

Machine Translation Summit XIV, Nice, France 2 - 6 September 2013

Big Data Adaptation DatAptor

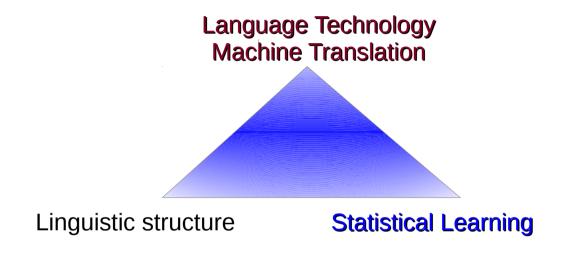
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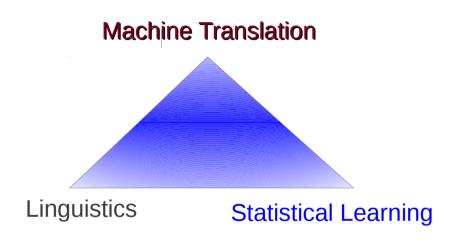
MT at ILLC-UvA

Statistical Language Processing and Learning Lab.



MT at ILLC-UvA

Statistical Language Processing and Learning Lab.



Main topics within SLPL

- Syntax-driven SMT (learning, decoding)
- Learning latent reordering for translation
- Hierarchical models with morph. And syntax
- Data-powered Adaptation
- ..

- **SLPL Lab** (Growing: 8 PhD students; 4 postdocs; Programmer)
- Five projects on Statistical MT (2012—2019)

This talk: Big Data and DatAptor (Feb 2013 - Feb 2017)

BIG DATA What Comes to Mind?

BIG DATA

Data Data Data ... (Repeated many many times)

BIG DATA

Everyone wants big data

(Does anyone know what for?)

BIG DATA What comes to mind?

Efficient computing

Big storage + Fast search

Diversity

Quality differences

Noise; Difficult statistics ...

Saturated statistics

Just count and divide

Simple models are enough

[cf. The Unreasonable Effectiveness of Data. Halevy, Norvig and Pereira 2009.]

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Diversity also offers advantages

- Language use (different domains)
- Translators practice and guidelines

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DatAptor Project

DatAptor Project 2013-2017

Technology Foundation STW (NWO)

- Data-Powered
- Domain-Specific
- Translation Services
- on Demand

Researchers (SLPL, ILLC, UvA)

- Dr Khalil Sima'an (principal investigator)
- Dr Christof Monz (senior researcher)
- Dr Bart Mellebeek (postdoc: Jan 2013)
- Milos Stanojevic (PhD student: March 2013)
- Vacancies: postdoc + programmer

Partners (User Board)

- TAUS
- EC DGT
- Intel Inc.
- Symantec

DatAptor Motivation

Technology Foundation STW (NWO)

Versatile adaptation needed

- Potential demand vs. current demand
 - Continuously increasing text volumes
 - Large variability in kinds of texts (domains)
- Changing translation market
 - Changing domains, e.g. shifting international trade/cultural/... exchange
 - Changing acceptance for automatic translation

Versatile adaptation of MT Engines? How?

So many domains... so little time...

Versatile Build an MT System

For every domain of language use —

sports; news; politics; financial; banking; automotive; drugs; food; appliance manuals; hardware/software manuals; scientific articles;

Automatically and rapidly

→ Minin

Minimal human intervention

On demand: user specs.

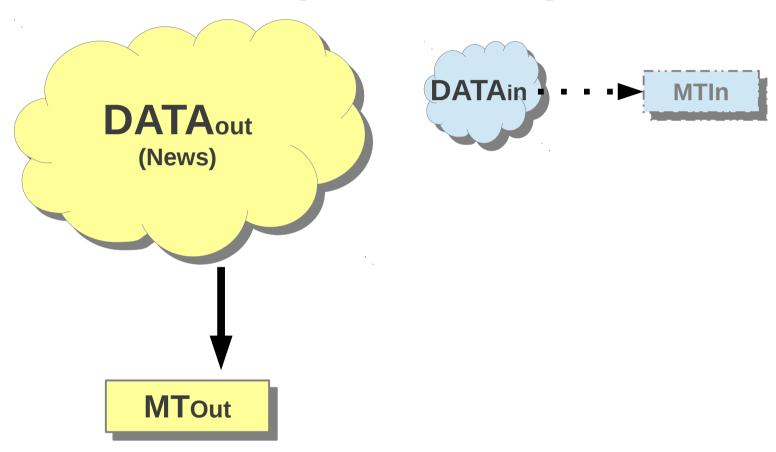
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User supplied example texts to be translated

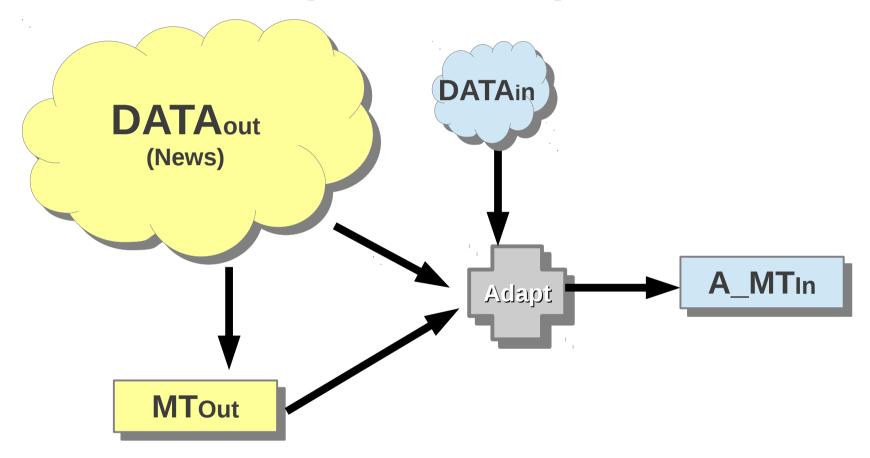
A population of MT Systems!

Current Practice Tiny Data Adaptation

Current Practice Tiny Data Adaptation



Current Practice Tiny Data Adaptation



Tiny Data Adaptation Current Practice

Task

Build MT system from tiny in-domain data using whatever out-of-domain data exists

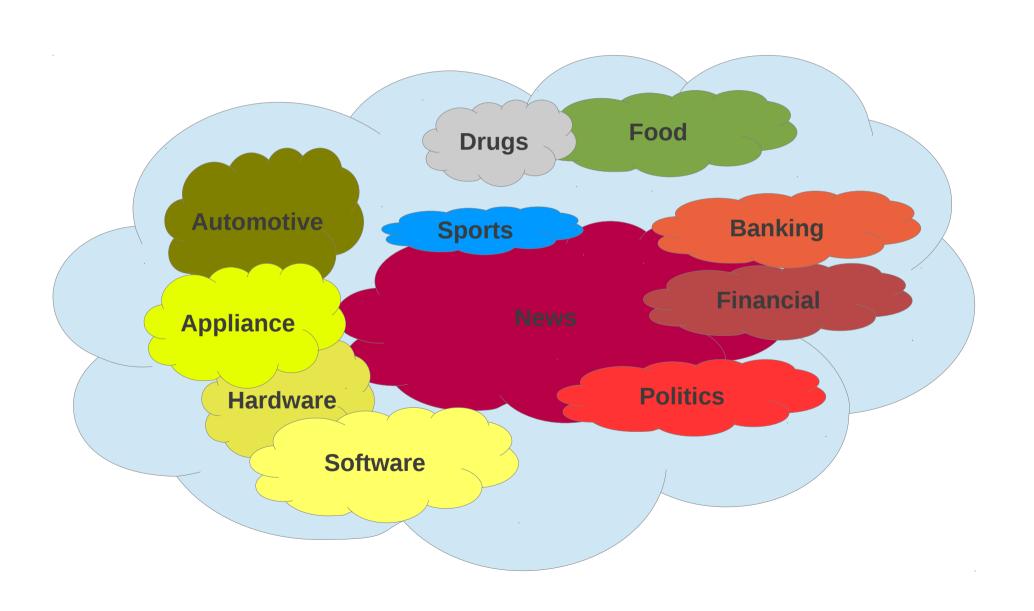
Theoretically Challenging, interesting, very difficult

Practice Assumptions maybe too strong

Alternative Scenario

BIG DATA Adaptation

Big Data == Diversity



DatAptor Hypothesis

Big Data

Metaphore

Imagine a world of translators

- A translator: background + experience
- A translator for every situation

For every new translation order Find the best suitable translator



DatAptor Hypothesis

Big Data == Diversity

Metaphore

Imagine a world of translators

- · A translator for every situation
- A translator with own background and experience

For every new translation order

Find the best suitable translator



If (Big Data == ``A World of Domains")

Diversity enables rapid adaptation to new domain

``Find the most suitable MT system in the Data"

DatAptor Challenges I

INPUT: User documents from some domain + BIG DATA OUTPUT: SMT system adapted for domain

- Distill from BIG DATA a suitable training data Weigh some documents as more relevant than others
- Train SMT system on distilled data.

Map of BIG DATA

- Efficiency for distilling suitable training data
 Map: the more related, the closer to each other on the map
- How to measure domain similarity?
 Statistical (hierarchical-)topic similarity; translation-equivalence and instance weighting ...

BIG DATA vs. BIG Trans. DATA

BIG T-DATA

Translation Data

Translation Equations

$$VIII = 7$$

Meaning Equations

$$\forall \div [] \equiv \forall []$$

Semantic Equation

BIG DATA

Domain Diversity

Topic Similarity(?)

Semantic Similarity(?)



BIG T-DATA

Translation Data

Translation Equations

$$V = 7$$

Meaning Equations

$$V+11 = V11$$
 5+2 = 7

$$5+2=7$$

Semantic Equation

$$VII = (V+II) = (5+2) = 7$$

BIG DATA



Domain Diversity

Topic Similarity(?)

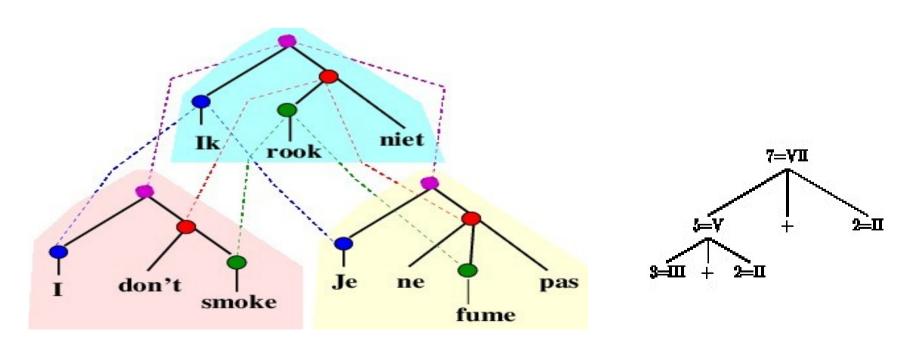
Semantic Similarity(?)

More ... Many More **Accurate Topic/Semantic Equations/Similarity relations**

DatAptor Challenges II

- User-Driven Data-Powered Adaptation
- Recursive, Hierarchical Meaning Equations

Structures translation equations; Better reordering
Better fit with morpho-syntax; ``Deeper" meaning equations



To conclude

Big Trans. Data

- Enables Data-Powered Adaptation (DatAptation)
- Statistics over ``Meaning Equations"
- More than MT? Language understanding!

Thank you! k.simaan@uva.nl



