

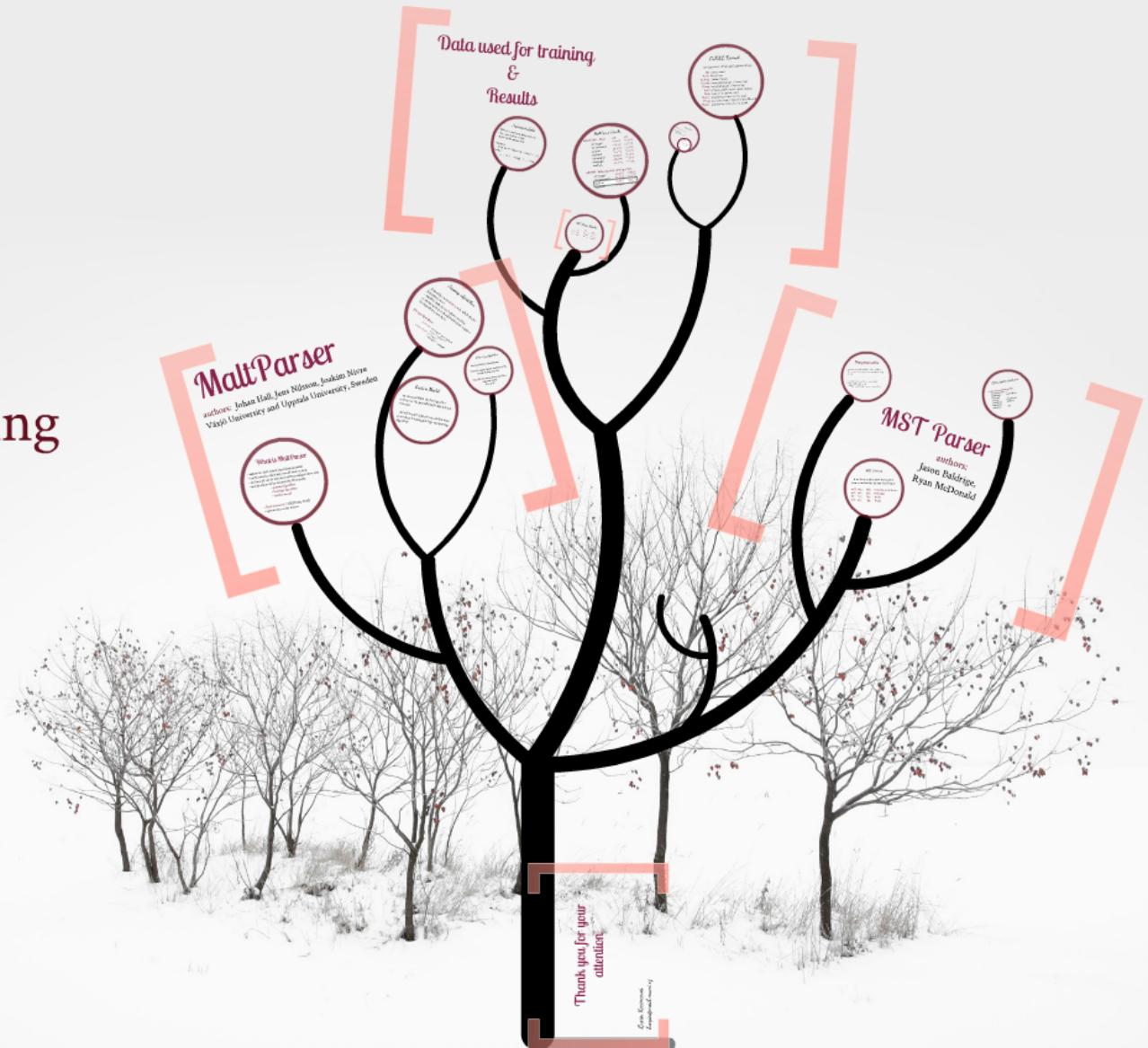
# Trees

or what is left after parsing



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# MaltParser

authors: Johan Hall, Jens Nilsson, Joakim Nivre  
Växjö University and Uppsala University, Sweden

## What is MaltParser

- system for data driven dependency parsing
- can be used for inducing a model from corpus
- such model can be used for further parsing of new data
  - specification can be divided into three parts:
    - parsing algorithm
    - learning algorithm
    - feature model
- latest version 1.7.2 (25th Sep. 2012)
- open source coded in Java

**Parsing algorithm**  
- defined by a transition system which derives dependency trees  
- together with an oracle that is used for reconstruction of each valid transition sequence for dependency structures

**Feature Model**  
- an external XML file that specifies features of the partially built dependency structure  
- default model: depends on combination of machine learning package and parsing algorithm

**Learning algorithm**  
- two machine learning packages:  
**LIBSVM**: support vector machines with kernels (Chang and Lin)  
**LIBLINEAR**: various linear classifiers including SVMs (Fan et al.)

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# *Parsing algorithm*

- defined by a **transition system** which derives dependency trees
- together with an **oracle** that is used for reconstruction of each valid transition sequence for dependency structures

## Parsing algorithms:

**projective:** nivreager\*, nivrestandard\*,  
covproj^, stackproj\*\*

**non-projective:** covnonproj^,  
stackeager\*\*, stacklazy\*\*

## *Learning algorithm*

- two machine learning packages:

**LIBSVM**: support vector machines with kernels (Chang and Lin)

**LIBLINEAR**: various linear classifiers including SVMs  
(Fan et al.)

# *Feature Model*

- an external XML file that specifies features of the partially built dependency structure
- default model: depends on combination of machine learning package and parsing algorithm

# MST Parser

authors:

Jason Baldridge,  
Ryan McDonald

## MST Format

- from latest version supports CoNLL  
format but has also its own data format:

w(1) w(2) .. w(n) - n words of a sentence  
p(1) p(2) .. p(n) - POS tags  
l(1) l(2) .. l(n) - labels  
d(1) d(2) .. d(n) - heads

## How parser works

- non-projective dependency parser based on  
searching minimum spanning trees over directed  
graphs  
- latest version 0.5.0 (23rd January 2012)  
- system developed in Java  
and distributed under Apache Licence V1.0

## Optimization features

- options that need to be specified:  
• training-iteration: 10  
• neuposi/proj  
• decode-type: 1  
• training-k:  
• loss-type:  
• create-dict:  
• order/scope:  
of features:  
1/2

# How parser works

- non-projective dependency parser based on searching maximum spanning trees over directed graphs
- latest version 0.5.0 (23th January 2012)
- system developed in Java  
and distributed under Apache Licence V2.0

## *Optimization features*

- options that need to be specified:
  - training-iterations: 10
  - decode-type: nonproj/proj
  - training-k: 1
  - loss-type: punc/nopunc
  - create-forest: false/true
  - order/scoped  
of features: 1/2

## *MST Format*

- from latest version supports CoNNL format but has also its own data format:

w(1) w(2) .. w(n) - n words of a sentence

p(1) p(2) .. p(n) - POS tags

l(1) l(2) .. l(n) - labels

d(1) d(2) .. d(n) - heads

# Data used for training & Results

## Training data

- PDT 2.0 used for training purposes
- data annotated on a-layer
- manually disambiguated

sentences  
train: 68 495 | dev: 9 270 | test: 10 148  
tokens  
1 171 191 | 158 962 | 173 586

## MaltParser Results

	UA	LA
nivreager	79.99 %	71.89 %
nivrestandard	71.43 %	64.73 %
cproj	80.13 %	71.43 %
stackproj	79.67 %	73.99 %
cvnvnpnproj	80.58 %	74.95 %
stackeager	82.54 %	77.14 %
stacklary	83.17 %	77.74 %

	UA	LA
nivreager	83.21 %	78.42 %
nivrestandard	81.51 %	76.37 %
stacklary	85.02 %	80.05 %
stackproj	85.00 %	77.47 %

## MST Parser Results

	LA	UA
-PDT - 2.0	42.19 %	77.72 %
-PDT - 4.0	72.44 %	83.07 %
-PDT - 4.5	72.00 %	83.06 %

## CoNLL Format

- introduced in CoNLL shared task workshops

- ID token counter
- Form word form
- Lemma lemma of word
- Cpostag coarse-grained part-of-speech tag
- Postag fine-grained part-of-speech tag
- Feats syntactic and/or morphological features
- Head head of the current token
- Deprel dependency relation to the head
- Phead projective head of current token (ID or 0)
- Pdeprel dependency relation to the phead

## Evaluation

CoNLL shared task workshops  
http://conll.csail.mit.edu/2006/index.html

# *Training data*

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sentences

train: 68 495 | dtest: 9 270 | etest: 10 148

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# *CoNNL Format*

- introduced in CoNNL shared task workshops

**ID** token counter

**Form** word form

**Lemma** lemma of word

**Cpostag** coarse-grained part-of-speech tag

**Postag** fine-grained part-of speech tag

**Feats** syntactic and/or morphological features

**Head** head of the current token

**Deprel** dependency relation to the head

**Phead** projective head of current token (ID or 0)

**Pdeprel** dependency relation to the phead

## *MaltParser Results*

**LIBLINEAR ~ 2hod.**

	UA	LA
nivreeager	79.99 %	71.89 %
nivrestandard	71.43 %	64.73 %
covproj	80.13 %	71.43 %
stackproj	79.67 %	73.99 %
covnonproj	80.58 %	74.95 %
stackeager	82.54 %	77.14 %
stacklazy	83.17 %	77.74 %

**LIBSVM ~20hod. (použitý splitting trick)**

nivreeager	83.21 %	78.42 %
nivrestandard	81.51 %	76.37 %
stacklazy	85.02 %	80.05 %
stackproj	83.00 %	77.47 %

# *MST Parser Results*

	LA	UA
opt. 1 ~ 5 h.	69.19 %	77.73 %
opt. 2 ~ 4 h.	75.34 %	83.01 %
opt. 3 ~ 4.5 h.	75.39 %	83.04 %

# *Evaluation*

CoNNL shared Task results:

unlabeled accuracy ~ 87.30 %

labeled accuracy ~ 80.38

## *Evaluation script*

CoNNL Shared tasks used  
various scripts, eg.  
Randomized Parsing  
Evaluation Comparator

own script developed



Thank you for your  
attention

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