

# Models of Synchronous Grammar Induction for SMT

Workshop 2010

The Center for Speech and Language Processing  
Johns Hopkins University

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# Team members

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Desai Chen (CMU)

# Statistical machine translation

Urdu → English

اس حملہ کے بعد بڑی تعداد میں مقامی باشندوں نے علاقوں کو خالی کر دیا ہے .



- Statistical machine translation: Learn how to translate from parallel corpora.

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After this incident, a large number of local residents fled from these areas.

- Statistical machine translation: Learn how to translate from parallel corpora

# Statistical machine translation: Before

Urdu → English

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In this attack a large number of local residents has should vacate areas.

- Current state-of-the-art translation models struggle with language pairs which exhibit large differences in structure.

# Statistical machine translation: After

Urdu → English

اس حملہ کے بعد بڑی تعداد میں مقامی باشندوں نے علاقوں کو خالی کر دیا ہے .



After this attack, a large number of local residents have to vacate the areas.

- In this workshop we've made some small steps towards better translations for difficult language pairs.

# Statistical machine translation: limitations

## Structural divergence between languages:

English	Who wrote this letter?
Arabic	من الذي كتب هذه الرسالة؟ (function-word) (who) (wrote) (this) (the-letter)
Chinese	这封信是谁写的？ (this) (letter) (be) (who) (write) (come-from) (function-word)

# Statistical machine translation: limitations

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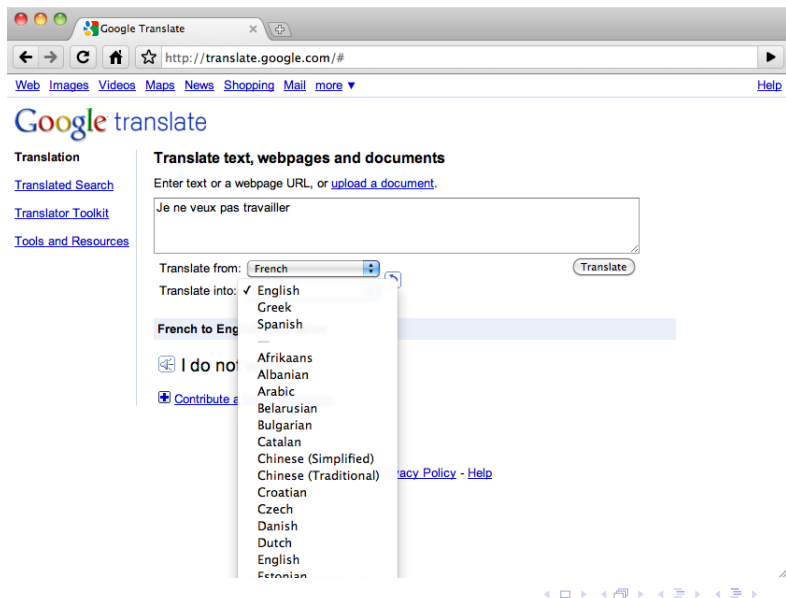
# Statistical machine translation: limitations

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- Phrasal translation equivalences (existing models)
- **Constituent reordering (this workshop!)**
- Morphology (Next year?)

# Statistical machine translation: successes



# Workshop overview

## Input:

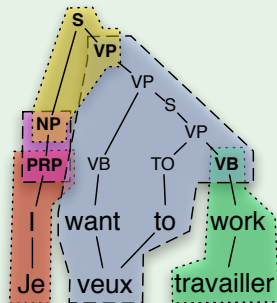
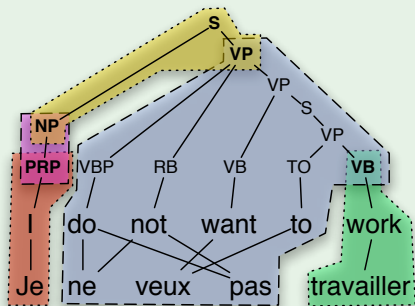
- Existing procedures for synchronous grammar extraction

## Output:

- New unsupervised models for large scale synchronous grammar extraction,
- A comparison and analysis of the existing and proposed models,
- Extended decoders (cdec/Joshua) capable of working efficiently with these models.

# Models of translation

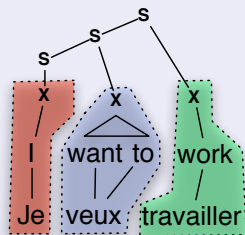
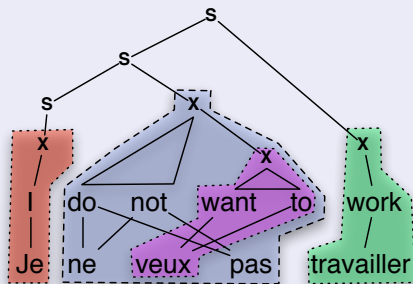
## Supervised SCFG: Syntactic Tree-to-String



- Strong model of sentence structure.
- Reliant on a treebank to train the parser.

# Models of translation

## Unlabelled SCFG: Hiero



- Only requires the parallel corpus.
- But weak model of sentence structure.

# Using syntax in Machine Translation:

## Synchronous Context Free Grammar (SCFG)

$$S \rightarrow \langle X_{[1]}, X_{[1]} \rangle$$
$$X \rightarrow \langle X_{[1]} X_{[2]}, X_{[2]} X_{[1]} \rangle$$
$$X \rightarrow \langle \textit{Sie}, \textit{She} \rangle$$
$$X \rightarrow \langle \textit{eine Tasse Kaffee}, \textit{a cup of coffee} \rangle$$
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## Example Derivation

$$\begin{aligned} S &\Rightarrow \langle X_{[1]}, X_{[1]} \rangle \Rightarrow \langle X_{[2]} X_{[3]}, X_{[2]} X_{[3]} \rangle \\ &\Rightarrow \langle \textit{Sie} X_{[3]}, \textit{She} X_{[3]} \rangle \Rightarrow \langle \textit{Sie} X_{[4]} X_{[5]}, \textit{She} X_{[4]} X_{[5]} \rangle \\ &\Rightarrow \langle \textit{Sie will} X_{[5]}, \textit{She wants to} X_{[5]} \rangle \Rightarrow \langle \textit{Sie will} X_{[6]} X_{[7]}, \textit{She wants to} X_{[7]} X_{[6]} \rangle \\ &\Rightarrow \langle \textit{Sie will eine Tasse Kaffee} X_{[7]}, \textit{She wants to} X_{[7]} \textit{ a cup of coffee} \rangle \\ &\Rightarrow \langle \textit{Sie will eine Tasse Kaffee trinken}, \textit{She wants to drink a cup of coffee} \rangle \end{aligned}$$

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# Models of translation

## Phrase extraction:

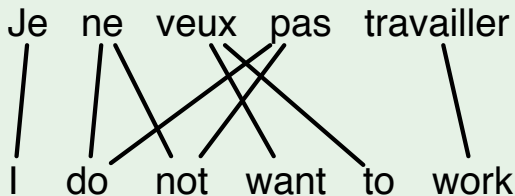
Je ne veux pas travailler

I do not want to work



# Models of translation

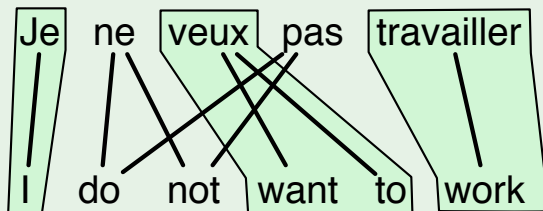
## Phrase extraction:



- Use a word-based translation model to annotate the parallel corpus with word-alignments

# Models of translation

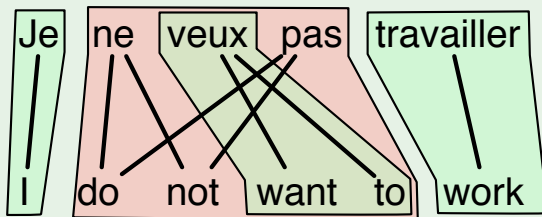
## Phrase extraction:



- $\langle \text{Je, I} \rangle$ ,  $\langle \text{veux, want to} \rangle$ ,  $\langle \text{travailler, work} \rangle$

# Models of translation

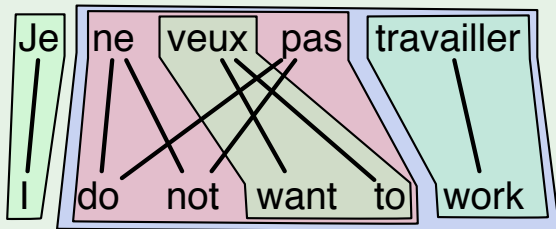
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# Models of translation

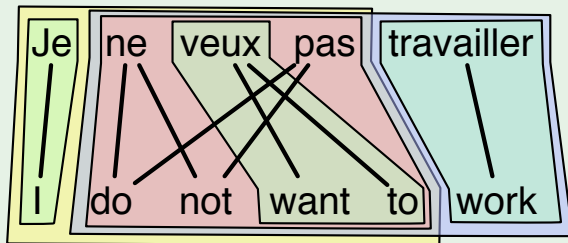
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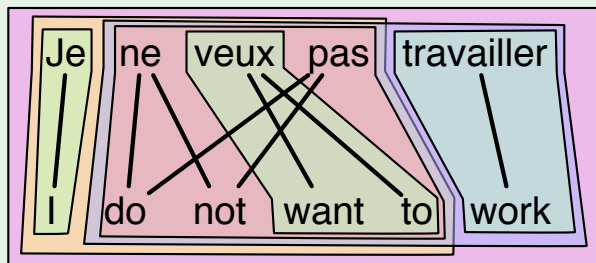
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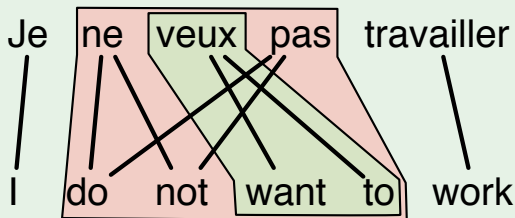
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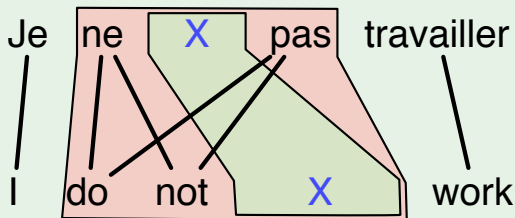
## SCFG Rule extraction:



- $X \rightarrow \langle \text{ne veux pas, do not want to} \rangle$

# Models of translation

## SCFG Rule extraction:

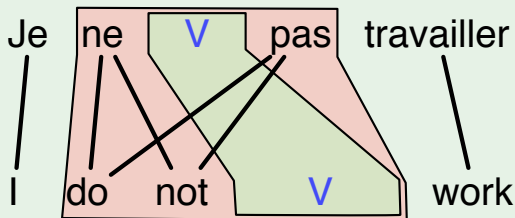


- $X \rightarrow \langle \text{ne veux pas, do not want to} \rangle$ ,
- $X \rightarrow \langle \text{ne } X_{\boxed{1}} \text{ pas, do not } X_{\boxed{1}} \rangle$



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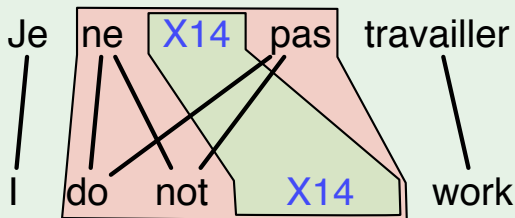
## SCFG Rule extraction:



- $VP/NN \rightarrow \langle \text{ne veux pas, do not want to} \rangle$ ,
- $VP/NN \rightarrow \langle \text{ne } V_{\boxed{1}} \text{ pas, do not } V_{\boxed{1}} \rangle$

# Models of translation

## SCFG Rule extraction:



- $X_{10} \rightarrow \langle \text{ne veux pas, do not want to} \rangle$ ,
- $X_{10} \rightarrow \langle \text{ne } X_{14}_{[1]} \text{ pas, do not } X_{14}_{[1]} \rangle$


















# Impact

Language	Words	Domain
English	4.5M	Financial news
Chinese	0.5M	Broadcasting news
Arabic	300K (1M planned)	News
Korean		Military

**Table:** Major treebanks: data size and domain

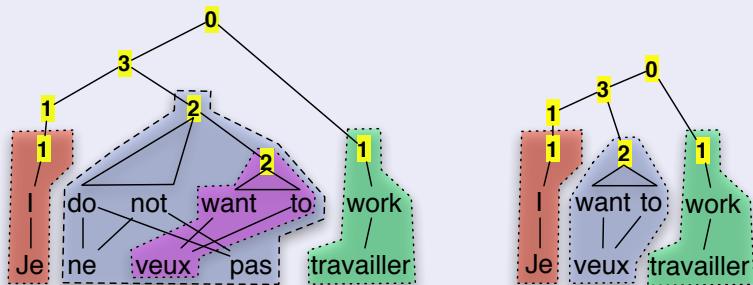
# Impact

Parallel corpora far exceed treebanks (millions of words):

																			
	7	90	83	55	40	50	55	28	29	12	12	8	10	8	7	21	6	6	9
	90	7	34	24	29	12	10	11	11	9	11	7	6	6	7	4	5	5	6
	83	34	7	17	16	12	10	12	11	9	10	8	6	6	7	6	6	5	6
	52	24	17	6	14	12	9	9	10	9	10	7	5	5	6	3	5	5	4
	39	29	16	14	6	9	10	7	8	8	10	8	6	6	6	3	5	5	4
	48	12	12	12	9	3	25	5	5	22	6	2	3	2	3	3	3	3	2
	55	10	10	9	10	26	2	2	2	8	5	2	2	2	2	2	2	2	1
	26	11	12	9	7	5	2	7	12	3	4	6	5	4	7	3	5	5	4
	29	11	11	10	8	5	2	12	6	3	4	6	6	5	6	3	5	5	4
	12	9	9	9	8	23	8	3	3	2	6	1	2	2	2	2	2	2	2
	11	11	10	10	10	6	5	4	4	6	4	5	3	3	4	1	3	3	3
	8	7	8	7	8	2	2	6	6	1	5	5	4	4	5	2	4	4	3

# Models of translation

## Hierarchical



- AIM: Implement a large scale open-source synchronous constituent learning system.
- AIM: Investigate and understand the relationship between the choice of synchronous grammar and SMT performance,
- AIM: and fix our decoders accordingly.

# Evaluation goals

We will predominately evaluate using BLEU, but also use automatic structured metrics and perform small scale human evaluation:

- Evaluate phrasal, syntactic, unsupervised syntactic,
- Aim 1: Do no harm (not true of existing syntactic approach)
- Aim 2: Exceed the performance of current non-syntactic systems.
- Aim 3: Meet or exceed performance of existing syntactic systems.

# Workshop Streams

- Implement scalable SCFG grammar extraction algorithms.
- Improve SCFG decoders to efficiently handle the grammars produced.
- Investigate discriminative training regimes that leverage features extracted from these grammars.

# Language pairs (small)

- BTEC Chinese-English:
  - ▶ 44k sentence pairs, short sentences
  - ▶ Widely reported 'prototyping' corpus
  - ▶ Hiero baseline score: 52.4 (16 references)
  - ▶ Prospects: BTEC always gives you good results
- NIST Urdu-English:
  - ▶ 50k sentence pairs
  - ▶ Hiero baseline score: MT05 - 23.7 (4 references)
  - ▶ Major challenges: major long-range reordering, SOV word order
  - ▶ Prospects: small data, previous gains with supervised syntax



# Language pairs (large)

- NIST Chinese-English:

- ▶ 1.7M sentence pairs, Standard NIST test sets
- ▶ Hiero baseline score: MT05 - 33.9 (4 references)
- ▶ Major challenges: large data, mid-range reordering, lexical ambiguity
- ▶ Prospects: supervised syntax gains reported

- NIST Arabic-English:

- ▶ 900k sentence pairs
- ▶ Hiero baseline score: MT05 - 48.9 (4 references)
- ▶ Major challenges: strong baseline, local reordering, VSO word order
- ▶ Prospects: difficult

- Europarl Dutch-French:

- ▶ 1.5M sentence pairs, standard Europarl test sets
- ▶ Hiero baseline score: Europarl 2008 - 26.3 (1 reference)
- ▶ Major challenges: V2 / V-final word order, many non-literal translations
- ▶ Prospects: ???

# Summary

- Scientific Merit:
  - ▶ A systematic comparison of existing syntactic approaches to SMT.
  - ▶ An empirical study of how constituency is useful in SMT.
  - ▶ An evaluation of existing theories of grammar induction in a practical application (end-to-end evaluation).
- Potential Impact:
  - ▶ Better MT systems, for more languages, across a range of domains.
  - ▶ More accessible high performance translation models for researchers.
- Feasibility:
  - ▶ A great team with a wide range of both theoretical and practical experience.
  - ▶ Solid preparation.
- Novelty:
  - ▶ First attempt at large scale unsupervised synchronous grammar induction.