your query into French and German, and find French and German	nan results.
Greek	ly, we translate the French and German results back into your language	
Haitian Creole		
Hebrew		
Hindi		
Hungarian		
Icelandic	Mans	
Indonesian		
Irish		
Italian		
Japanese		
Korean		
Latvian	A	
Lithuanian	<b>V</b>	
French	English Translate	



### Language Tools

#### Translated search

Type a search phrase in your language. Google will find results in other languages and translate them for you to read.

Search for:			Translate and Search
Estonian	pages written in:	My language:	
Filipino	omatically selected languages	English v	
Finnish	cific languages		
French	Cinc languages		
Galician	ch for Bern tourist information.		
German	ranslate your query into French and German, a	and find French and German resul	lts.
Greek	ly, we translate the French and German result	s back into your language.	
Haitian Creole			
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Icelandic	Mans		
Indonesian			
Irish			
Italian			
Japanese			
Korean			
Latvian	▲		
Lithuanian	¥		
French	English Translate		

## 2756 language pairs!













### 虽然北风呼啸,但天空依然十分清澈。

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。

 Although north wind howls , but sky still very clear .

 虽然 北 风 呼啸 , 但 天空 依然 十分 清澈 。





















Despite the strong northerly winds , the sky remains very clear .



Despite the strong northerly winds , the sky remains very clear . The sky was still crystal clear , though the north wind was howling . Although a north wind was howling , the sky remained clear and blue .



Despite the strong northerly winds , the sky remains very clear . The sky was still crystal clear , though the north wind was howling . Although a north wind was howling , the sky remained clear and blue .



1010人のようには、「「「「「「「」」」」。

WAA TEITE AT HILAA HILEPEIT OA NY ETEINAAN TALKA AN F HAA TEITE AT HILAA HILEPEIT OA NY ETEILAAN TALKA AN F HATA TALEATA AN MENOY F TO TO Y & UA GEOTENIA AN THIAN TO TALEATA AN MENOY F TO TO Y & UA GEOTENIA AN THIAN HATA TO Y LAT LAELA NTATE ON ENAVERATION TALY MATTER HATA TO Y LAT LAELA NTATE ON ENAVERATION TALY MATTER HATA TO Y LAT LAELA NTATE ON TO Y & ANALY TALY EIKE MEISTA IEPAA PFYPIKAE TE KANTITINA TO PATA HATA TALEATA TALEATA INA TALEATA IIA TALEATA HATA TALEATA NIEPEA PATA TALEATA IIA TALEATA HATA TE PITA NIEPEA PATA TALEATA IIA TALEATA HATA TALEATA NATA TALEATA ILA TALEATA HATA TALEATA NATA TALEATA INA TALEATA HATA TALEATA NATA TALEATA INA TALEATA HATA TALEATA NATA TALEATA ILA TALEATA HATA TALEATA NATA TALEATA INA TALEATA HATA TALEATA NATA TALEATA INA TALEATA HATA TALEATA INA ANTA TALEATA TALEATA HATA TALEATA I TALEATA TALEATA TALEATA HATA TALEATA INA ANTA TALEATA TALEATA HATA TALEATA TALEATA TALEATA TALEATA TALEATA HATA TALEATA TALEATA TALEATA TALEATA TALEATA HATA TALEATA TALEATA TALEATA TALEATA TALEATA TALEATA HATA TALEATA TALE










			CLASSIC SOUPS	Sm.	Lg.
燉雞		57.	House Chicken Soup (Chicken, Celery,		
			Potato, Onion, Carrot)	1.50	2.75
飯	-	58.	Chicken Rice Soup	1.85	3.25
麵	*	59.	Chicken Noodle Soup	1.85	3.25
東雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
茄蛋	*	61.	Tomato Clear Egg Drop Soup	1.65	2.95
る	湯	62.	Regular Wonton Soup	1.10	2.10
辣	湯	63. 🍋	Hot & Sour Soup	1.10	2.10
it	*	64.	Egg Drop Soup	1.10	2.10
\$	湯	65.	Egg Drop Wonton Mix	1.10	2.10
腐 菜	*	66.	Tofu Vegetable Soup	NA	3.50
玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
肉玉米	湯	68.	Crab Meat Corn Cream Soup	NA	3.50
鮮	*	69.	Seafood Soup	NA	3.50
	烟 飯麵 東茄 腐玉肉雞 飯麵 季茄 香辣花蛋 玉鲜雞	燉 飯麵 東茄 香辣花蛋 腐玉肉雞 飯麵 霉蛋 香辣花蛋 玉鲜雞 渴渴者香湯渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴渴	燉 雞 湯 57. 燉 雞 湯 58. 飯 麵 雪 雪 香 60. 東 斎 春 60. 東 斎 春 62. 本 湯 湯 63. 本 湯 湯 64. 富 菜 米 米 番 65. 本 米 米 番 65. 本 本 米 米 番 65. 本 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 本 春 65. 本 本 本 春 65. 本 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 本 春 65. 本 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 春 65. 本 本 本 春 65. 本 本 春 65. 本 春 7. 本 春 7. 春	広にASSIC SOUPS         燉雞湯 57.       House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)         飯湯 58.       Chicken Rice Soup         麵湯 59.       Chicken Noodle Soup         麵 36.       59.         蔥 零 60.       Cantonese Wonton Soup         茄 雪 湯 61.       Tomato Clear Egg Drop Soup         茄 雪 湯 63.       Hot & Sour Soup         蒜 63.       Hot & Sour Soup         菜 湯 65.       Egg Drop Soup         膏 湯 65.       Egg Drop Soup         葦 湯 66.       Tofu Vegetable Soup         玉 米 湯 67.       Chicken Corn Cream Soup         戽 玉 米 湯 68.       Crab Meat Corn Cream Soup         鲜 湯 69.       Seafood Soup	皮雞 雞 湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)1.50飯 湯 58.Chicken Rice Soup1.85麵 湯 59.Chicken Noodle Soup1.85蔥 湯 60.Cantonese Wonton Soup1.50茄 審 湯 61.Tomato Clear Egg Drop Soup1.65杏 湯 62.Regular Wonton Soup1.10辣 湯 63.Hot & Sour Soup1.10菜 湯 65.Egg Drop Soup1.10菜 湯 65.Egg Drop Soup1.10素 湯 65.Egg Drop Soup1.10素 湯 65.Egg Drop Soup1.10素 湯 65.Egg Drop Soup1.10素 湯 66.Tofu Vegetable SoupNA五 米 湯 67.Chicken Corn Cream SoupNA鲜 湯 69.Seafood SoupNA

			CLASSIC SOUPS	Sm.	Lg.
燉雞	*	57.	House Chicken Soup (Chicken, Celery,		
			Potato, Onion, Carrot)	. 1.50	2.75
飯		58.	Chicken Rice Soup	1.85	3.25
麵	*	59.	Chicken Noodle Soup	1.85	3.25
東雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
茄蛋	*	61.	Tomato Clear Egg Drop Soup	1.65	2.95
る	湯	62.	Regular Wonton Soup	. 1.10	2.10
辣	湯	63. 🍋	Hot & Sour Soup	. 1.10	2.10
花	*	64.	Egg Drop Soup	.1.10	2.10
\$	湯	65.	Egg Drop Wonton Mix	1.10	2.10
腐 菜	*	66.	Tofu Vegetable Soup	NA	3.50
玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
肉玉米	湯	68.	Crab Meat Corn Cream Soup	NA	3.50
鮮	*	69.	Seafood Soup	NA	3.50
	烟 的复数 电茄 医盖皮外 的复数 医胆子 医外子 医胆子 医丁二氏 医丁二氏 医马马斯 化丁二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化	燉 飯麵 東茄 香辣花蛋 腐玉肉雞 飯麵 雪蛋 香辣花蛋 玉鲜雞 温温香香浸活浸浸清清清清清清清清清清清清清清清清清清清清清清清清清清清清清清清清	燉 雞 湯 57. 燉 雞 湯 58. 蔥 麵 雲 雪 香 60. 素 3 湯 香 61. ● 3 湯 香 第 63. ■ 3 湯 3 湯 3 湯 3 湯 3 湯 3 湯 3 64. ■ 3 湯 3 湯 3 湯 3 湯 3 湯 3 湯 3 64. ■ 3 湯 3 64. ■ 4 65. ■ 58. ■ 59. ■	上CLASSIC SOUPS燉 雞 湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)飯 湯 58.Chicken Rice Soup麵 湯 59.Chicken Noodle Soup麵 湯 59.Chicken Noodle Soup蔥 零 杏 60.Cantonese Wonton Soup茄 審 湯 61.Tomato Clear Egg Drop Soup茄 審 湯 63.Hot & Sour Soup※ 湯 63.Hot & Sour Soup※ 湯 65.Egg Drop Soup☞ 湯 65.Egg Drop Wonton Mix☞ 湯 66.Tofu Vegetable Soup玉 米 湯 67.Chicken Corn Cream Soup戶 玉 米 湯 68.Crab Meat Corn Cream Soup鲜 湯 69.Seafood Soup	広LASSIC SOUPSSm.燉雞湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)1.50飯 湯 58.Chicken Rice Soup1.85麵 湯 59.Chicken Noodle Soup1.85蔥 湯 61.Tomato Clear Egg Drop Soup1.65杏 湯 62.Regular Wonton Soup1.10辣 湯 63.Hot & Sour Soup1.10菜 湯 65.Egg Drop Soup1.10菜 湯 65.Egg Drop Soup1.10菜 湯 65.Egg Drop Soup1.10素 湯 65.Egg Drop Soup1.10素 湯 65.Egg Drop Soup1.10素 湯 65.Egg Drop SoupNA五 米 湯 67.Chicken Corn Cream SoupNA鲜 湯 69.Seafood SoupNA

			CLASSIC SOUPS	Sm.	Lg.
燉 雞		57.	House Chicken Soup (Chicken, Celery,		
			Potato, Onion, Carrot)	1.50	2.75
飯		58.	Chicken Rice Soup	1.85	3.25
麵	*	59.	Chicken Noodle Soup	1.85	3.25
東 雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
茄香	*	61.	Tomato Clear Egg Drop Soup	1.65	2.95
る	湯	62.	Regular Wonton Soup	1.10	2.10
辣	湯	63. 🍋	Hot & Sour Soup	1.10	2.10
祀	*	64.	Egg Drop Soup	1.10	2.10
ቻ	湯	65.	Egg Drop Wonton Mix	1.10	2.10
腐 菜	*	66.	Tofu Vegetable Soup	NA	3.50
玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
肉玉米	湪	68.	Crab Meat Corn Cream Soup	NA	3.50
鮮	-	69.	Seafood Soup	NA	3.50
	烟 的复数 电茄 医盖皮外 的复数 医胆子 医外子 医胆子 医丁二氏 医丁二氏 医马马斯 化丁二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化二乙基 化	燉 飯麵 東茄 香辣花蛋 腐玉肉雞 飯麵 雪蛋 香辣花蛋 玉鲜雞 温温香香浸活浸浸清清清清清清清清清清清清清清清清清清清清清清清清清清清清清清清清	燉 雞 湯 57. 燉 雞 湯 58. 飯 麵 雪 雪 59. 東 湯 湯 60. 東 湯 湯 60. 東 斎 湯 湯 63. 湯 湯 64. 富 菜 米 米 湯 65. 本 米 湯 65. 本 米 湯 65. 本 米 米 湯 65. 本 米 湯 65. 本 米 湯 65. 本 米 湯 65. 本 米 湯 65. 本 湯 65. 本 米 湯 65. 本 湯 65. 本 湯 65. 本 米 湯 65. 本 米 湯 65. 本 湯 65. 本 湯 65. 本 湯 65. 本 米 湯 65. 本 3 6. 本 5. 本 3 6. 本 5. 本 3 6. 本 5. 本 3 6. 本 5. 本 5. 本 5. 本 3. む	上CLASSIC SOUPS燉雞湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)飯湯 58.Chicken Rice Soup麵湯 59.Chicken Noodle Soup麵湯 59.Chicken Noodle Soup蔥 零 60.Cantonese Wonton Soup茄 零 湯 61.Tomato Clear Egg Drop Soup茄 零 湯 63.Hot & Sour Soup菜 湯 63.Hot & Sour Soup荒 湯 64.Egg Drop Soup夏 Drop SoupEgg Drop Soup玉 湯 65.Egg Drop Soup玉 米 湯 67.Chicken Corn Cream Soup萬 玉 湯 68.Crab Meat Corn Cream Soup鲜 湯 69.Seafood Soup	広LASSIC SOUPSSm.燉雞湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)1.50飯 湯 58.Chicken Rice Soup1.85麵 湯 59.Chicken Noodle Soup1.85蔥 湯 61.Tomato Clear Egg Drop Soup1.65杏 湯 62.Regular Wonton Soup1.10辣 湯 63.Hot & Sour Soup1.10菜 湯 65.Egg Drop Soup1.10菜 湯 65.Egg Drop Soup1.10意 菜 湯 65.Egg Drop Soup1.10意 菜 湯 65.Egg Drop Soup1.10意 菜 湯 66.Tofu Vegetable SoupNA五 米 湯 67.Chicken Corn Cream SoupNA鲜 湯 69.Seafood SoupNA

				CLASSIC SOUPS	Sm.	Lg.
清	燉雞	***	57.	House Chicken Soup (Chicken, Celery,		
				Potato, Onion, Carrot)	. 1.50	2.75
雞	飯	-	58.	Chicken Rice Soup	. 1.85	3.25
雞	麵	*	59.	Chicken Noodle Soup	. 1.85	3.25
廣	東雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
蕃	茄蛋	-	61.	Tomato Clear Egg Drop Soup	. 1.65	2.95
霯	る	湯	62.	Regular Wonton Soup	. 1.10	2.10
歐	辣	湯	63. 🍋	Hot & Sour Soup	.1.10	2.10
靀	1E	*	64.	Egg Drop Soup	.1.10	2.10
÷	Ŧ	**	65.	Egg Drop Wonton Mix	.1.10	2.10
豆	腐菜	*	66.	Tofu Vegetable Soup	NA	3.50
雞	玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
	肉玉米	湯	68.	Crab Meat Corn Cream Soup	NA	3.50
海	鮮	*	69.	Seafood Soup	NA	3.50

				CLASSIC SOUPS	Sm.	Lg.
清	燉雞	***	57.	House Chicken Soup (Chicken, Celery,		
				Potato, Onion, Carrot)	. 1.50	2.75
雞	飯	-	58.	Chicken Rice Soup	. 1.85	3.25
雞	麵	*	59.	Chicken Noodle Soup	. 1.85	3.25
廣	東雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
蕃	茄蛋	-	61.	Tomato Clear Egg Drop Soup	. 1.65	2.95
霯	る	湯	62.	Regular Wonton Soup	. 1.10	2.10
歐	辣	湯	63. 🍋	Hot & Sour Soup	.1.10	2.10
靀	1E	*	64.	Egg Drop Soup	.1.10	2.10
÷	Ŧ	**	65.	Egg Drop Wonton Mix	.1.10	2.10
豆	腐菜	*	66.	Tofu Vegetable Soup	NA	3.50
雞	玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
	肉玉米	湯	68.	Crab Meat Corn Cream Soup	NA	3.50
海	鮮	*	69.	Seafood Soup	NA	3.50

			CLASSIC SOUPS	Sm.	Lg.
燉雞		57.	House Chicken Soup (Chicken, Celery,		
			Potato, Onion, Carrot)	1.50	2.75
飯	-	58.	Chicken Rice Soup	1.85	3.25
麵	*	59.	Chicken Noodle Soup	1.85	3.25
東 雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
茄蛋	*	61.	Tomato Clear Egg Drop Soup	1.65	2.95
る	湯	62.	Regular Wonton Soup	1.10	2.10
辣	湯	63. 🍋	Hot & Sour Soup	.1.10	2.10
7E	湯	64.	Egg Drop Soup	1.10	2.10
Ŧ	湯	65.	Egg Drop Wonton Mix	1.10	2.10
腐菜	*	66.	Tofu Vegetable Soup	NA	3.50
玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
肉玉米	湯	68.	Crab Meat Corn Cream Soup	NA	3.50
●羊	-	69.	Seafood Soup	NA	3.50
	烟 飯麵 東茄 香辣花蛋 腐玉肉雞 飯麵 雪蛋 香辣花蛋 美米米	燉 飯麵 東茄 香辣花蛋 腐玉肉雞 飯麵 霉蛋 香辣花蛋 羔羊	燉 雞 湯 57. 燉 雞 湯 58. 飯 麵 雪 雪 59. 東 斎 春 60. 東 斎 春 61. 本 湯 湯 62. 本 湯 湯 63. 本 湯 湯 65. 本 湯 湯 65. 本 米 米 湯 65. 本 米 湯 65. 本 米 米 湯 65. 本 米 湯 65. 本 4. 本 湯 65. 本 米 湯 65. 本 4. 本 5. 本 5. 5. 本 5. 本	版 雞 湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)飯 湯 58.Chicken Rice Soup麵 湯 59.Chicken Noodle Soup蔥 湯 61.Tomato Clear Egg Drop苏 審 湯 61.Tomato Clear Egg Drop添 着 62.Regular Wonton Soup辣 湯 63.Hot & Sour Soup菜 湯 65.Egg Drop夏 DropSoup葉 湯 65.Egg Drop夏 X 湯 66.Tofu Vegetable Soup玉 米 湯 67.Chicken Corn Cream Soup鮮 湯 69.Seafood Soup	皮雞雞湯57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)1.50飯湯58.Chicken Rice Soup1.85麵湯59.Chicken Noodle Soup1.85東雪各60.Cantonese Wonton Soup1.50茄蛋湯61.Tomato Clear Egg Drop Soup1.65香湯62.Regular Wonton Soup1.10辣湯63.Hot & Sour Soup1.10菜湯65.Egg Drop Soup1.10菜湯65.Egg Drop Soup1.10素湯65.Egg Drop Soup1.10素湯65.Egg Drop Soup1.10素素湯66.Tofu Vegetable SoupNA点米湯68.Crab Meat Corn Cream SoupNA鲜湯69.Seafood SoupNA

				CLASSIC SOUPS	Sm.	Lg.
清	燉雞		57.	House Chicken Soup (Chicken, Celery,		0.05
				Potato, Onion, Carrot)	. 1.50	2.75
雞	飯	-	58.	Chicken Rice Soup	1.85	3.25
雞	麵	*	59.	Chicken Noodle Soup	1.85	3.25
廣	東雲	呑	60.	Cantonese Wonton Soup	1.50	2.75
퐇	茄蛋	**	61.	Tomato Clear Egg Drop Soup	1.65	2.95
雲	呑	湯	62.	Regular Wonton Soup	1.10	2.10
酸	辣	湯	63. 🍋	Hot & Sour Soup	.1.10	2.10
<del>ም</del>	祀	湯	64.	Egg Drop Soup	1.10	2.10
雲	ቻ	湯	65.	Egg Drop Wonton Mix	1.10	2.10
፱	腐 菜	湯	66.	Tofu Vegetable Soup	NA	3.50
雞	玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
봫	肉玉米	*	68.	Crab Meat Corn Cream Soup	NA	3.50
海	鮮	*	69.	Seafood Soup	NA	3.50

				CLASSIC SOUPS	Sm.	Lg.
눍	燉雞	-	57.	House Chicken Soup (Chicken, Celery,		
				Potato, Onion, Carrot)	. 1.50	2.75
雞	飯	: <mark>0</mark>	58.	Chicken Rice Soup	. 1.85	3.25
雞	麵	湯	59.	Chicken Noodle Soup	. 1.85	3.25
廣	東雲	呑	60.	Cantonese Wonton Soup	. 1.50	2.75
퐇	茄蛋	湯	61.	Tomato Clear Egg Drop Soup	. 1.65	2.95
霯	る	湯	62.	Regular Wonton Soup	. 1.10	2.10
酸	辣	湯	63. 🍋	Hot & Sour Soup	. 1.10	2.10
Ŧ	祀	湯	64.	Egg Drop Soup	. 1.10	2.10
雲	7	湯	65.	Egg Drop Wonton Mix	. 1.10	2.10
፹	腐菜	**	66.	Tofu Vegetable Soup	NA	3.50
雞	玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
臀	肉玉米	**	68.	Crab Meat Corn Cream Soup	NA	3.50
海	鮮	-	69.	Seafood Soup	NA	3.50

			CLASSIC SOUPS	Sm.	Lg.
燉雞	*	57.	House Chicken Soup (Chicken, Celery,		
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る	*	62.	Regular Wonton Soup	. 1.10	2.10
辣	*	63. 🍋	Hot & Sour Soup	.1.10	2.10
祀	*	64.	Egg Drop Soup	. 1.10	2.10
委	*	65.	Egg Drop Wonton Mix	.1.10	2.10
腐 菜	-	66.	Tofu Vegetable Soup	NA	3.50
玉米	湯	67.	Chicken Corn Cream Soup	NA	3.50
肉玉米	*	68.	Crab Meat Corn Cream Soup	NA	3.50
鮮	*	69.	Seafood Soup	NA	3.50
	烟 飯麵 東茄 香辣花蛋 腐玉肉雞 飯麵 雲蛋 香辣花蛋 玉鲜	燉 飯麵 東茄 香辣花蛋 腐玉肉雞 鼠麵 雲蛋 香辣花蛋 菜米米丁 马莱米米米 马马马马马马马马马马马马马马马马马马马马马马马马马马马马马马	<ul> <li>燉雞湯 57.</li> <li>燉雞湯 58.</li> <li>汤麵雪雪 59.</li> <li>枣茄 60.</li> <li>枣茄 60.</li> <li>枣茄 61.</li> <li>香辣花 56.</li> <li>香辣花 56.</li> <li>高玉 366.</li> <li>玉米 366.</li> <li>玉米 366.</li> <li>玉米 366.</li> <li>玉米 366.</li> <li>玉米 366.</li> </ul>	広口名SSIC SOUPS         燉 雞 湯 57.       House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)         飯 湯 58.       Chicken Rice Soup         麵 湯 59.       Chicken Rice Soup         蔥 湯 59.       Chicken Noodle Soup         蔥 雪 湯 61.       Tomato Clear Egg Drop Soup         亦 雪 湯 62.       Regular Wonton Soup         ※ 湯 63. №       Hot & Sour Soup         花 湯 64.       Egg Drop Soup         夏 子 湯 65.       Egg Drop Soup         玉 米 湯 67.       Chicken Corn Cream Soup         肉 玉 米 湯 68.       Crab Meat Corn Cream Soup	広LASSIC SOUPSSm.燉雞湯 57.House Chicken Soup (Chicken, Celery, Potato, Onion, Carrot)1.50飯 湯 58.Chicken Rice Soup1.85麵 湯 59.Chicken Noodle Soup1.85蔥 湯 61.Tomato Clear Egg Drop Soup1.65杏 湯 62.Regular Wonton Soup1.10辣 湯 63.Hot & Sour Soup1.10花 湯 64.Egg Drop Soup1.10花 湯 65.Egg Drop Soup1.10花 湯 66.Tofu Vegetable Soup1.10皮 菜 湯 67.Chicken Corn Cream SoupNA萬 玉 湯 68.Crab Meat Corn Cream SoupNA鲜 湯 69.Seafood SoupNA

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# A CELRONAL

p(heads)

1 - p(heads)



#### p(heads)?





 $p(data) = p(heads)^7 \times p(tails)^3$ 



#### $p(data) = p(heads)^7 \times [1 - p(heads)]^3$



p(heads)







## Optimization





#### p(heads)?

#### p(heads)?



However , the sky remained clear under the strong north wind .

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。

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This is a *latent variable* problem

More on learning from parallel data: Today's lab Tomorrow's lecture

#### Statistical Machine Translation

Develop a statistical *model* of translation that can be *learned* from *data* and used to *predict* the correct English translation of new Chinese sentences.

### Statistical Machine Translation

#### Develop a statistical *model* of translation that can be *learned* from *data* and used to *predict* the correct English translation of new Chinese sentences.
#### training data (parallel text)













#### What's a model?

#### What's a model?

For our purposes, a model will be **a probability distribution over sentence pairs**.

 Access to techniques developed and proven over hundreds of years that work on many problems.

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• Common sense in mathematical form!



The probabilities of all possible events must sum to 1.

p(e) = 1 $e \in E$ 



When an event consists of observations about more than one variable, it is a *joint probability*.



 $p(A = 1, B = 1) = \frac{2}{42}$ 



 $p(1,1) = \frac{2}{42}$ 



A probability distribution over a subset of variables is a *marginal probability*.



 $p(B = 1) = p(\cdot, 1) = \sum_{a \in A} p(A = a, B = 1) = \frac{1}{6}$ 



$$p(A = 1) = p(1, \cdot) = \sum_{b \in B} p(A = 1, B = b) = \frac{1}{6}$$



The probability of a variable under the condition that the other variables are fixed is the *conditional probability*.



 $p(B = 1|A = 1) = \frac{p(A = 1, B = 1)}{\sum_{b \in B} p(A = 1, B = b)} = \frac{2}{7}$ 



$$p(B=1|A=1) = \frac{p(A=1,B=1)}{\sum_{b\in B} p(A=1,B=b)} = \frac{2}{7} \qquad \frac{\text{joint}}{\text{marginal}}$$



We can still represent the joint distribution as a product of other distributions.



p(A = 1, B = 1) = p(A = 1, B = 1)



 $p(A = 1, B = 1) = \sum_{b \in B} p(A = 1, B = b) \frac{p(A = 1, B = 1)}{\sum_{b \in B} p(A = 1, B = b)}$ 



 $p(A = 1, B = 1) = p(A = 1) \cdot p(B = 1 | A = 1)$ 



 $p(A, B) = p(A) \cdot p(B|A)$ 

 $p(A,B) = p(A) \cdot p(B|A)$ 

 $p(A, B) = p(A) \cdot p(B|A) = p(B) \cdot p(A|B)$ 

 $p(A) \cdot p(B|A) = p(B) \cdot p(A|B)$ 

 $p(B|A) = \frac{p(B) \cdot p(A|B)}{p(A)}$ 

Bayes' Rule

# $p(B|A) = \frac{p(B) \cdot p(A|B)}{p(A)}$
#### Probabilistic Primer



Bayes' Rule

...But the probability that an event has happened is the same as the probability I have to guess right if I guess it has happened. Wherefore the following proposition is evident: If there be two subsequent events, the probability of the 2d b/N and the probability both together *P*/*N*, and it being 1st discovered that the 2d event has also happened, the probability I am right is P/b.



Thomas Bayes

configuration



configuration

p(image|English)

configuration

p(image|English)

# configuration





p(DNA)

#### p(mutation|DNA)

p(DNA)







However, the sky remained clear under the strong north wind .

However, the sky remained clear under the strong north wind .

p(Chinese|English)

However, the sky remained clear under the strong north wind .

p(Chinese|English)

虽然北风呼啸,但天空依然十分清澈。



When I look at an article in Russian, I say: "This is really written in English, but it has been coded in some strange symbols. I will now proceed to decode."

Warren Weaver (1949)



#### THE MATHEMATICAL THEORY OF COMMUNICATION

by Claude E. Shannon and Warren Weaver



#### Claude Shannon

p(English|Chinese) =

 $\frac{p(English) \times p(Chinese | English)}{p(Chinese)}$ prior likelihood

### Noisy Channel

p(English|Chinese) =

 $\frac{p(English) \times p(Chinese|English)}{p(Chinese)}$ signal model channel model

p(English|Chinese) =

 $\begin{array}{c|c} p(English) \times p(Chinese | English) \\ \hline & & & \\ \hline & & & \\ p(Chinese) \\ \hline & & & \\ language \ model \end{array} \qquad \begin{array}{c} p(Chinese | English) \\ \hline & & \\ f \end{array} \qquad \begin{array}{c} & \\ f \end{array} \end{array} \qquad \begin{array}{c} & \\ f \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array}$  \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \end{array} \end{array} \end{array} \qquad \begin{array}{c}

#### p(Chinese|English)

### 

#### English





p(English|Chinese) =

 $\begin{array}{c|c} p(English) \times p(Chinese | English) \\ \hline & & & \\ \hline & & & \\ p(Chinese) \\ \hline & & & \\ language \ model \end{array} \qquad \begin{array}{c} p(Chinese | English) \\ \hline & & \\ f \end{array} \qquad \begin{array}{c} & \\ f \end{array} \end{array} \qquad \begin{array}{c} & \\ f \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array}$  \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \end{array} \qquad \begin{array}{c} & f \end{array} \end{array} \end{array} \end{array} \end{array} \end{array} \qquad \begin{array}{c}

 $p(English|Chinese) \sim$ 

 $p(English) \times p(Chinese|English)$ 

 $p(English|Chinese) \sim$ 

 $p(English) \times p(Chinese|English)$ 

What is the probability of an English sentence?

 $p(English|Chinese) \sim$ 

#### $p(English) \times p(Chinese|English)$

#### What is the probability of an English sentence?

What is the probability of a Chinese sentence, given a particular English sentence?

We can think of our probabilistic model as a story that explains every single word in the sentence pair.

p(However|START)

However

p(However|START)

However,

p(, |However)

However, the

p(the|,)

However, the sky

p(sky|the)

However, the sky remained

p(remained|sky)

However, the sky remained clear

p(clear | remained)

# $p(English) = \prod_{i=1}^{length(English)} p(word_i|word_{i-1})$
# $p(English) = \prod_{i=1}^{length(English)} p(word_i | word_{i-1})$

Note: the prior probability that word<sub>0</sub>=START is 1.

$$p(English) = \prod_{i=1}^{length(English)} p(word_i | word_{i-1})$$

Note: the prior probability that word<sub>0</sub>=START is 1. This model explains every word in the English sentence.

$$p(English) = \prod_{i=1}^{length(English)} p(word_i | word_{i-1})$$

Note: the prior probability that word<sub>0</sub>=START is 1. This model explains every word in the English sentence. But it makes very strong conditional independence \_\_\_\_\_\_assumptions!

- The language model does not depend in any way on parallel data.
- How much English data should we train it on?

39

**G00g** 

Impact on size of language model training data (in words) on quality of Arabic-English statistical machine translation system (NIST test data)



41

GOC

Impact on size of language model training data (in words) on quality of Arabic-English statistical machine translation system



43

**J**O

Impact on size of language model training data (in words) on quality of Arabic-English statistical machine translation system



45

**GO** 

Impact on size of language model training data (in words) on quality of Arabic-English statistical machine translation system



#### 47 Goog Language Models



- There's no data like more data.
- Language models serve a similar function in speech recognition, optical character recognition, and other probabilistic models of text data.
- You'll learn a lot more about them from Nicola Bertoldi on Tuesday and Wednesday

What is a good story about how a Chinese sentence came into being, given that we already have an English sentence?

#### What is a good story about how a Chinese sentence came into being, given that we already have an English sentence?

Note: in this example I'll show you an English sentence, conditioned on a Chinese sentence. Note that we can apply the same technique in either direction.

虽然北风呼啸,但天空依然十分清澈。

p(English|Chinese)

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。

p(English|Chinese)

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。

However , the sky remained clear under the strong north wind .

#### p(English|Chinese)



However , the sky remained clear under the strong north wind .



However, the sky remained clear under the strong north wind.

#### p(English|Chinese)?

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。 人 虽然

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。 人 虽然

 $p_f(1|$ 虽然)



but sky Although north wind howls, still clear very 天空 依然 呼啸 清澈 十分 北风 虽然 但 , ,天空天空依然清澈 风 呼啸 北 虽然

 Although north wind howls , but sky still very clear .

 虽然 北 风 呼啸 , 但 天空 依然 十分 清澈 。

  $\downarrow$   $\downarrow$ 
 $\downarrow$   $\downarrow$ 





 $p_t(However|$ 虽然)

 Although north wind howls , but sky still very clear .

 虽然 北 风 呼啸 , 但 天空 依然 十分 清澈 。  $\varepsilon$  

 」 」 」 」 」 」 」 、 」 」 \_ / / / / / / /

 虽然 北 风 呼啸 , 天空 天空 依然 清澈 。  $\varepsilon \varepsilon$  

 」 】 】 】 】 、 / / / / / / / / / / / / / /

 日本地区 中啸 , 天空 天空 依然 清澈 。  $\varepsilon \varepsilon$ 

However north wind strong , the sky remained clear . under the

Although north wind howls, but sky still very clear 虽然北风呼啸,但天空依然十分清澈。 $\varepsilon$ 虽然北风呼啸,天空天空依然清澈。 $\varepsilon \varepsilon$ However north wind strong, the sky remained clear. under the

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 $p_d(0|However)$ 





 $p_d(8|north)$ 





However , the sky remained clear under the strong north wind .
# IBM Model 4



However , the sky remained clear under the strong north wind .  $p(English, alignment | Chinese) = \prod_{p_f} \prod_{p_t} \prod_{p_d}$ 

### IBM Model 4

虽然北风呼啸,但天空依然十分清澈。

However , the sky remained clear under the strong north wind .  $p(English, alignment | Chinese) = \prod_{p_f} \prod_{p_t} \prod_{p_d}$ 

### IBM Model 4

虽然北风呼啸,但天空依然十分清澈。

However, the sky remained clear under the strong north wind .  $p(English|Chinese) = \sum_{alignments} \prod_{p_f} \prod_{p_t} \prod_{p_d} \prod_{p_d} p_d$ 

• Fertility probabilities.

- Fertility probabilities.
- Word translation probabilities.

- Fertility probabilities.
- Word translation probabilities.
- Distortion probabilities.

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- Some problems:
  - Weak reordering model -- output is not fluent.
  - Many decisions -- many things can go wrong.

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- Word translation probabilities.
- Distortion probabilities.
- Some problems:
  - Weak reordering model -- outp
  - Many decisions -- many things



CONCISE English-Chinese Chinese-English DICTIONARY

- Fertility probabilities.
- Word translation probabilities.
- Distortion probabilities.
- Some problems:
  - Weak reordering model -- output is not fluent.
  - Many decisions -- many things can go wrong.

# IBM Model 4 Again



However , the sky remained clear under the strong north wind .

# IBM Model 4 Again



However , the sky remained clear under the strong north wind .

Although north wind howls, but sky still very clear. 虽然北风呼啸,但天空依然十分清澈。















 $p(English, alignment|Chinese) = \\p(segmentation) \cdot p(translations) \cdot p(reorderings)$ 

Segmentation probabilities.

- Segmentation probabilities.
- Phrase translation probabilities.

- Segmentation probabilities.
- Phrase translation probabilities.
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and Pinwinsthnoughout

- Segmentation probabilities.
- Phrase translation probabilities.
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- Some problems:
  - Weak reordering model -- output is not fluent.
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#### Overview



#### Overview



# Decoding

Probability models allow us to make predictions: Given a particular Chinese sentence, what is the most probable English sentence corresponding to it?

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#### In math:

 $\operatorname{argmax}_{English} p(English|Chinese)$ 

# Decoding

Probability models allow us to make predictions: Given a particular Chinese sentence, what is the most probable English sentence corresponding to it?

#### In math:

 $\operatorname{argmax}_{English} p(English|Chinese)$ 

problem: there are a lot of English sentences to choose from!





# Optimization

#### Overview



#### 北风呼啸。
segmentations substitutions permutations

segmentations $O(2^n)$ substitutionspermutations

segmentations substitutions

 $O(2^n)$  $O(5^n)$ 

permutations

segmentations $O(2^n)$ substitutions $O(5^n)$ permutationsO(n!)

segmentations $O(2^n)$ substitutions $O(5^n)$ permutationsO(n!)

240,000 possibilities!

#### the strong north wind .





Given a sentence pair and an alignment, we can easily calculate p(English, alignment|Chinese)



Given a sentence pair and an alignment, we can easily calculate p(English, alignment|Chinese)

Can we do this without enumerating  $O(10^n n!)$  pairs?

#### 北风呼啸。

#### the strong north wind .



#### 北风呼啸。 北风呼啸。 the strong north wind.

There are  $O(10^n n!)$  target sentences.

### 北风呼啸。 一 the strong north wind.

There are  $O(10^n n!)$  target sentences. But there are only  $O(5n^2)$  ways to start them.













coverage vector







coverage vector



### $p(north|START) \cdot p(\exists \forall | north)$ north

coverage vector













北风呼啸



### $p(north|START) \cdot p(\exists \forall | north)$ north

coverage vector



#### 

coverage vector

### $p(north|START) \cdot p(\exists \natural | north)$ north wind $p(wind|north) \cdot p(\exists \land | wind)$

coverage vector





Work done at sentence beginnings is shared across many possible output sentences!























Dynamic Programming
## Key Idea

#### amount of work:





#### amount of work: $O(5n^22^n)$

#### bad, but much better than $O(10^n n!)$



Key Idea

#### amount of work: $O(5n^22^n)$

bad, but much better than  $O(10^n n!)$ 

each edge labelled with a weight and a

Key Idea

word (or words)

#### amount of work: $O(5n^22^n)$

bad, but much better than  $O(10^n n!)$ 



Key Idea



## Weighted languages

- The lattice describing the set of all possible translations is a *weighted finite state automaton*.
- So is the language model.
- Since regular languages are closed under intersection, we can intersect the devices and run shortest path graph algorithms.
- Taking their intersection is equivalent to computing the probability under Bayes' rule.

 $O(5n^22^n)$  is still far too much work.

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Barry Haddow will tell you how we can still make it work on Thursday.





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- *All* of these models are weighted regular languages.
- Need dynamic programming with approximations.
- Is this the best we can do? Stay tuned on Thursday.

#### ارتفاع عجز الميزان التجاري الأردني



قيمة الواردات الأردنية بلغت 7.39 مليارات دولار في النصف الأول من العام (الجزيرة نت)

أفادت بيانات رسمية بأن العجز في الميزان التجاري الأردني ارتفع في النصف الأول من العام الحالي بنسبة 18.1% نتيجة زيادة حجم الواردات مقابل الصادرات، كما تراجع حجم القروض الائتمانية التي قدمتها البنوك الأردنية بنسبة 11% خلال الفترة نفسها.

وقالت بيانات لدائرة الإحصاءات العامة نشرت اليوم الاثنين إن قيمة العجز في الميزان التجاري بلغت 2.79 مليار دينار أردني (3.94 مليار دولار أميركي).

وأشارت البيانات إلى ارتفاع حجم الصادرات خلال النصف الأول من العام بنسبة 16.6%, حيث بلغت 2.05 مليار دينار (2.91 مليار دولار)، كما ارتفعت قيمة الواردات بنسبة 11.7% لتبلغ 5.23 مليارات دينار (7.39 مليارات دولار).



Value of Jordanian imports amounted to 7.39 billion dollars in the first half of the year (island Net)

According to official statements that the trade balance deficit rose Jordan in the first half of this year by 18.1% due to increased volume of imports versus exports, as the decline in the volume of credit provided by banks of Jordan by 11% during the same period.

The data for the Department of Statistics published on Monday that the value of the trade balance deficit amounted to 2.79 billion Jordanian dinars (3.94 billion U.S. dollars). The data indicated the high volume of exports during the first half of the year by 16.6%, reaching 2.05 billion dinars (2.91 billion dollars), as imports rose by 11.7% to 5.23 billion dinars (7.39 billion dollars).

The export value increased significantly for the countries of the Greater Arab Free Trade, including Saudi Arabia, and the countries of the free trade agreement for North America, including the United States, as well as non-Arab Asian

#### Overview



## Overview



However , the sky remained clear under the strong north wind .

## Overview



However , the sky remained clear under the strong north wind .



More has been written about machine translation evaluation than about machine translation itself.

#### Yorick Wilks

# Why evaluate? Rank systems Evaluate incremental changes Assess new ideas objectively

#### 美国愿和北韩谈判但拒绝再付出报酬

US willing to negotiate with North Korea but not to pay more compensation.

#### 美国愿和北韩谈判但拒绝再付出报酬

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The United States is willing to hold talks with North Korea but refused to pay remuneration. "奋进"号因机械手故障推迟到升空 Launch of "Endeavour" delayed by robotic arm problems. "奋进"号因机械手故障推迟到升空 Launch of "Endeavour" delayed by robotic arm problems.

"Progress" postponed because of mechanical hand into the sky.

Chinese people in the traditional Spring Festival is approaching, the CPC Central Committee this afternoon in Zhongnanhai on the 22nd non-Party personages to convene a forum in Spring Festival, invited the central committees of democratic parties, the leadership of the National Federation of Industry and Commerce and personages without party affiliation on behalf of comrades gathered together State yes, talked in length about the friendship, to greet the Chinese New Year. CPC Central Committee General Secretary and State President and Central Military Commission Chairman Hu Jintao on behalf of the CPC Central Committee, the State Council, to the central committees of democratic parties, leaders of the National Federation of Industry and Commerce and personages without party affiliation, to

members of the united front, to extend my New Year's blessing.

Although the northern wind shrieked across the sky, it was still very clear.

Although the northern wind shrieked across the sky, it was still very clear.

However , the sky remained clear under the strong north wind .

## Although the northern wind shrieked across the sky, it was still very clear.


Although the northern wind shrieked across the sky, it was still very clear.

However , the sky remained clear under the strong north wind .

Although a north wind was howling , the sky remained clear and blue .

The sky was still crystal clear , though the north wind was howling .

Despite the strong northerly winds , the sky remains very clear .

Although the northern wind shrieked across the sky, it was still very clear.





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unigram: 3/3 bigram: 2/2 trigram: 1/1

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BLEU-1 BLEU-4 BLEU-v11b BLEU-v12 METEOR-v0.6 NIST-v11b TER-v0.7.254-GRR ATEC1 ATEC2 ATEC3 ATEC4 Meteor-v0.7

**BEwT-E** Badger BadgerLite Bleu-sbp BleuSP CDer DP-Or DP-Orp DR-Or EDPM LET **METEOR-ranking** MaxSim

RTE RTE-MT SEPIA1 SEPIA2 SNR SR-Or SVM-Rank TERp ULCh ULCopt invWer mBLEU mTER

# A bit more on learning...

#### p(Chinese|English)

# 

### English





 $p(Chinese|English)^1$  $\times p(English)^1$  $\sim p(English|Chinese)$ English



## $p(Chinese|English)^{1/2}$

 $\times p(English)^{1}$ 

 $\sim p(English|Chinese)$ 

## $p(Chinese|English)^0$



#### $0 \cdot \log p(Chinese|English)$



#### p(English|Chinese) =

# $\frac{1}{Z} \left[ \lambda_1 \log p(Chinese|English) + \lambda_2 \log p(English) \right]$

# $p(English|Chinese) = \frac{1}{Z} \left[ \sum_{i} \lambda_{i} h_{i}(Chinese, English) \right]$

$$\underset{English}{\operatorname{argmax}} p(English|Chinese) = \frac{1}{Z} \left[ \sum_{i} \lambda_{i} h_{i}(Chinese, English) \right]$$

log-linear model conditional random field case-factor diagram undirected model

 $\begin{array}{l} \operatorname{argmax} p(English|Chinese) = \\ \frac{1}{Z} \left[ \sum_{i} \lambda_{i} h_{i}(Chinese, English) \right] \end{array}$ 

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#### linear model

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- We could optimize lambdas for likelihood (this would be a log-linear model).
- Good news: optimization is convex.
- Bad news: computing Z is intractable.
- Question: why should we bother with likelihood?

## $BLEU(MT \ output)$

#### $BLEU(\operatorname{argmax} score(English|Chinese))$ English

# $\sum_{Chinese \in Test} BLEU(\operatorname{argmax} score(English|Chinese))$



# Ôptimization



#### Max Blues
# Things to Remember

- Probability gives us a well-founded framework in which to explore a wide variety of models.
- We get many tools for learning and prediction.
- We can express many models in terms of weighted languages.



# Optimization

## More Information

#### Related ESSLLI course web page: http://homepages.inf.ed.ac.uk/alopez/esslli2010.html

#### Statistical Machine Translation

ADAM LOPEZ University of Edinburgh

Statistical machine translation (SMT) treats the translation of natural language as a machine learning problem. By examining many samples of human-produced translation, SMT algorithms automatically learn how to translate. SMT has made tremendous strides in less than two decades, and new ideas are constantly introduced. This survey presents a tutorial overview of the state of the art. We describe the context of the current research and then move to a formal problem description and an overview of the main subproblems: translation modeling, parameter estimation, and decoding. Along the way, we present a taxonomy of some different approaches within these areas. We conclude with an overview of evaluation and a discussion of future directions.

Categories and Subject Descriptors: G.3 [Probability and Statistics]: Statistical computing; 1.2.6 [Artificial Intelligence]: Learning-Parameter learning; 1.2.7 [Artificial Intelligence]: Natural Language Processing-Machine translation; 1.5.1 [Pattern Recognition]: Models-Statistical General Terms: Algorithms

General Terms: Algorithms Additional Key Words and Phrases: Natural language processing, machine translatio

ACM Reference Format: Lopez, A. 2008. Statistical machine translation. ACM Comput. Surv., 40, 3, Article 8 (August 2008), 49 pages DOI = 10.1145/1380584.1380586 http://doi.acm.org/10.1145/1380584.1380586

#### 1. INTRODUCTION

Machine translation (MT) is the automatic translation from one natural language into another using computers. Interest in MT is nearly as old as the electronic computer popular accounts trace its modern origins to a letter written by Warren Weaver in 1949, only a few years after ENIAC came online.<sup>1</sup> It has since remained a key application in the field of natural language processing (NLP). A god historical overview is given by Hutchins [2007], and a comprehensive general survey is given by Dorr, Jordan, and Benoit (1999).

<sup>1</sup>This letter is reproduced as Weaver [1955].

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## More Information

#### Related ESSLLI course web page: http://homepages.inf.ed.ac.uk/alopez/esslli2010.html



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