

# TectoMT: Machine Translation System

**Martin Popel**

ÚFAL (Institute of Formal and Applied Linguistics)  
Charles University in Prague



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# Outline

- Treex (framework) vs. TectoMT (MT system)
- Demo translation step by step
- Annotation of translation errors
- Details
  - Hidden Markov Tree Models (HMTM)
  - Combining dictionaries
  - Maximum Entropy dictionary
- Results
- Examples of translation

# Treex vs. TectoMT



2005

...

2011

**NLP framework**  
*TectoMT*

**MT system**  
*TectoMT*

lemmatization

tagging

parsing

Main author:  
Zdeněk Žabokrtský

**multi-purpose  
NLP framework**  
*Treex*

**MT system**  
*TectoMT*

lemmatization

tagging

parsing

coreference

PEDT preprocessing

CzEng analysis

treebank conversions

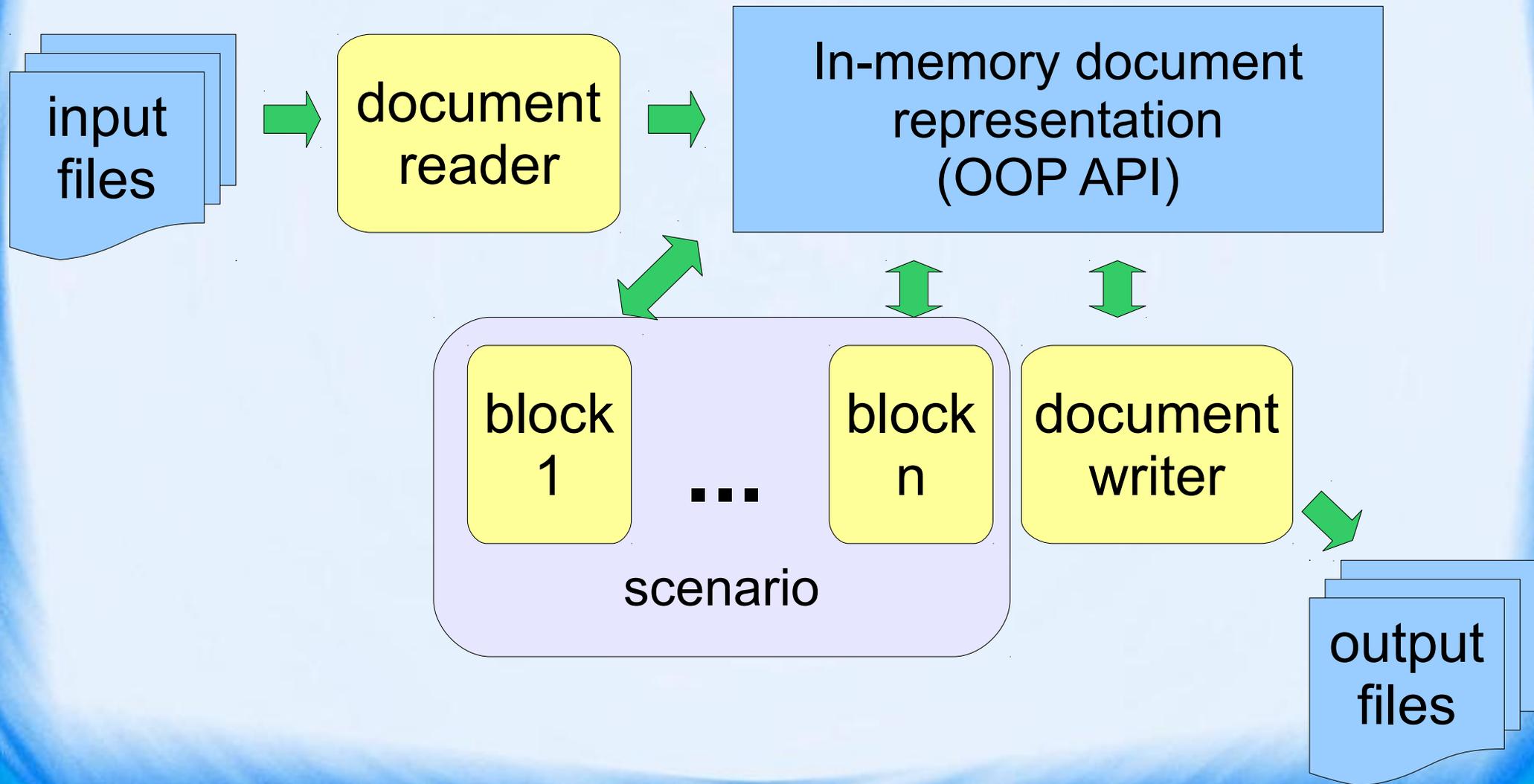
named entity r.

alignment (word,tree)

SMT preproc.

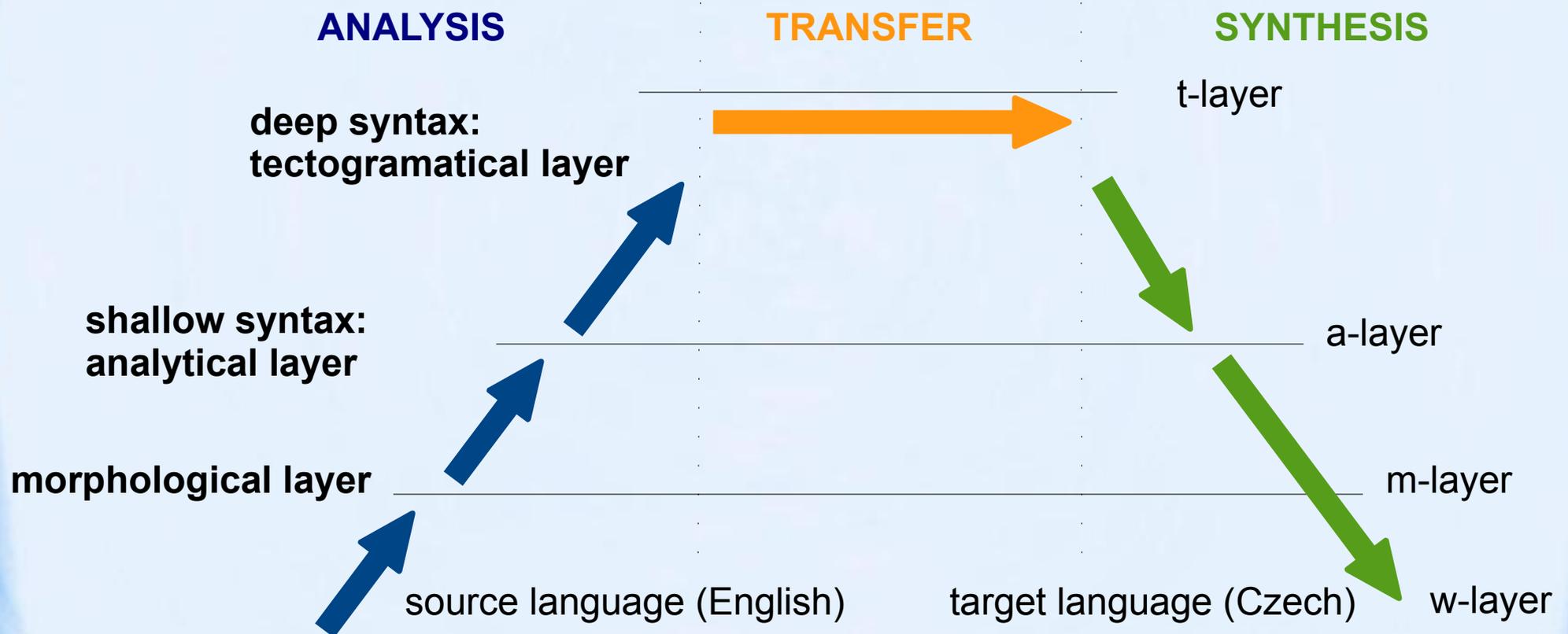
etc.

# Treex architecture



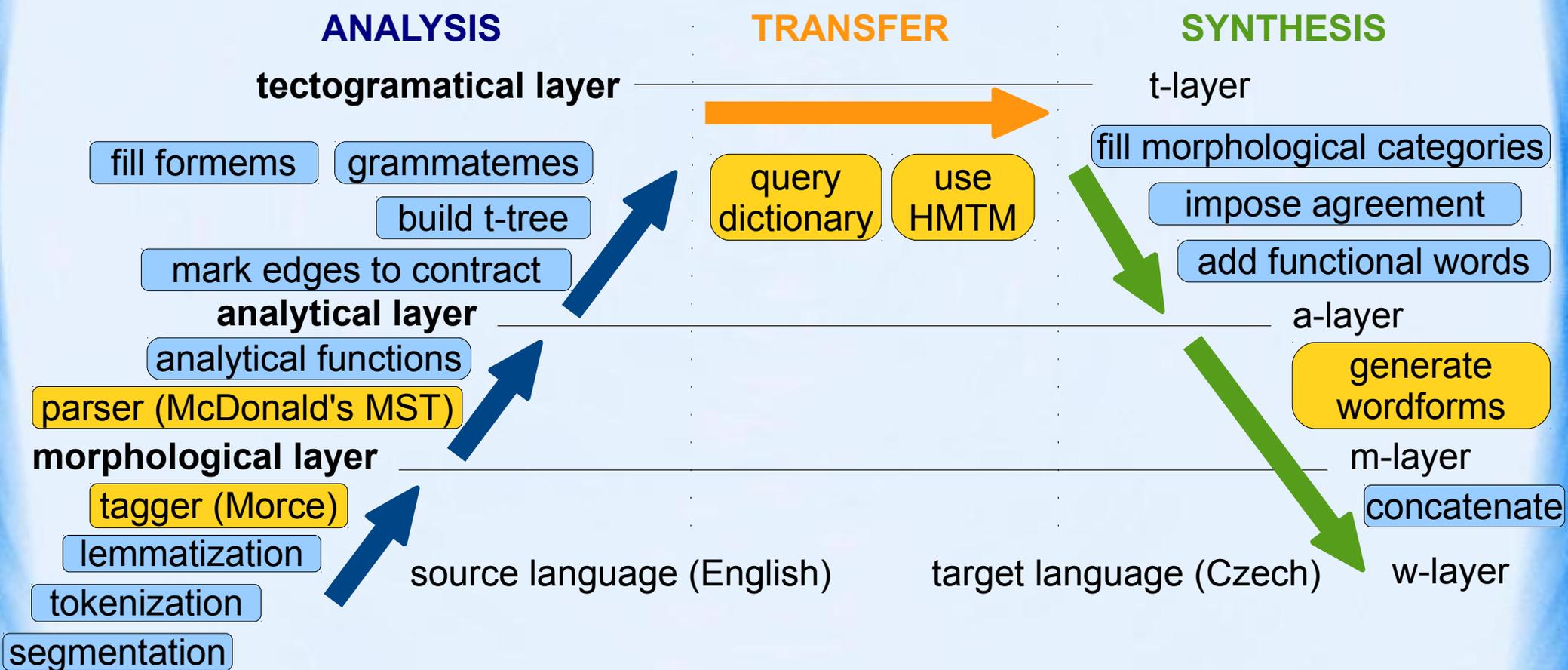
# Translation scheme

transfer over the tectogrammatical layer



# Translation scheme

rule based & statistical blocks



# Demo Translation – Analysis

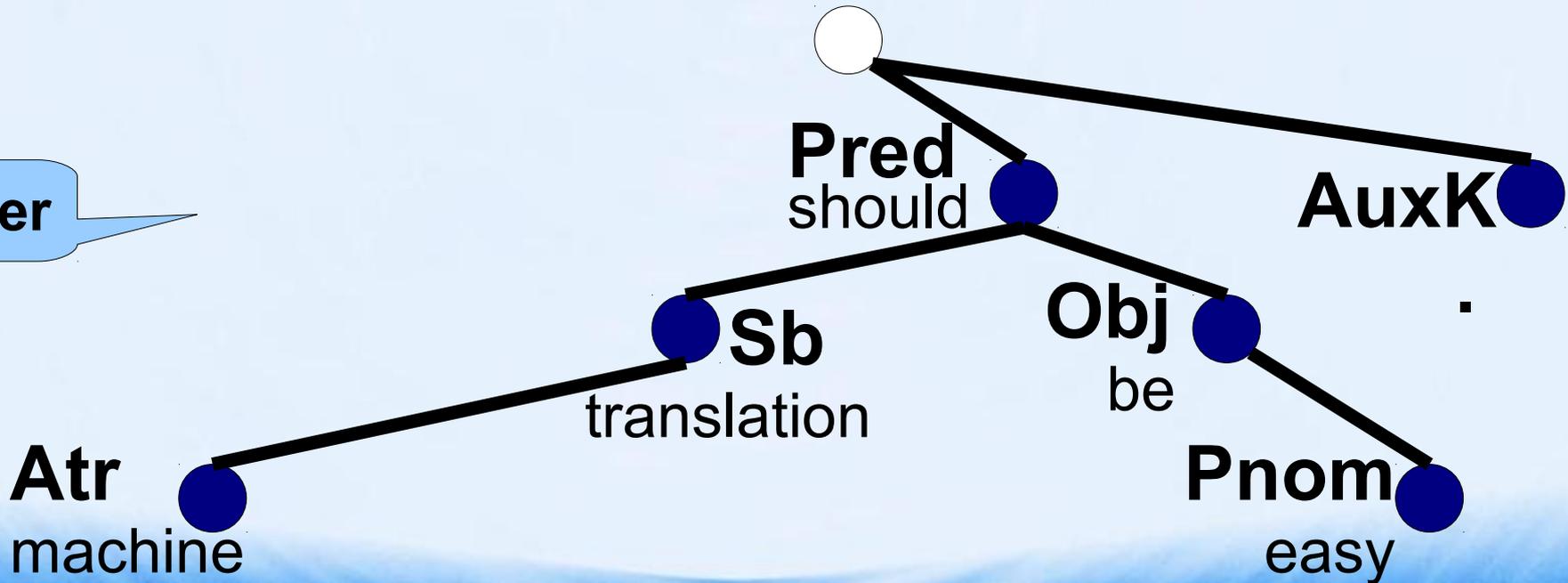
raw text

Machine translation should be easy.

m-layer

●	●	●	●	●	●
machine	translation	should	be	easy	.
NN	NN	MD	VB	JJ	.

a-layer



# Demo Translation – Analysis

raw text

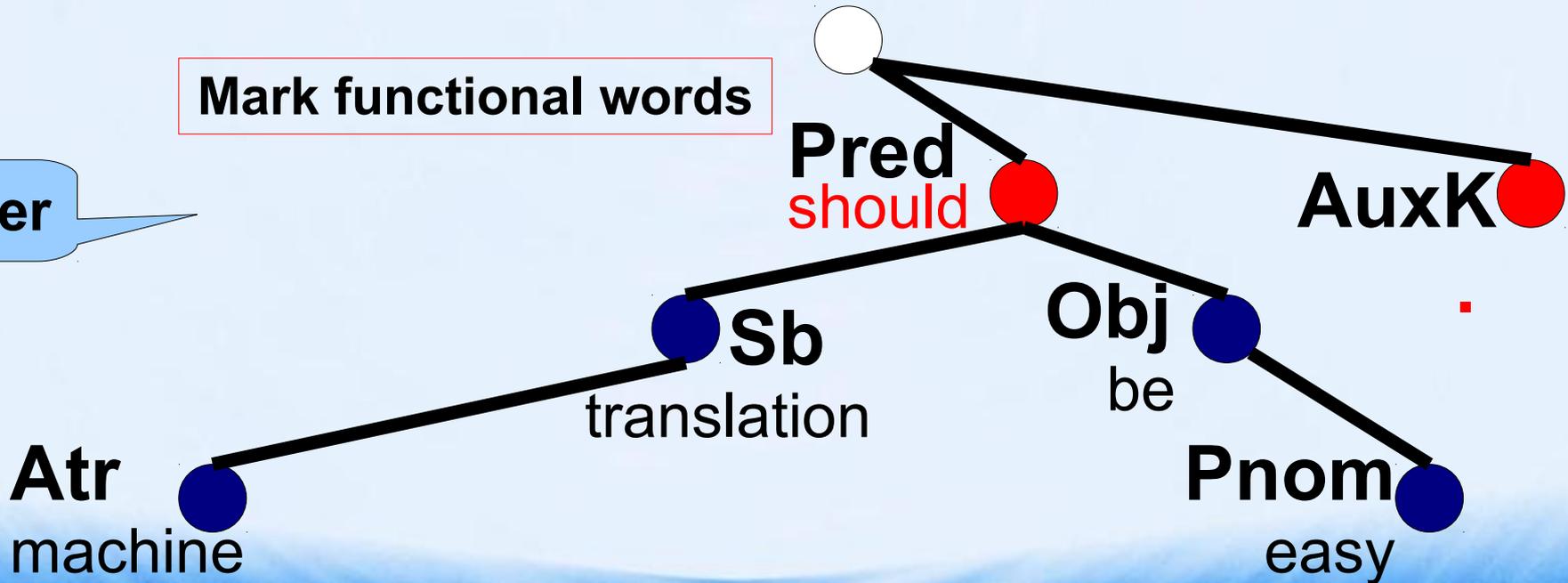
Machine translation should be easy.

m-layer

●	●	●	●	●	●
machine	translation	should	be	easy	.
NN	NN	MD	VB	JJ	.

Mark functional words

a-layer



# Demo Translation – Analysis

raw text

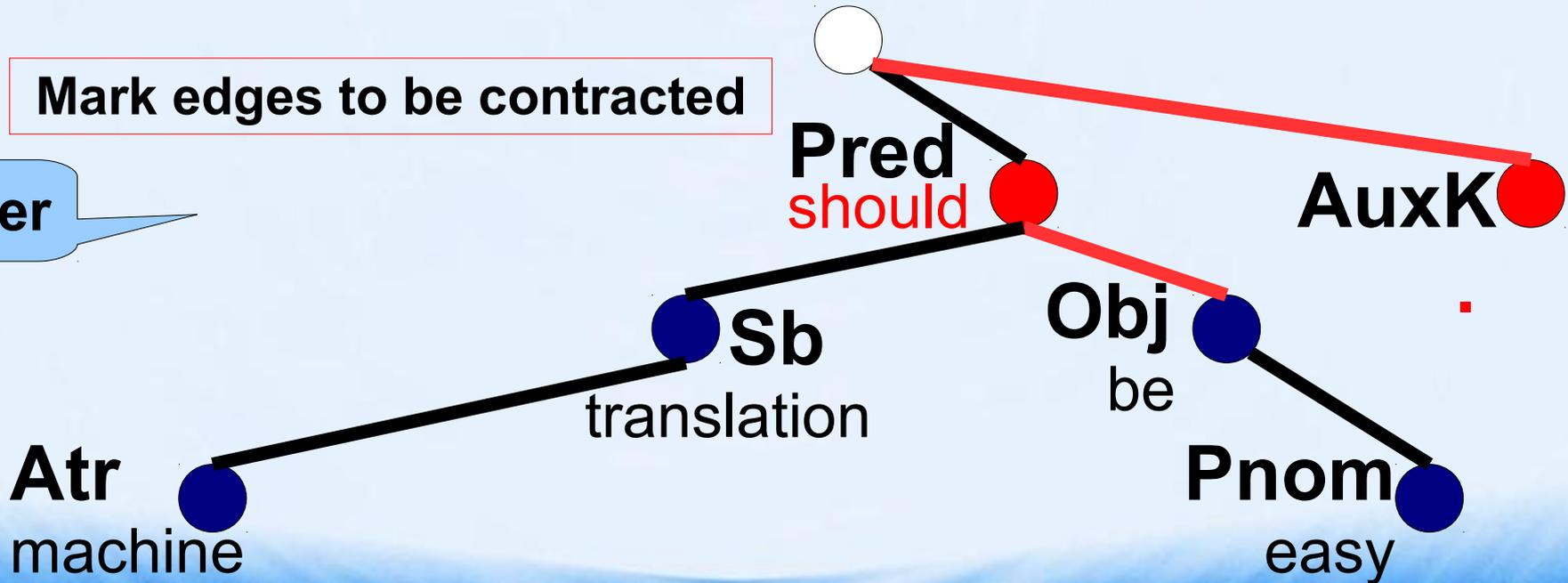
Machine translation should be easy.

m-layer

●	●	●	●	●	●
machine	translation	should	be	easy	.
NN	NN	MD	VB	JJ	.

Mark edges to be contracted

a-layer



# Demo Translation – Analysis

raw text

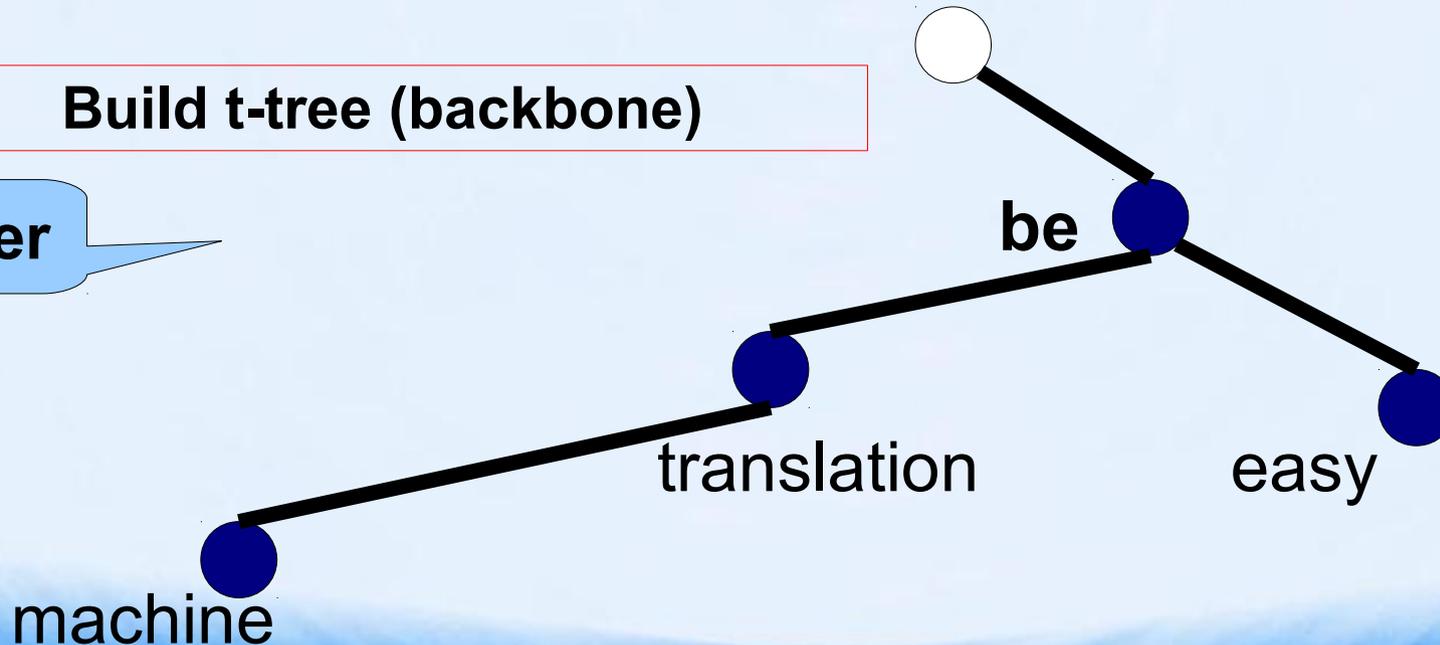
Machine translation should be easy.

m-layer

●	●	●	●	●	●
machine	translation	should	be	easy	.
<b>NN</b>	<b>NN</b>	<b>MD</b>	<b>VB</b>	<b>JJ</b>	.

Build t-tree (backbone)

t-layer



# Demo Translation – Analysis

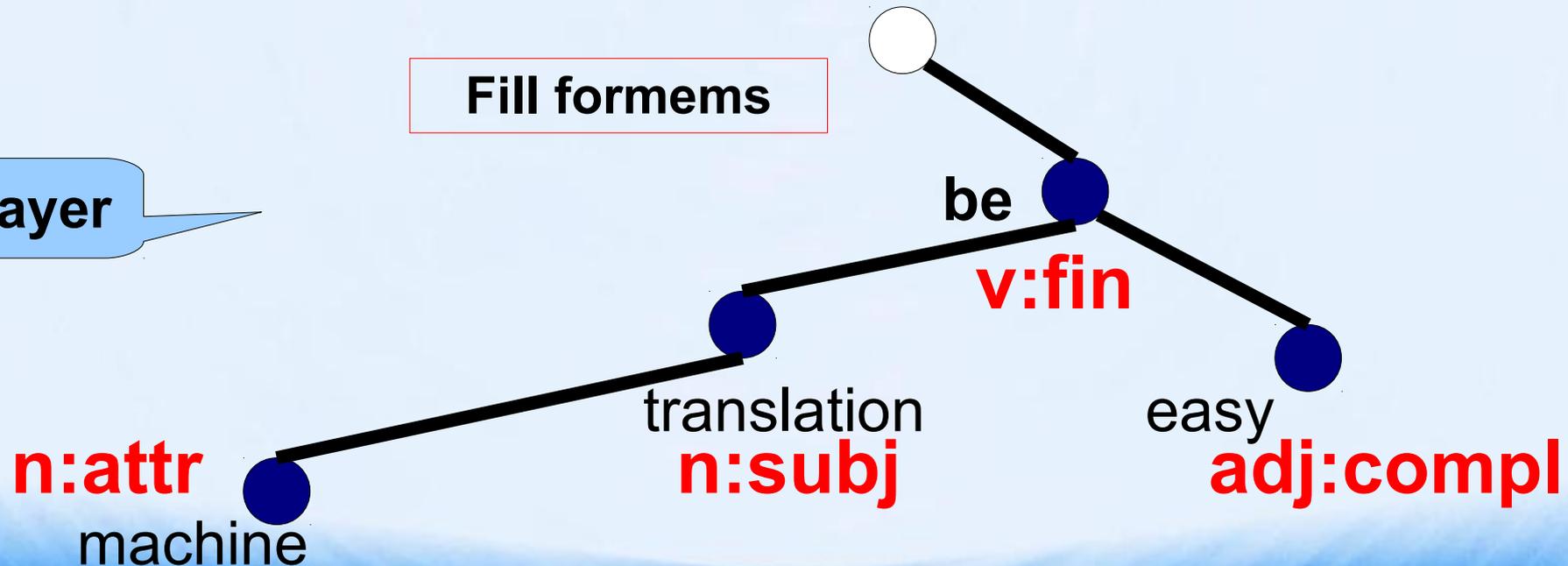
raw text

Machine translation should be easy.

m-layer

●	●	●	●	●	●
machine	translation	should	be	easy	.
<b>NN</b>	<b>NN</b>	<b>MD</b>	<b>VB</b>	<b>JJ</b>	.

t-layer



# Demo Translation – Analysis

raw text

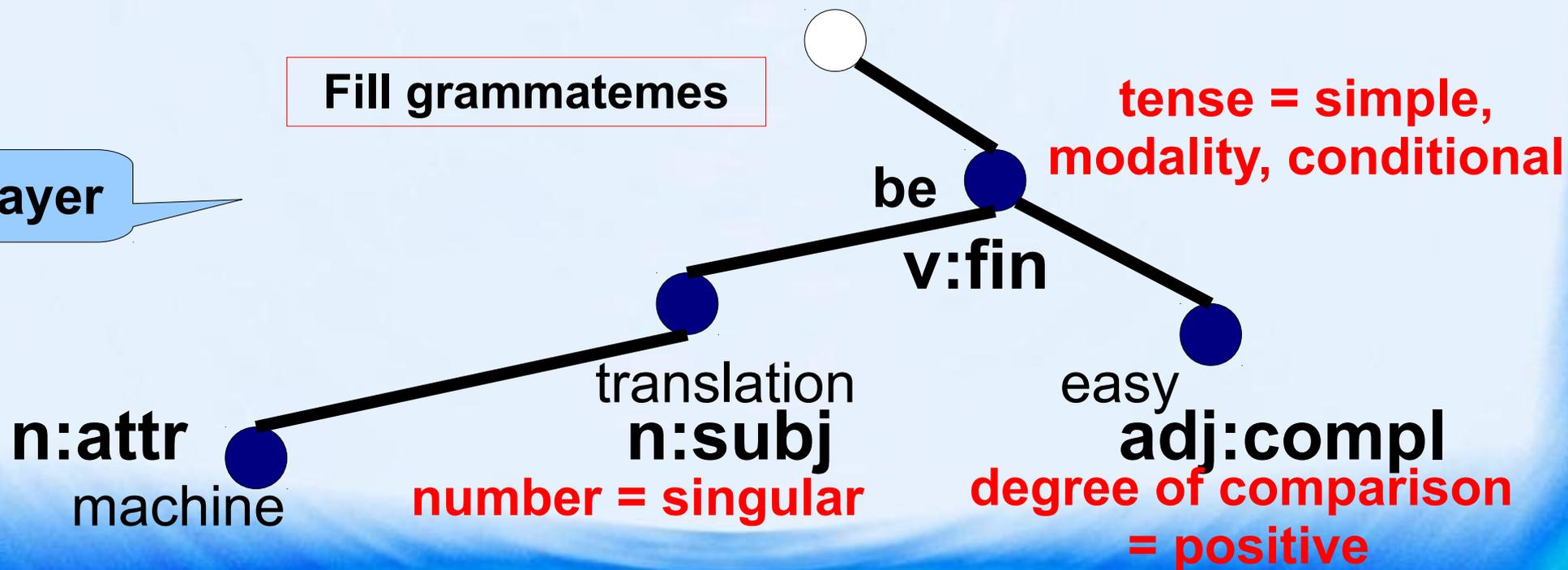
Machine translation should be easy.

m-layer

●	●	●	●	●	●
machine	translation	should	be	easy	.
<b>NN</b>	<b>NN</b>	<b>MD</b>	<b>VB</b>	<b>JJ</b>	.

Fill grammatememes

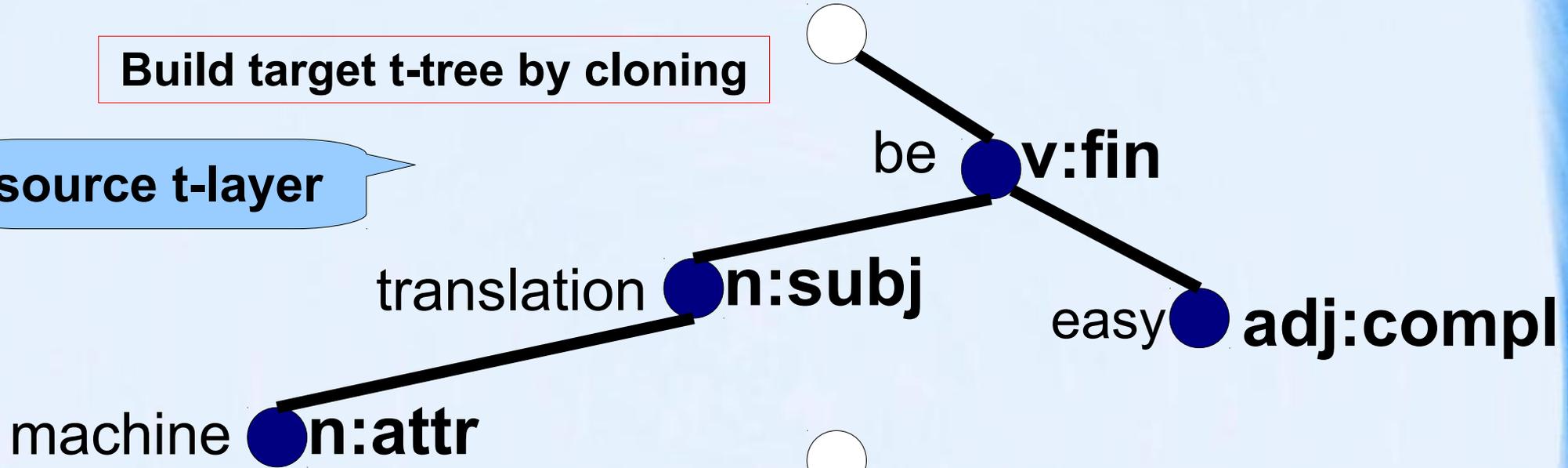
t-layer



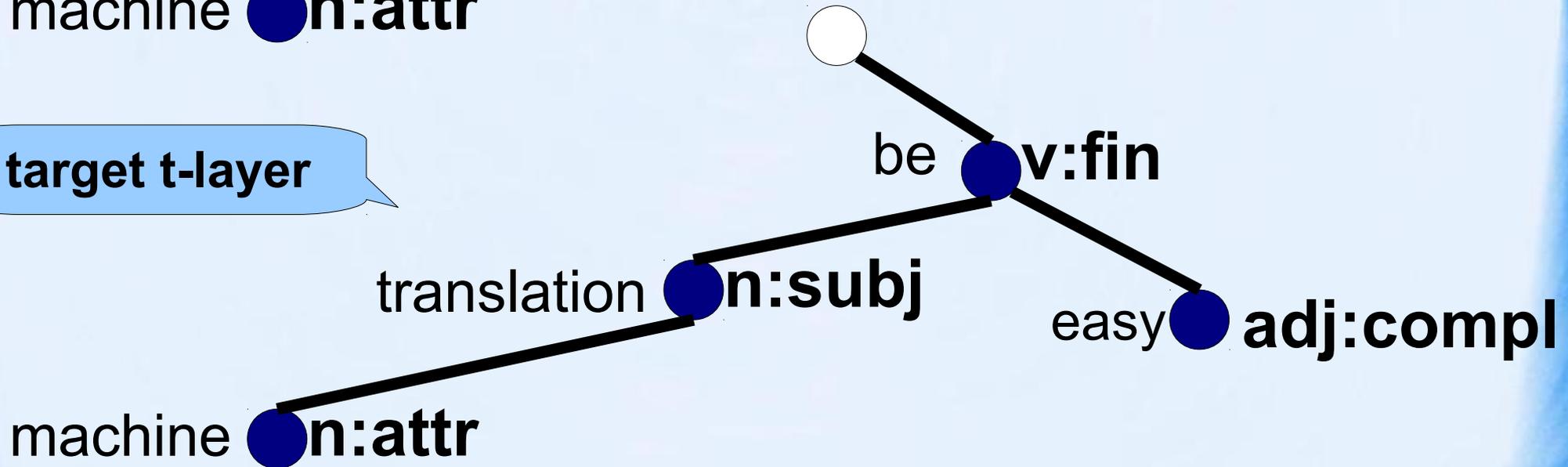
# Demo Translation – Transfer

Build target t-tree by cloning

source t-layer



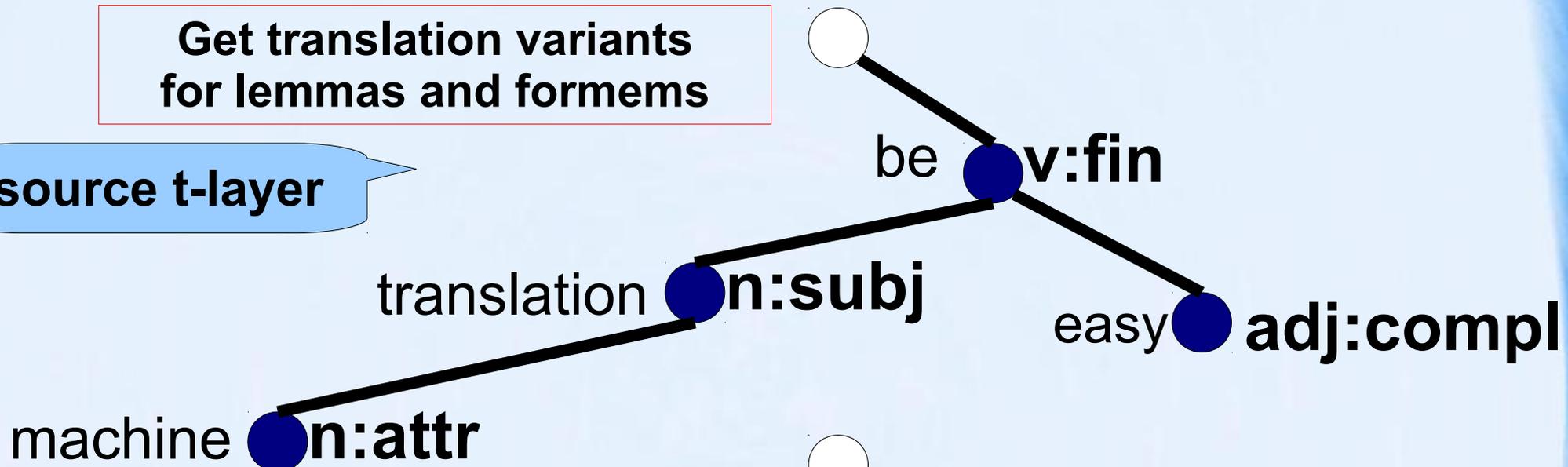
target t-layer



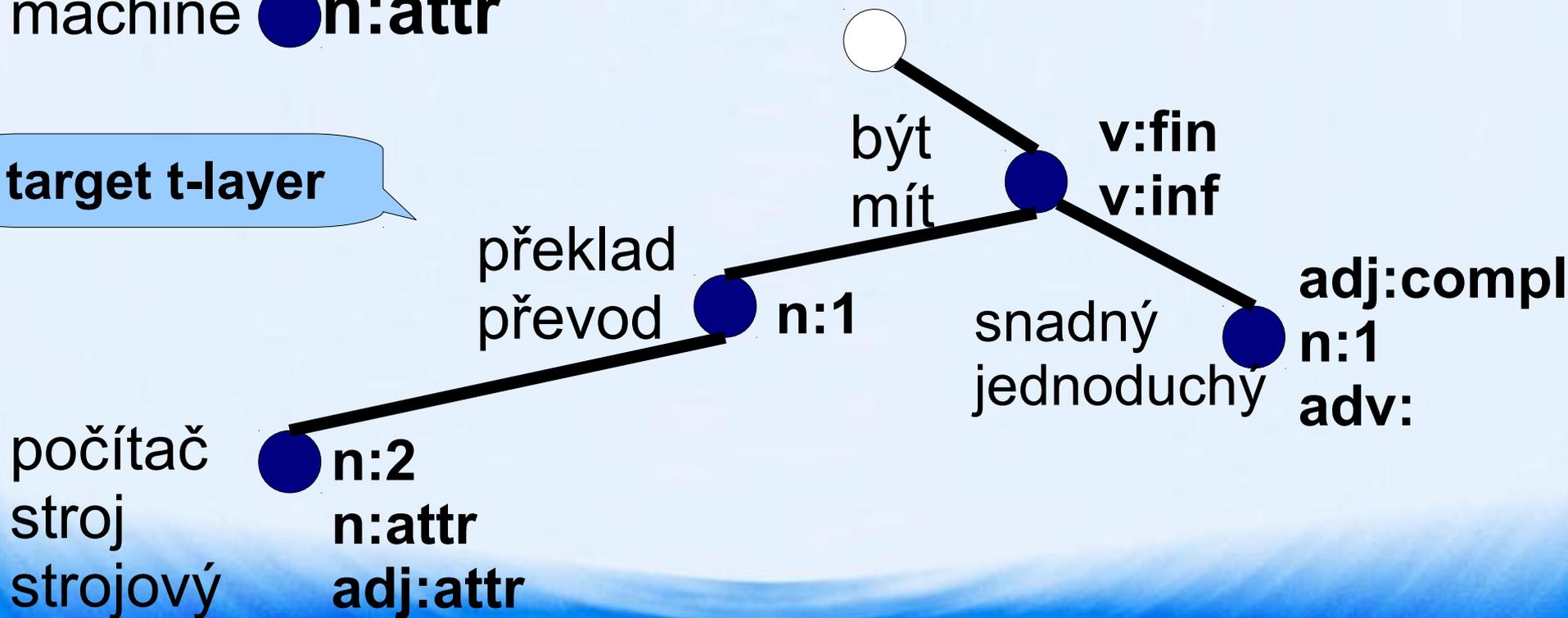
# Demo Translation – Transfer

Get translation variants for lemmas and formems

source t-layer



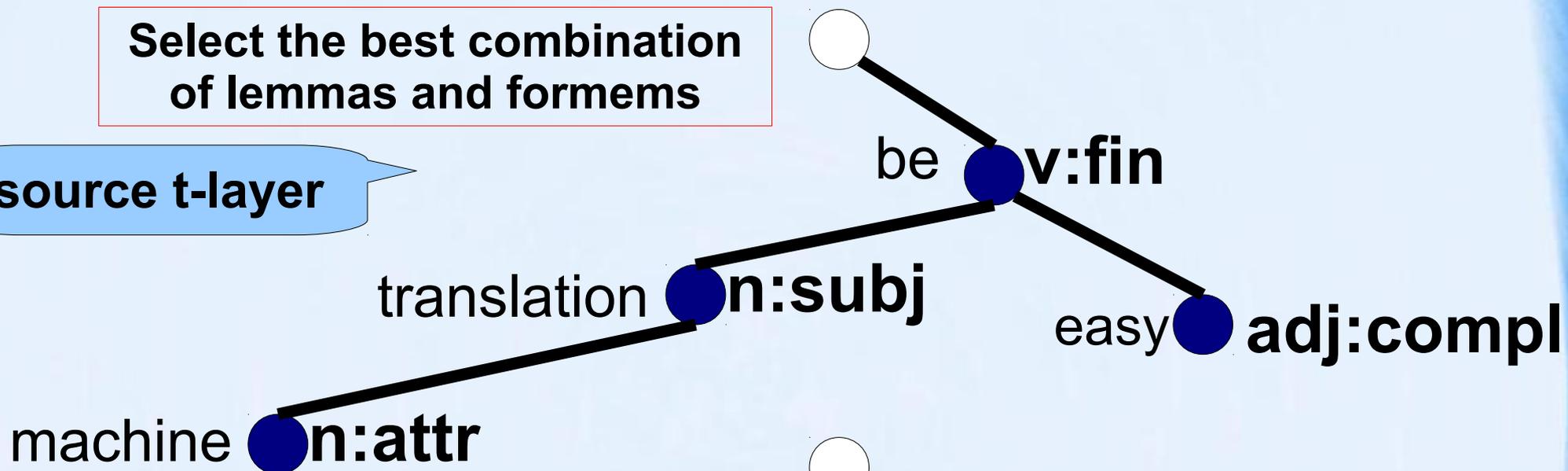
target t-layer



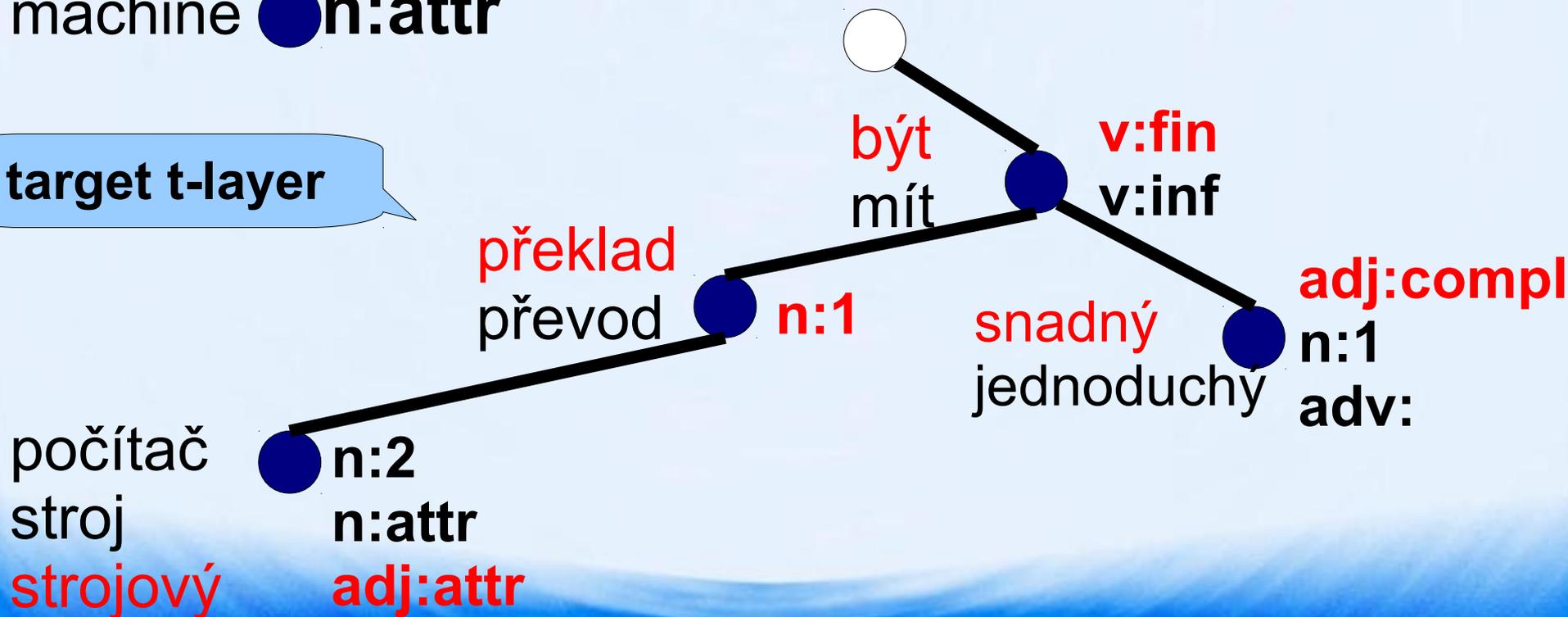
# Demo Translation – Transfer

Select the best combination of lemmas and formems

source t-layer



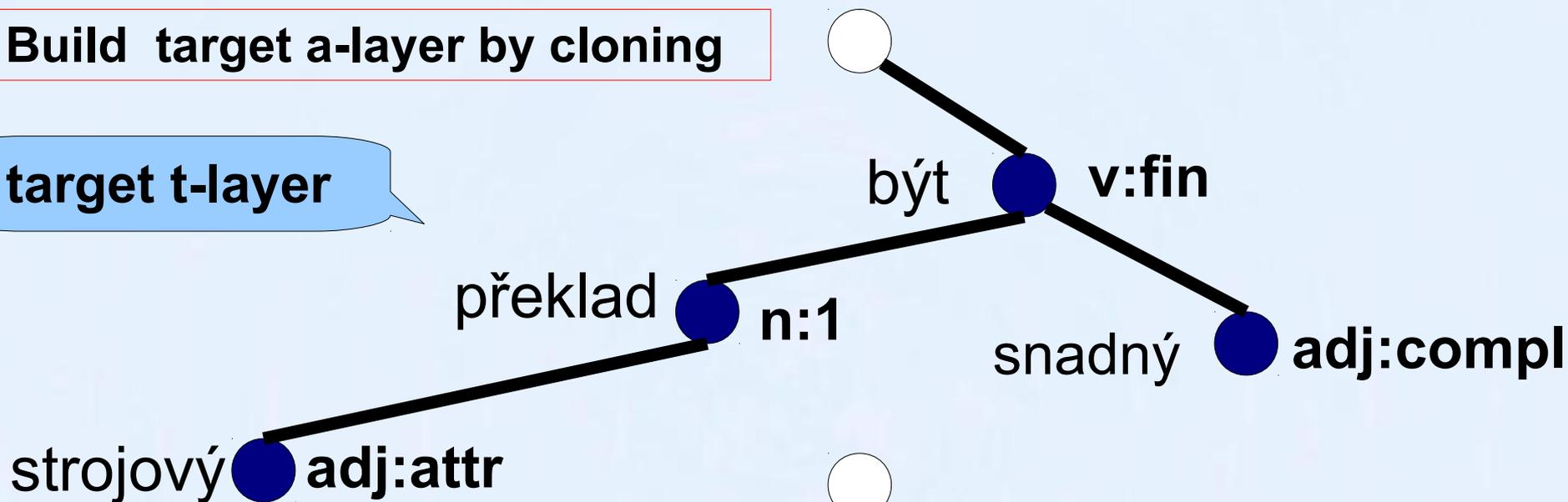
target t-layer



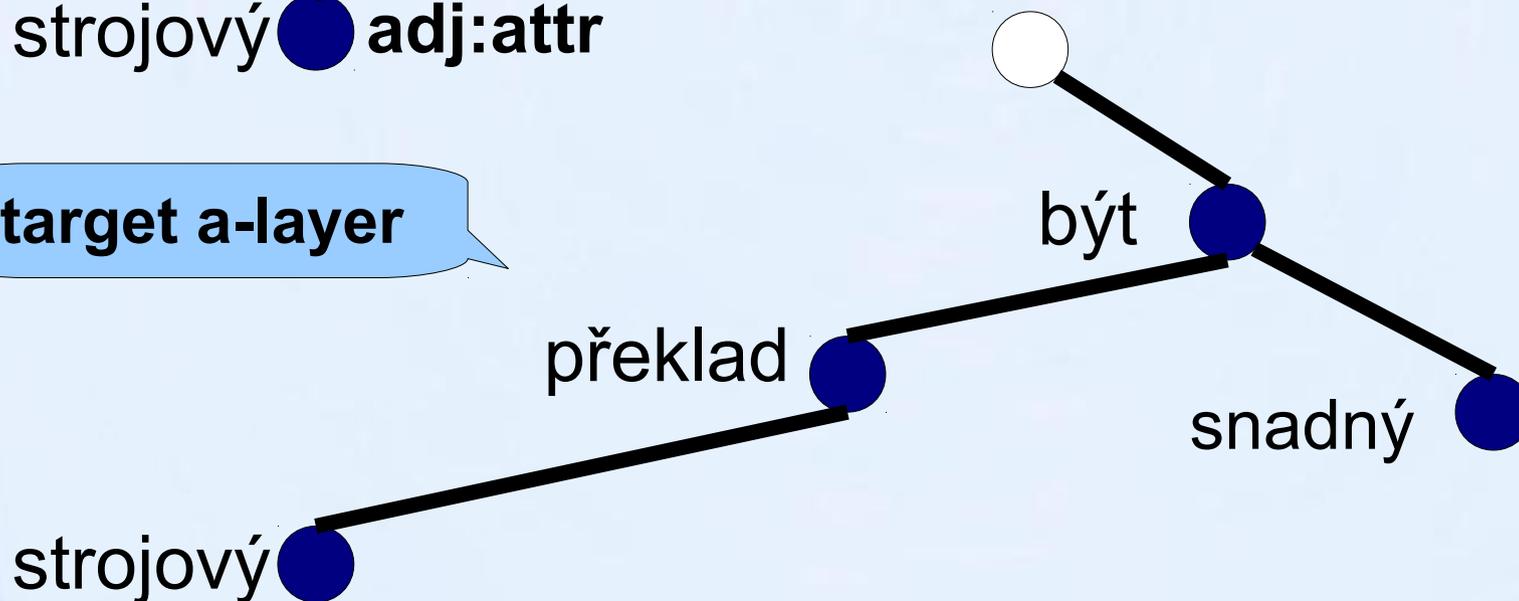
# Demo Translation – Synthesis

Build target a-layer by cloning

target t-layer



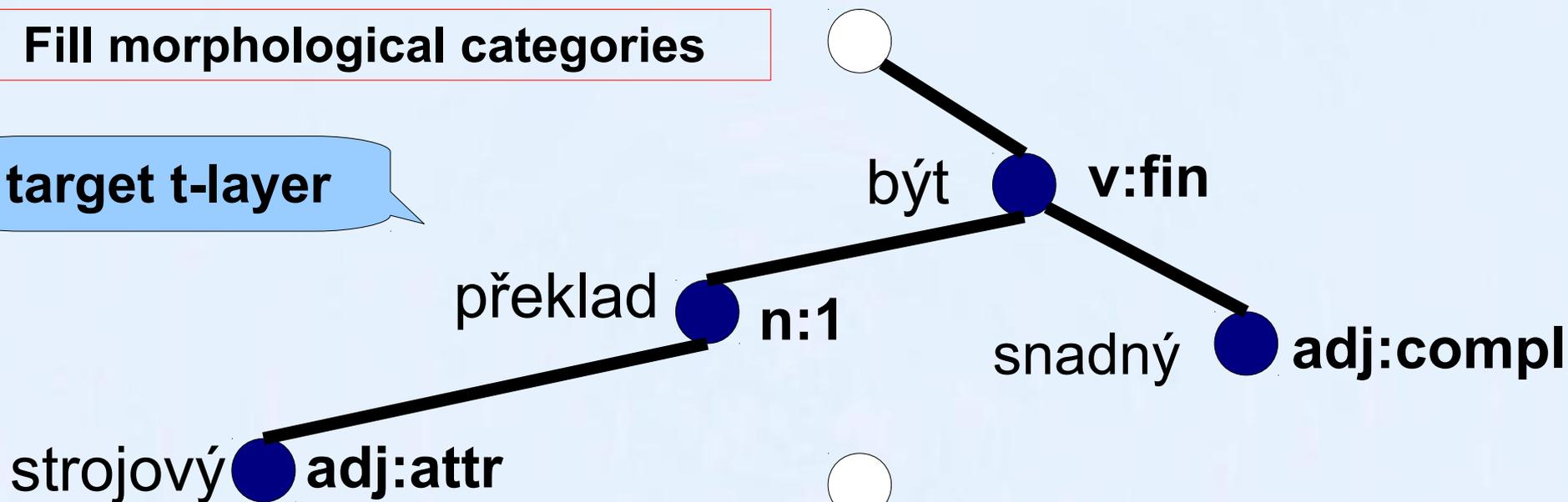
target a-layer



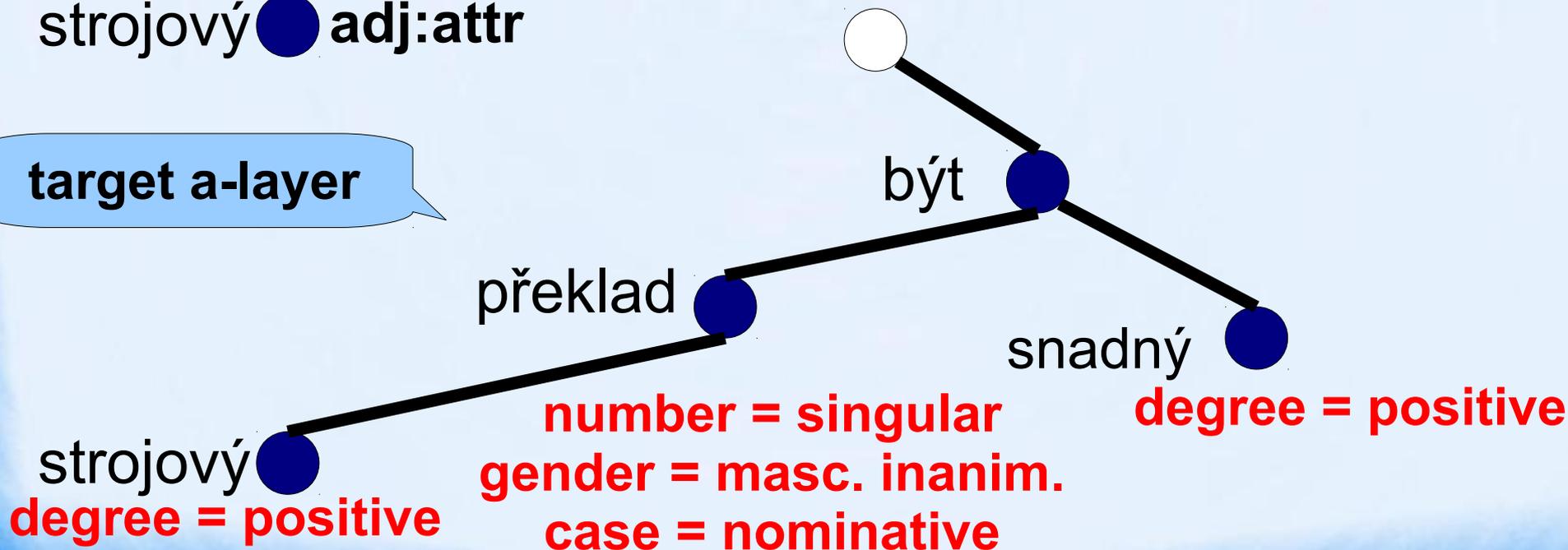
# Demo Translation – Synthesis

Fill morphological categories

target t-layer



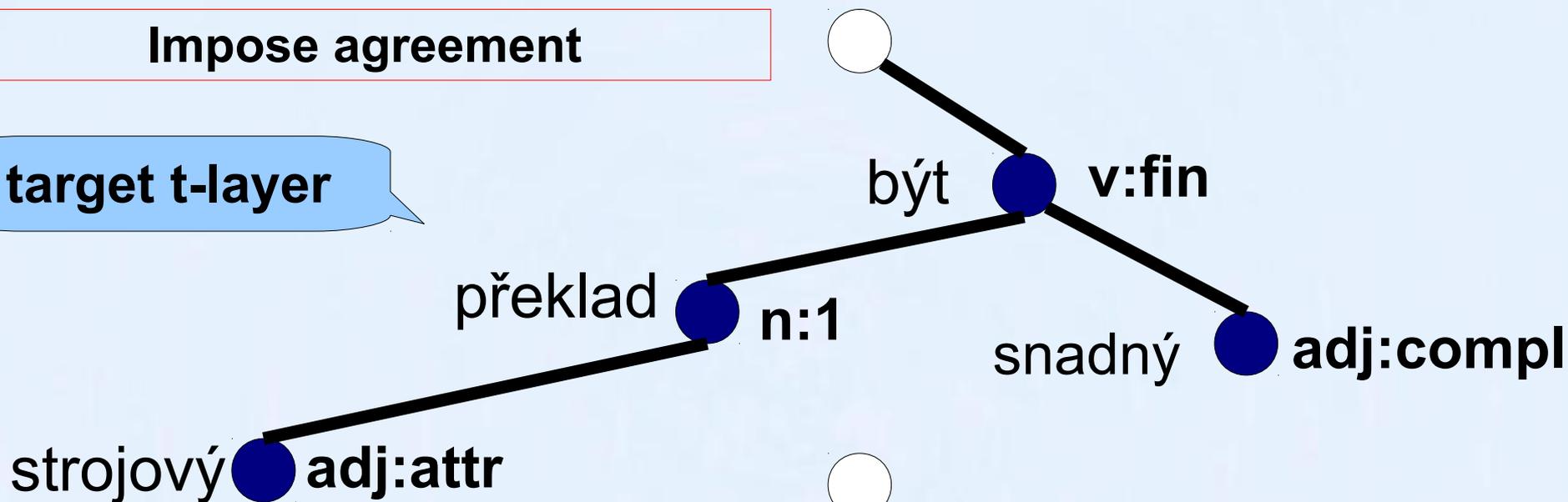
target a-layer



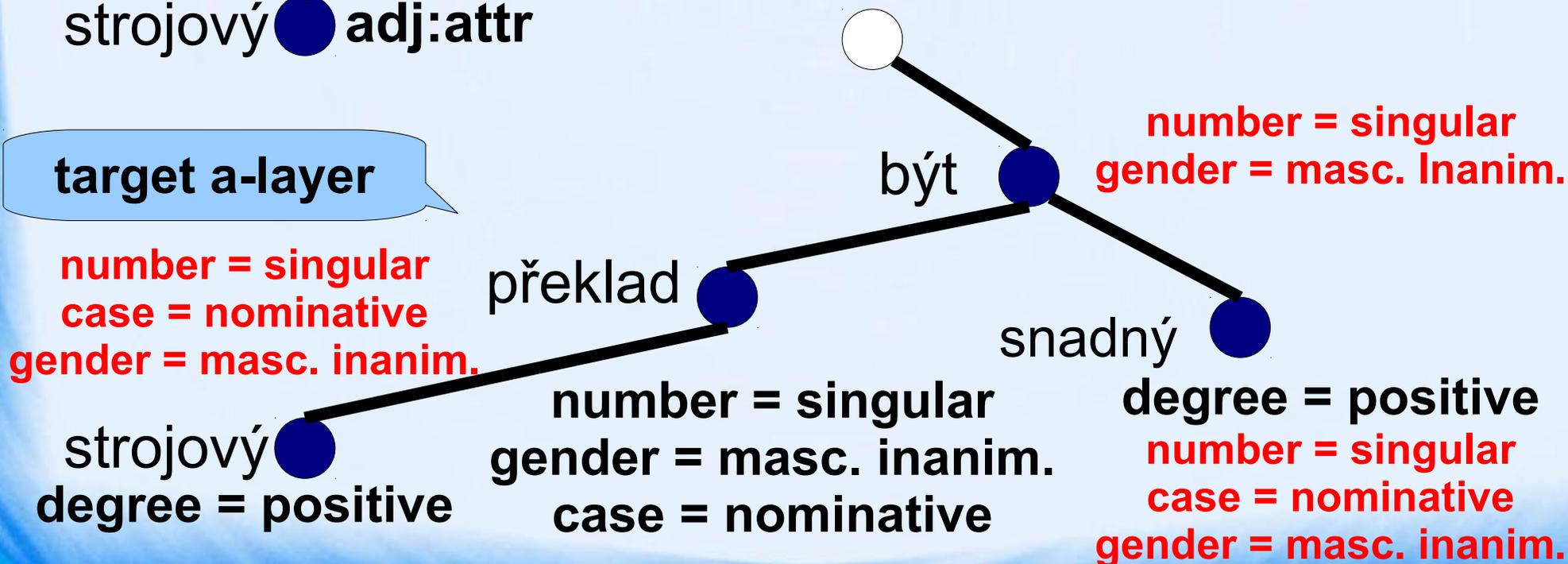
# Demo Translation – Synthesis

Impose agreement

target t-layer



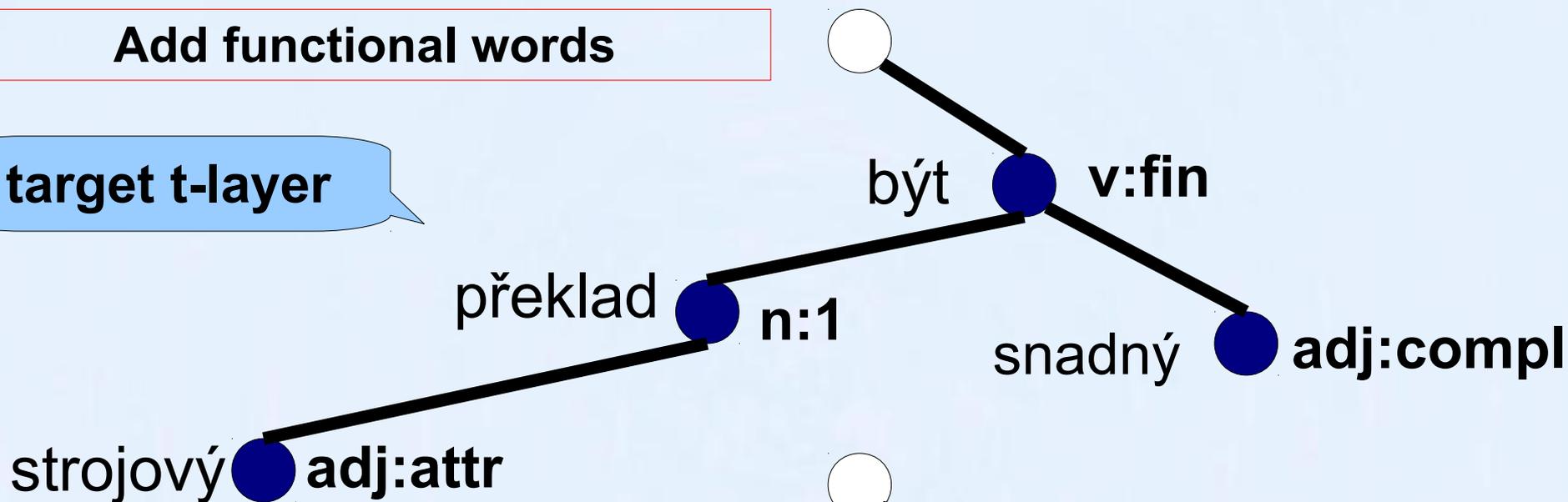
target a-layer



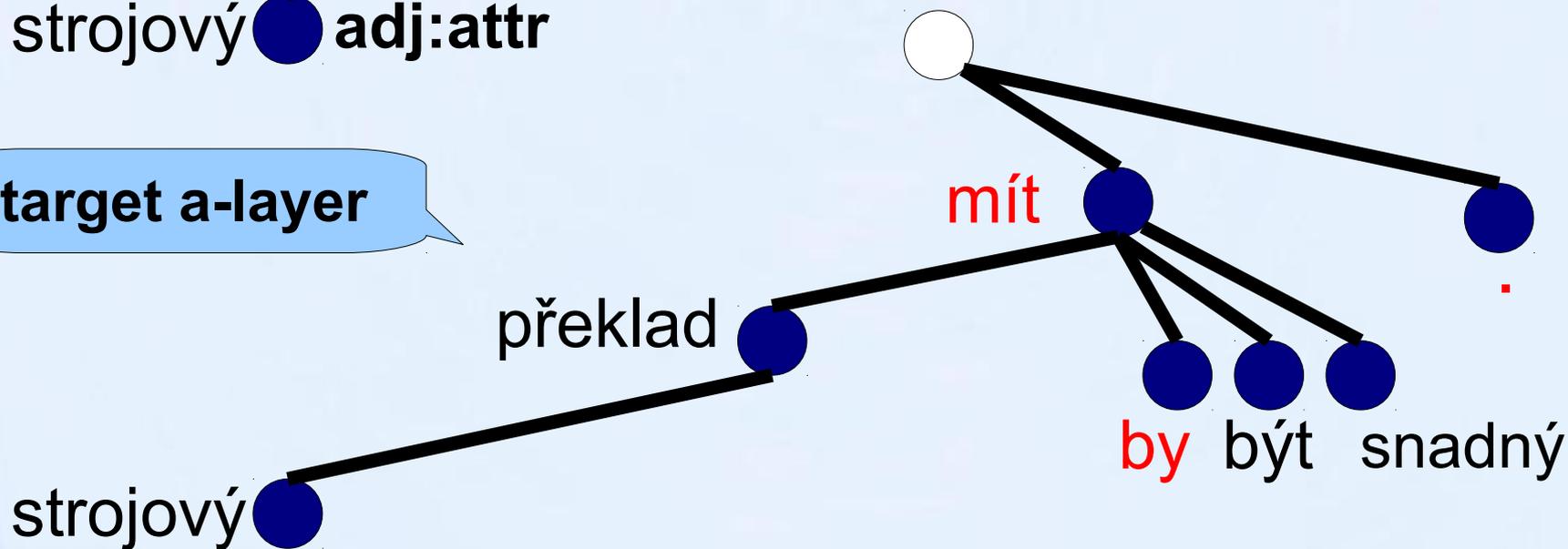
# Demo Translation – Synthesis

Add functional words

target t-layer



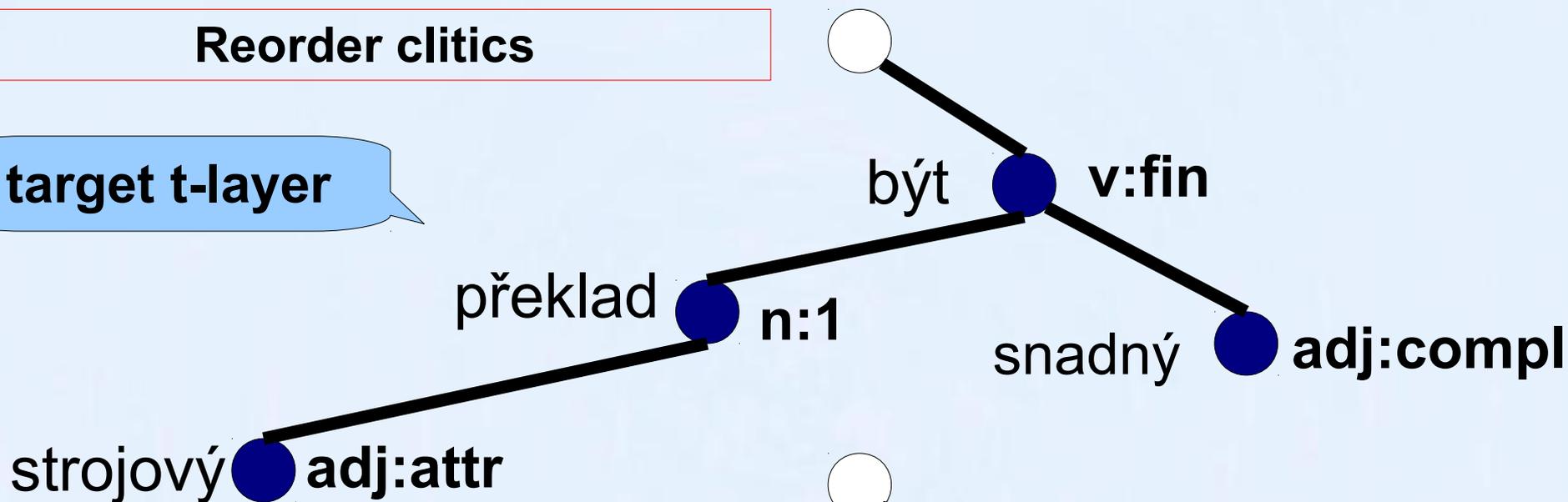
target a-layer



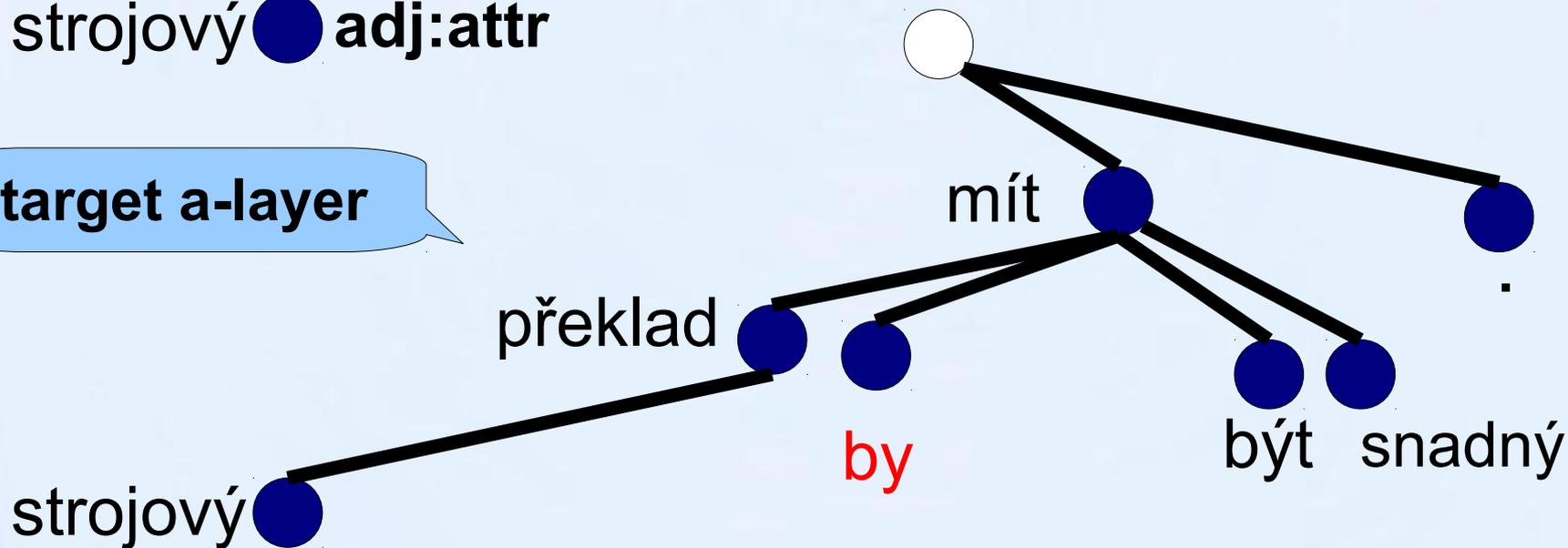
# Demo Translation – Synthesis

Reorder clitics

target t-layer



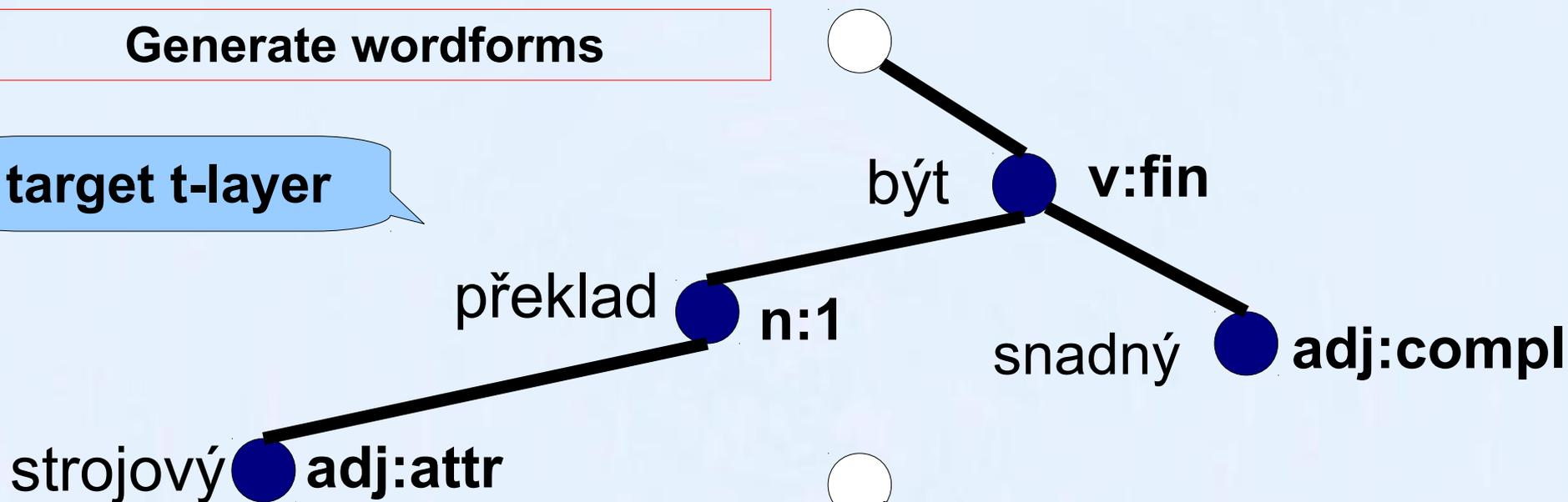
target a-layer



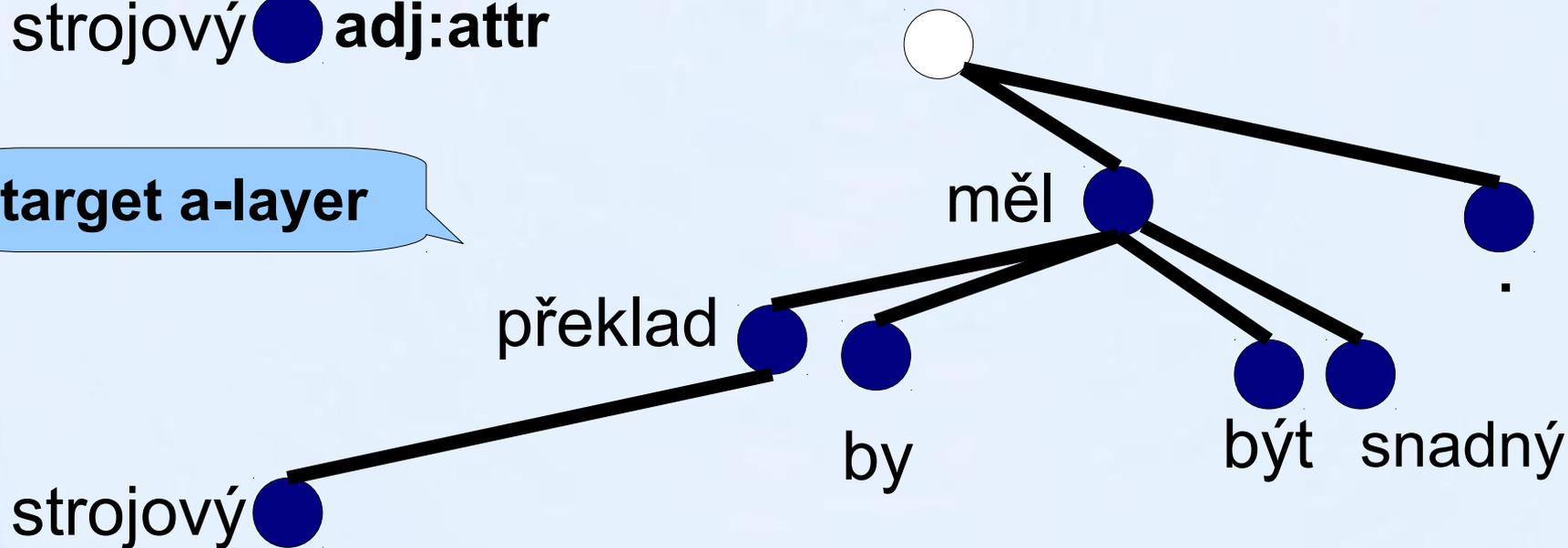
# Demo Translation – Synthesis

Generate wordforms

target t-layer



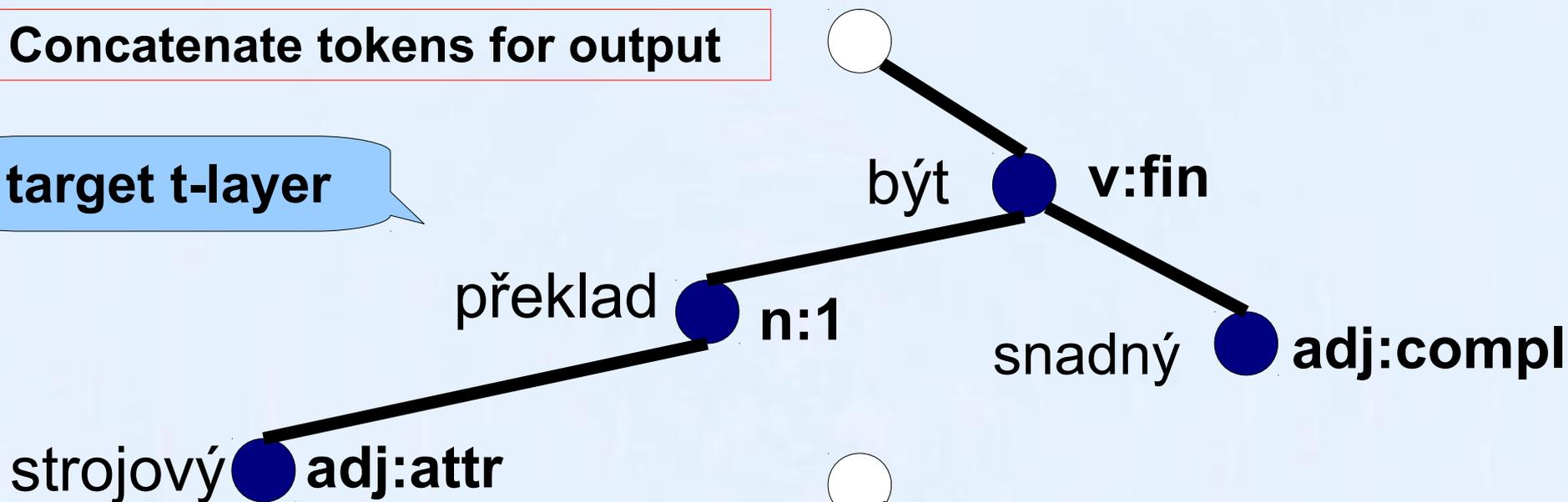
target a-layer



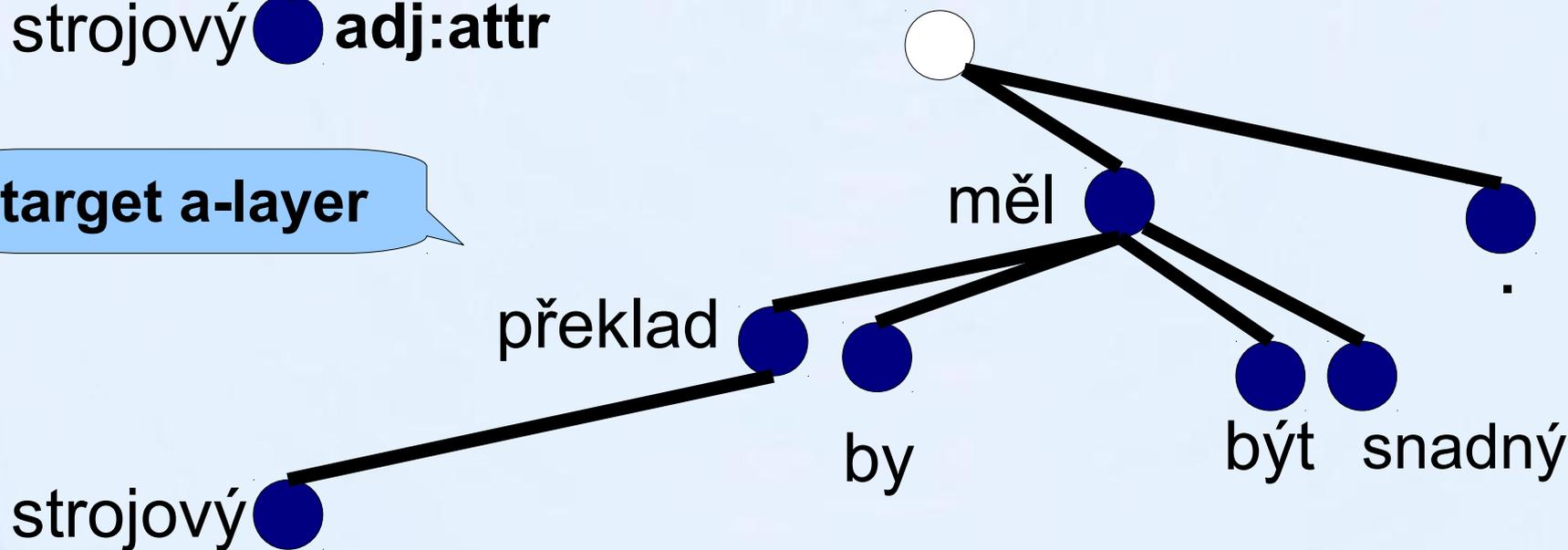
# Demo Translation – Synthesis

Concatenate tokens for output

target t-layer



target a-layer



**Strojový překlad by měl být snadný.**

# Demo Translation – Real Scenario



## MORPHOLOGY:

ResegmentSentences

Tokenize

NormalizeForms

FixTokenization

TagMorce

FixTags

Lemmatize

## NAMED ENTITIES:

StanfordNamedEntities

DistinguishPersonalNames

## A-LAYER:

MarkChunks

ParseMST

SetIsMemberFromDeprel

RehangConllToPdtStyle

FixNominalGroups

FixIsMember

FixAtree

FixMultiwordPrepAndConj

FixDicendiVerbs

SetAfunAuxCPCoord

SetAfun

## T-LAYER:

MarkEdgesToCollapse

MarkEdgesToCollapseNeg

BuildTtree

SetIsMember

MoveAuxFromCoordToMembers

FixTlemmas

SetCoapFunctors

FixEitherOr

FixIsMember

MarkClauseHeads

MarkPassives

SetFunctors

MarkInfin

MarkRelClauseHeads

MarkRelClauseCoref

MarkDspRoot

MarkParentheses

SetNodetype

SetGrammatemes

SetFormeme

RehangSharedAttr

SetVoice

FixImperatives

SetIsNameOfPerson

SetGenderOfPerson

AddCorAct

FindTextCoref

## TRANSFER:

CopyTtree

TrLFPPhrases

TrLFJointStatic

DeleteSuperfluousTnodes

TrFTryRules

TrFAddVariants

TrFRerank

TrLTryRules

TrLAddVariants

TrLFNumeralsByRules

TrLFilterAspect

TransformPassiveConstructions

PrunePersonalNameVariants

RemoveUnpassivableVariants

TrLFCompounds

CutVariants

RehangToEffParents

TrLFTreeViterbi

RehangToOrigParents

CutVariants

FixTransferChoices

ReplaceVerbWithAdj

DeletePossPronBeforeVlastni

TrLFemaleSurnames

AddNounGender

MarkNewRelClauses

AddRelpronBelowRc

ChangeCorToPersPron

AddPersPronBelowVfin

AddVerbAspect

FixDateTime

FixGrammatemesAfterTransfer

FixNegation

MoveAdjsBeforeNouns

MoveGenitivesRight

MoveRelClauseRight

MoveDicendiCloserToDsp

MovePersPronNextToVerb

MoveEnoughBeforeAdj

MoveJesteBeforeVerb

FixMoney

OverridePpWithPhraseTr

FindGramCorefForRefIpron

NeutPersPronGenderFromAntec

ValencyRelatedRules

SetClauseNumber

TurnTextCorefToGramCoref

## SYNTHESIS TO A-LAYER:

CopyTtree

DistinguishHomonymous.

ReverseNumberNounDep.

InitMorphcat

FixPossessiveAdjs

MarkSubject

ImposePronZAgr

ImposeRelPronAgr

ImposeSubjpredAgr

ImposeAttrAgr

ImposeComplAgr

DropSubjPersProns

AddPrepos

AddSubconj

AddReflexParticles

AddAuxVerbCompoundPassive

AddAuxVerbModal

AddAuxVerbCompoundFuture

AddAuxVerbConditional

AddAuxVerbCompoundPast

AddClausalExpletivePronouns

ResolveVerbs

ProjectClauseNumber

AddParentheses

AddSentFinalPunct

AddSubordClausePunct

AddCoordPunct

AddAppositionPunct

ChooseMlemmaForPersPron

GenerateWordforms

MoveCliticsToWackernagel

DeleteSuperfluousPrepos

DeleteEmptyNouns

VocalizePrepos

CapitalizeSentStart

CapitalizeNamedEntities.

FillTagFromMorphcat

## SYNTHESIS TO TEXT:

ConcatenateTokens

ApplySubstitutions

DetokenizeUsingRules

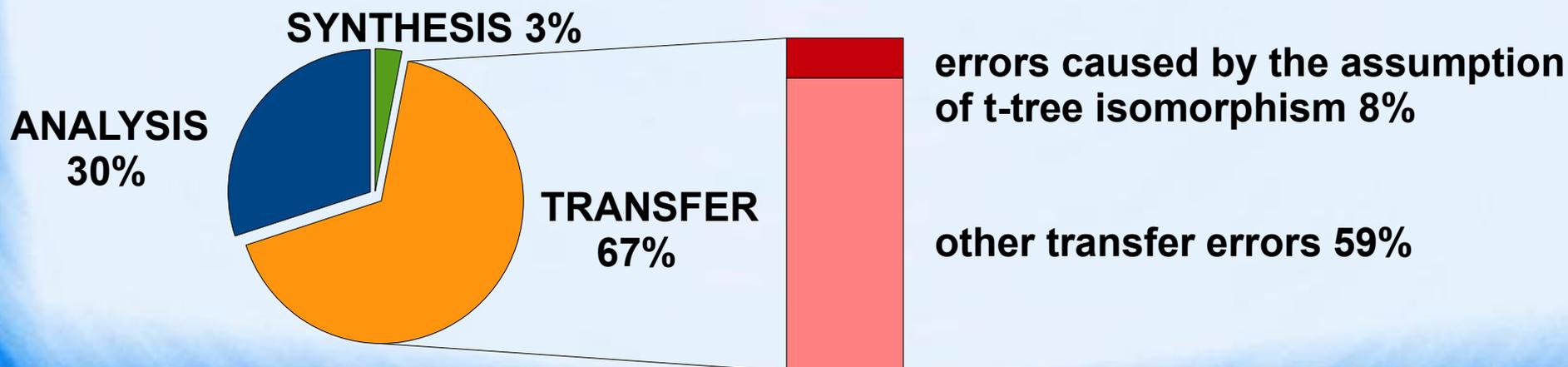
RemoveRepeatedTokens

NormalizePunctuationForWMT

# Annotation of Translation Errors

sample of 250 sentences, 1463 errors in total

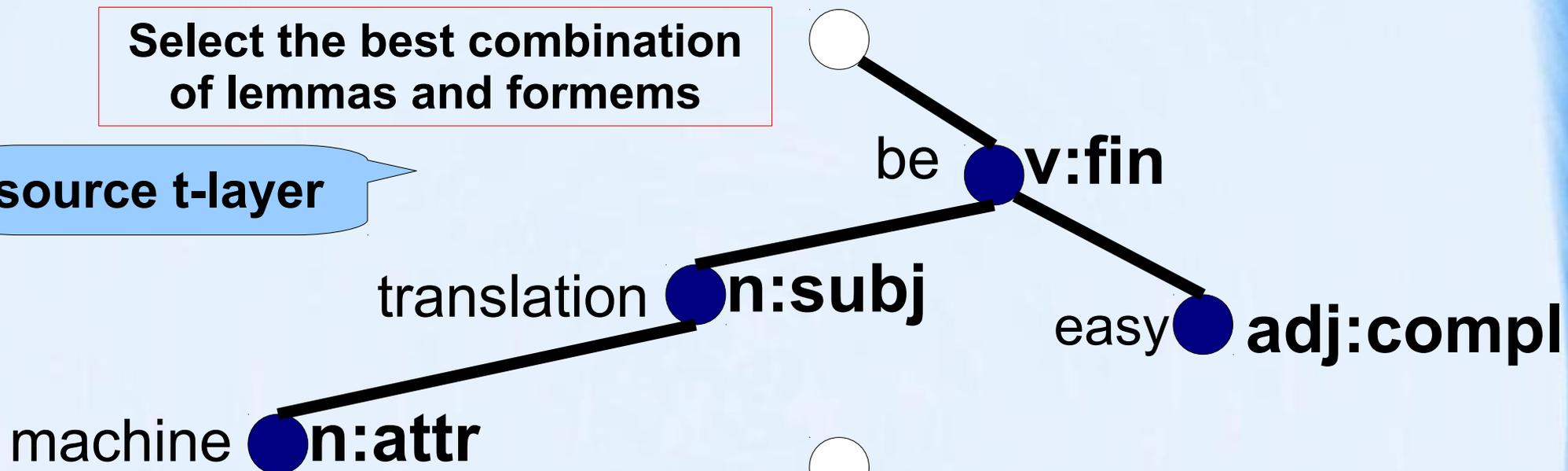
<b>Type</b>	lemma, formeme, gram., w. order,...
<b>Subtype</b>	gram: gender, person, tense,...
<b>Seriousness</b>	serious, minor
<b>Circumstances</b>	coordination, named entity, numbers
<b>Source</b>	tok, lem, tagger, parser, tecto, trans, x, syn, ?



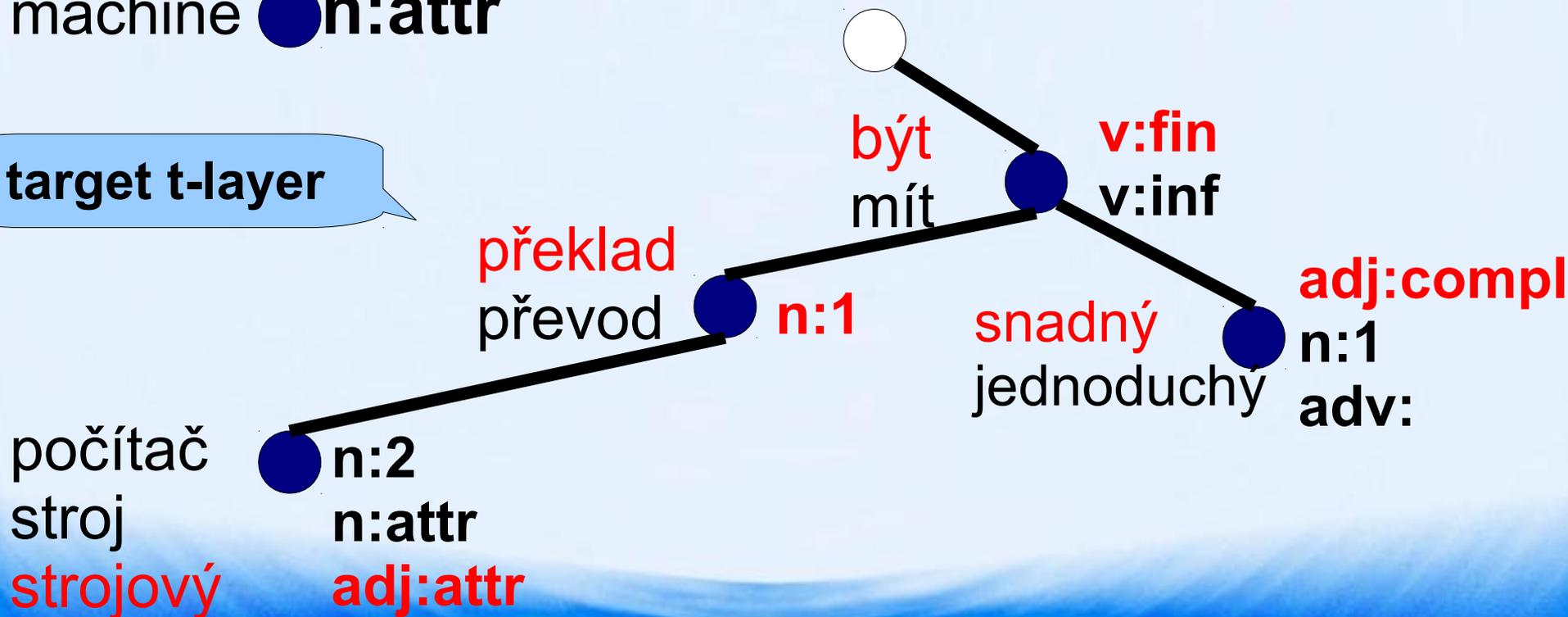
# HMTM – Motivation

Select the best combination of lemmas and formems

source t-layer



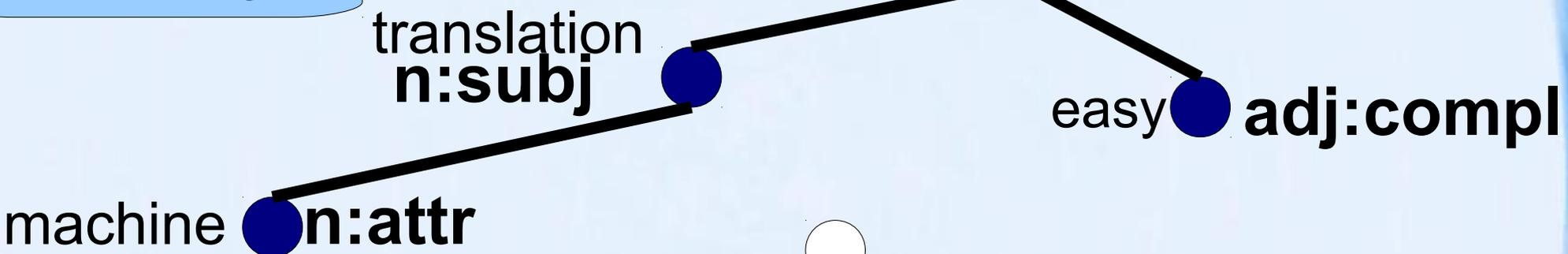
target t-layer



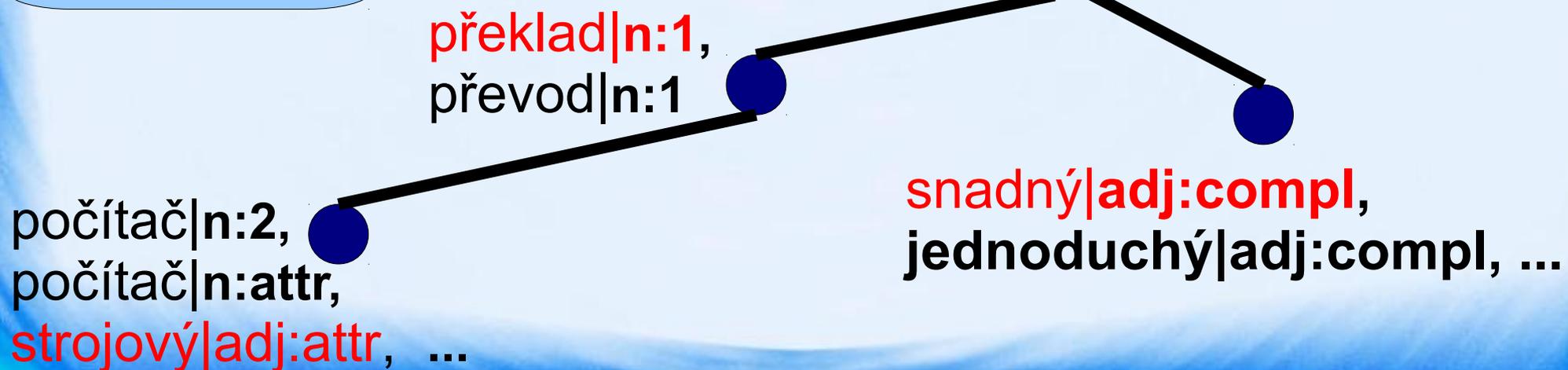
# HMTM – Motivation

Select the best label for each node

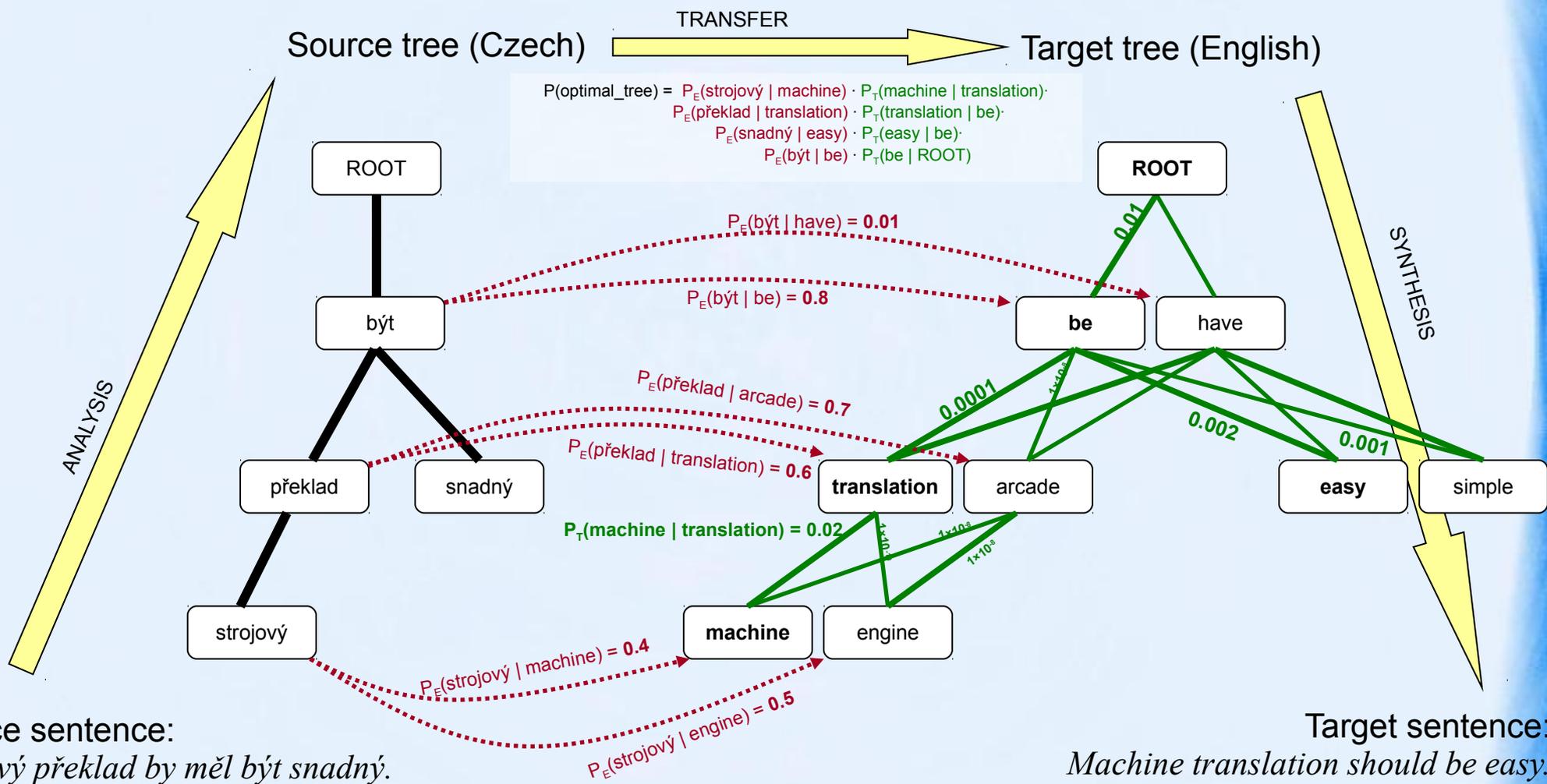
source t-layer



target t-layer



# HMTM in MT



$P_E(\text{source | target})$  ... emission probabilities ... **translation model**  
 $P_T(\text{dependent | governing})$  ... transition probabilities ... **target-language tree model**

# Combining Dictionaries

- new general interface (for lemmas and formems)  
`$dict->get_translations($input_label, $features)`  
returns a list of translation variants including probabilities
- OOP style, dictionary constructor can take another dictionary (or more) as a parameter → hierarchy

- Four basic types of dictionaries:

**Static plain**

loaded from a file „lemma → lemma“

**Context**

loaded from a file „lemma,features → lemma“

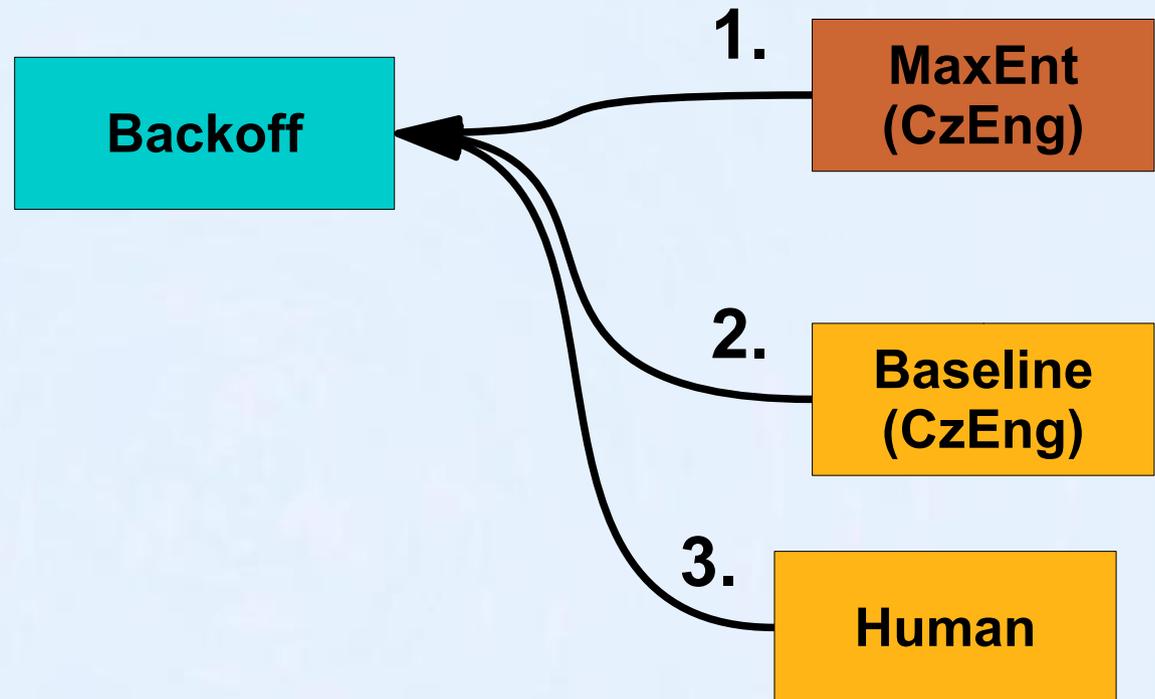
**Derivational**

translations derived dynamically, input dictionary

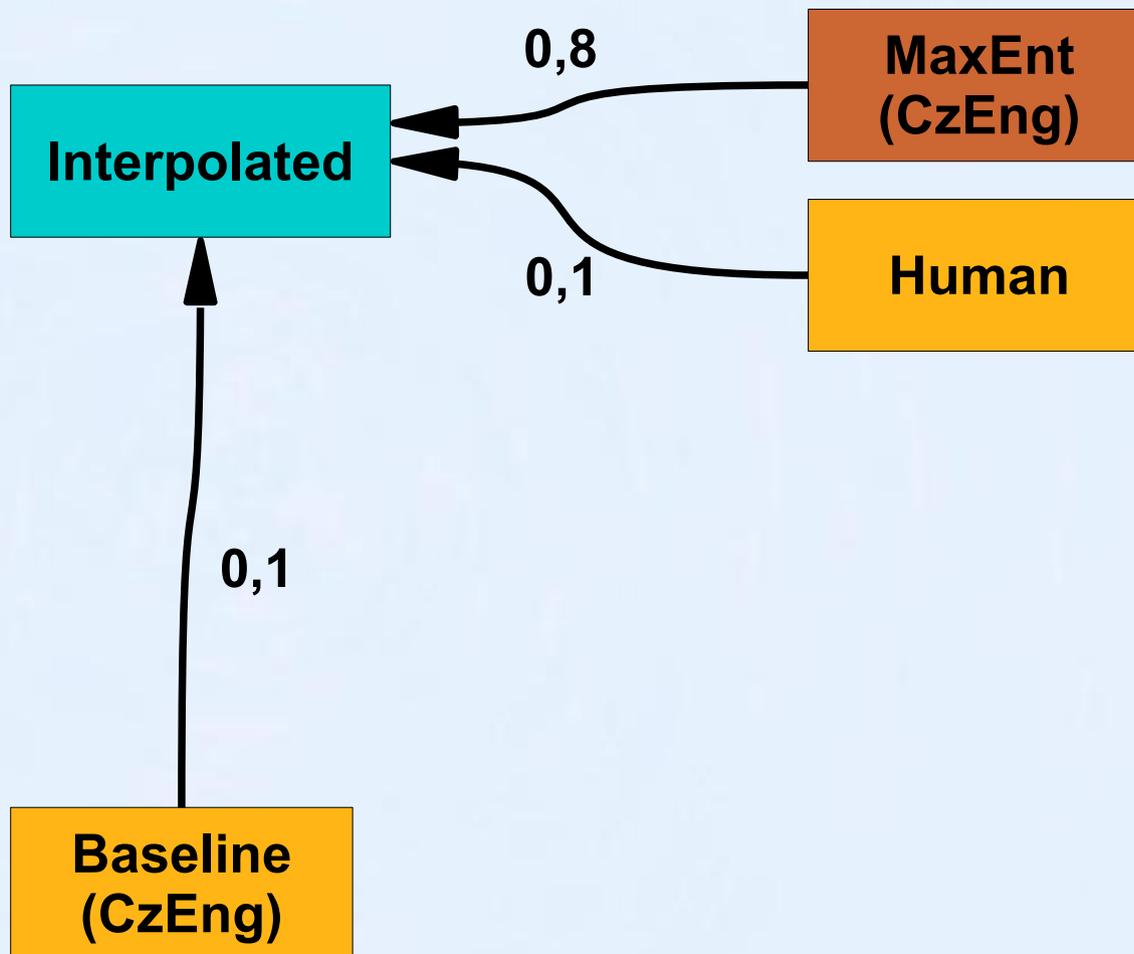
**Combinational**

combination of more input dictionaries

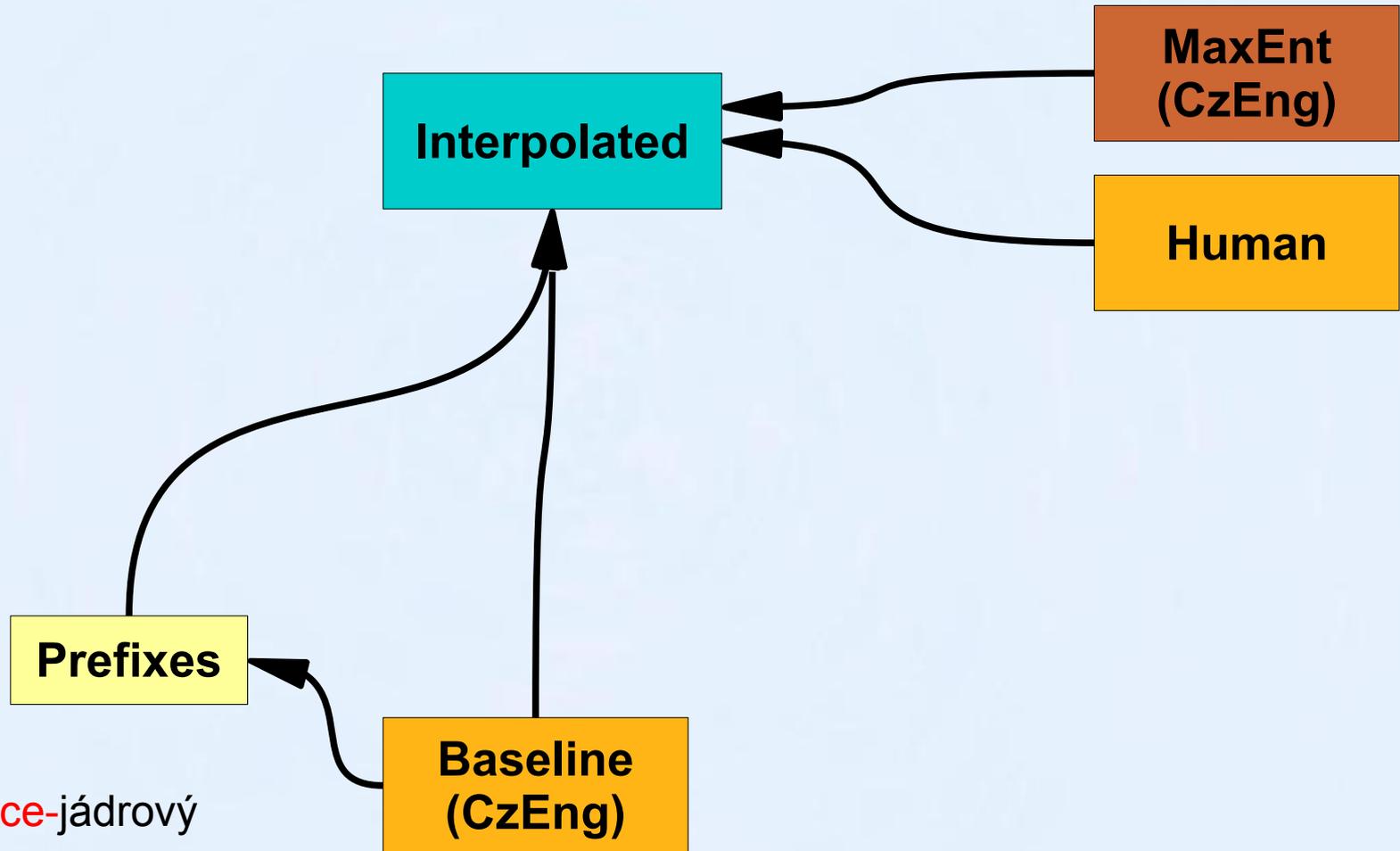
# Hierarchy of lemma dictionaries



# Hierarchy of lemma dictionaries

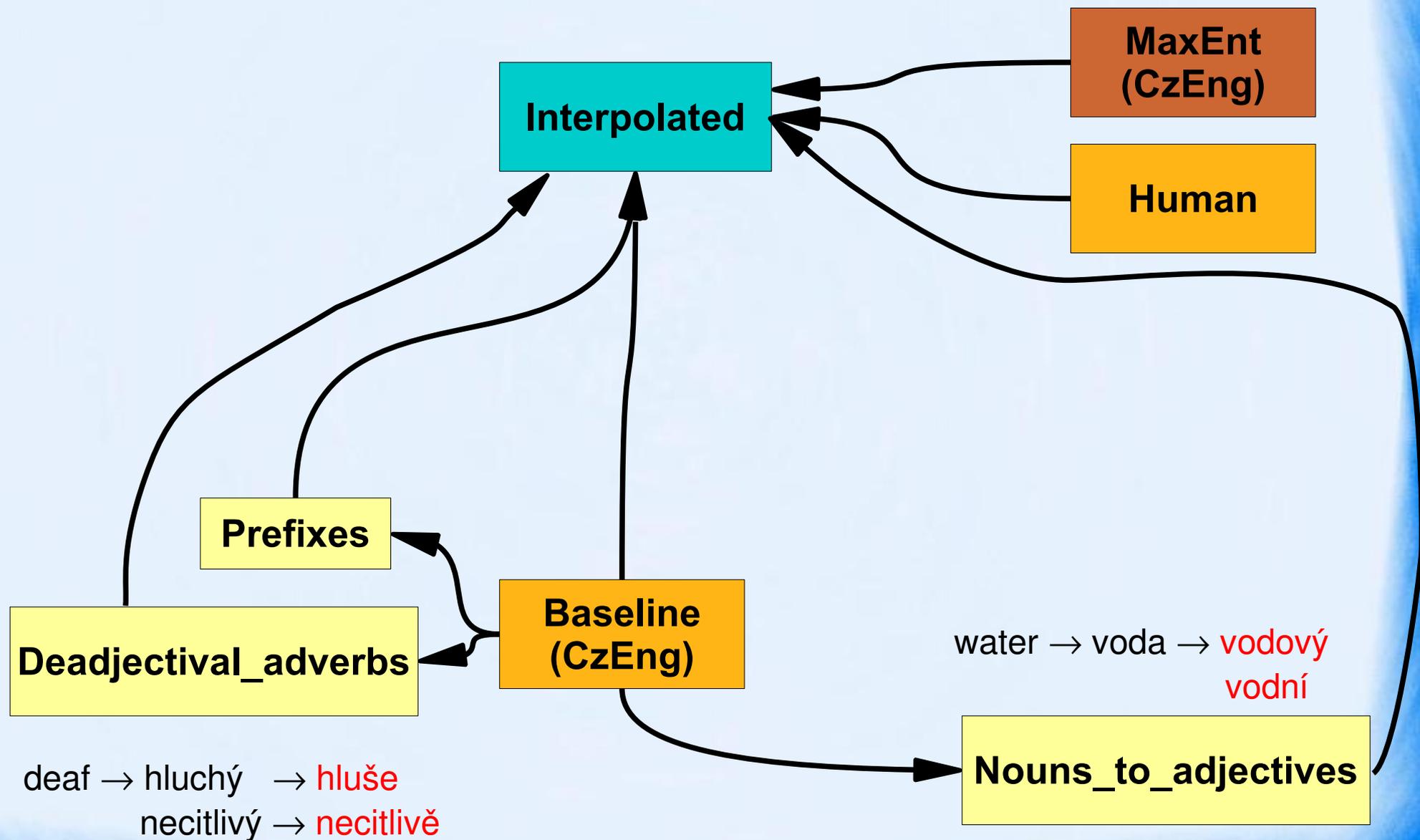


# Hierarchy of lemma dictionaries

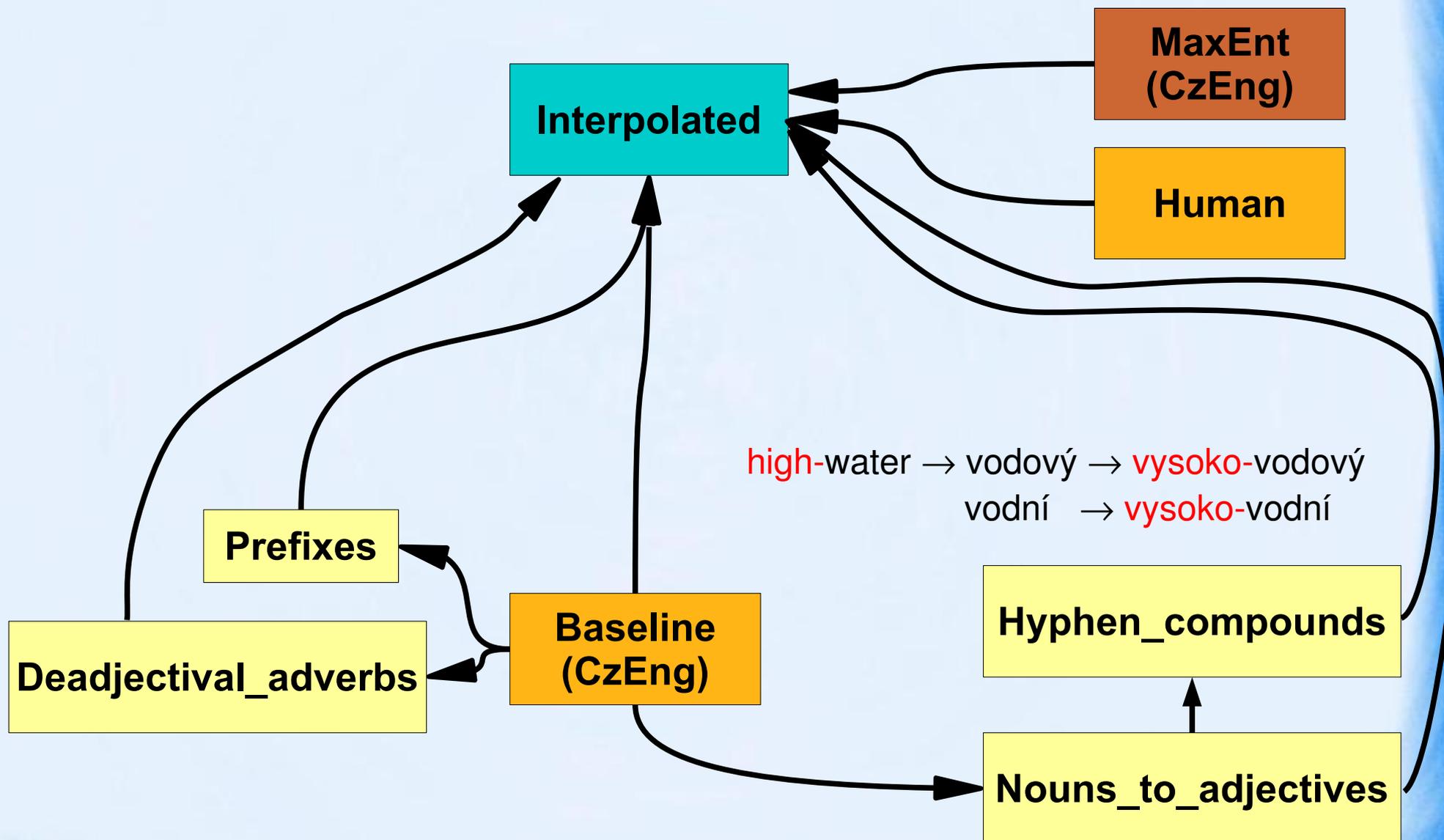


multi-core → více-jádrový  
více-jádro  
multi-jádrový  
multi-jádro

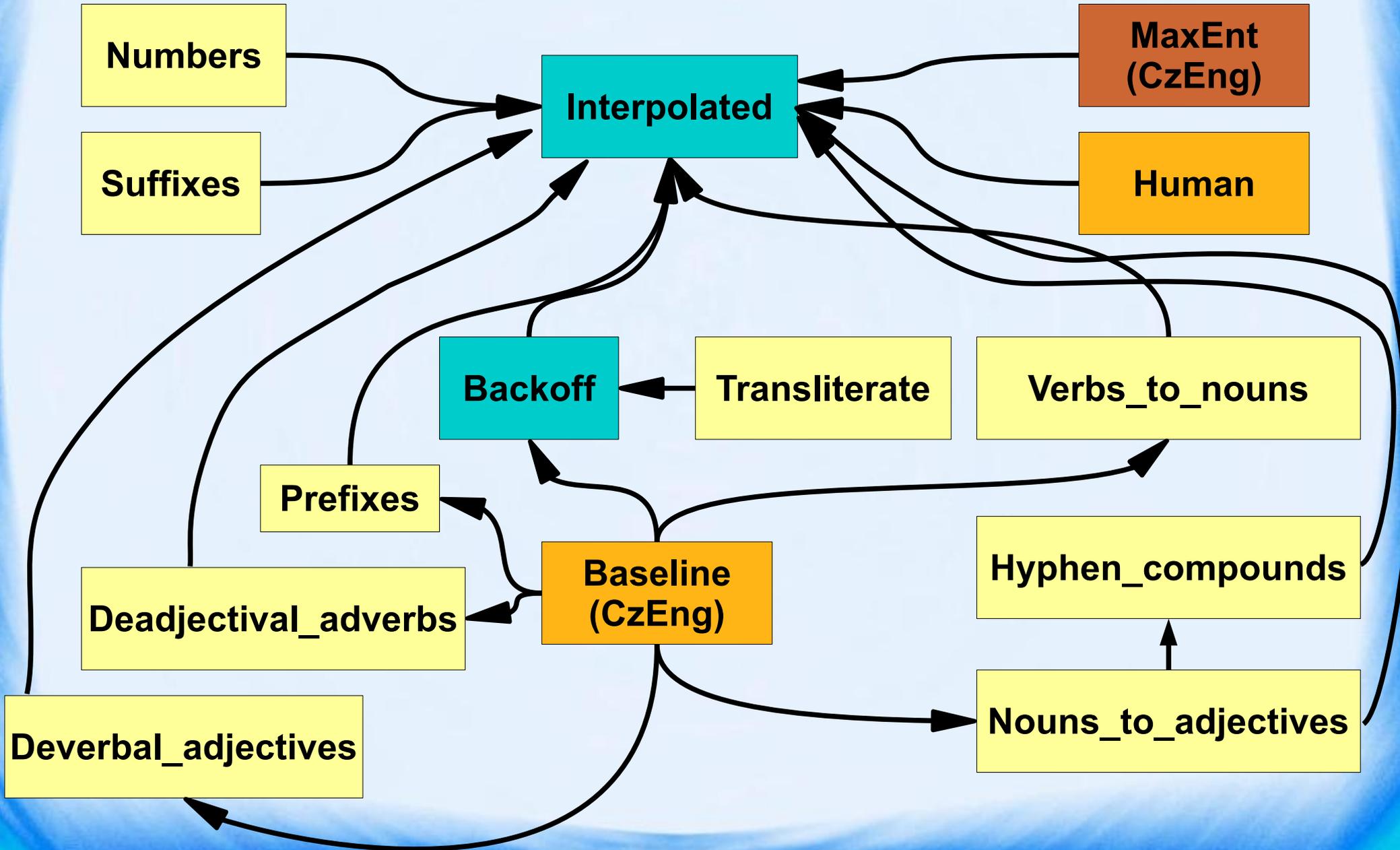
# Hierarchy of lemma dictionaries



# Hierarchy of lemma dictionaries



# Hierarchy of lemma dictionaries



# Maximum Entropy Dictionary

## Baseline Dictionary

$$p(y|x) = \frac{\text{count}(x, y)}{\text{count}(x)}$$

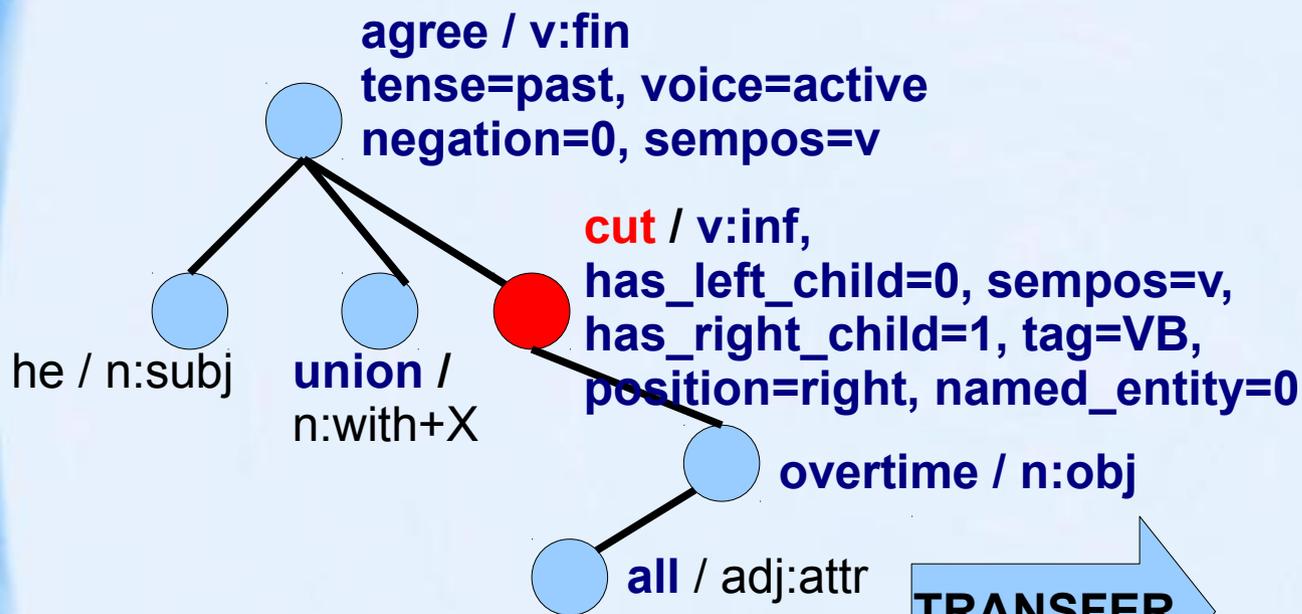
- Maximum likelihood estimates  
(from the training sections of CzEng 0.9)
- Pruned by thresholds on  $p(x|y)$  and  $p(y|x)$
- No context used  
x = source lemma  
y = target lemma

## MaxEnt Dictionary

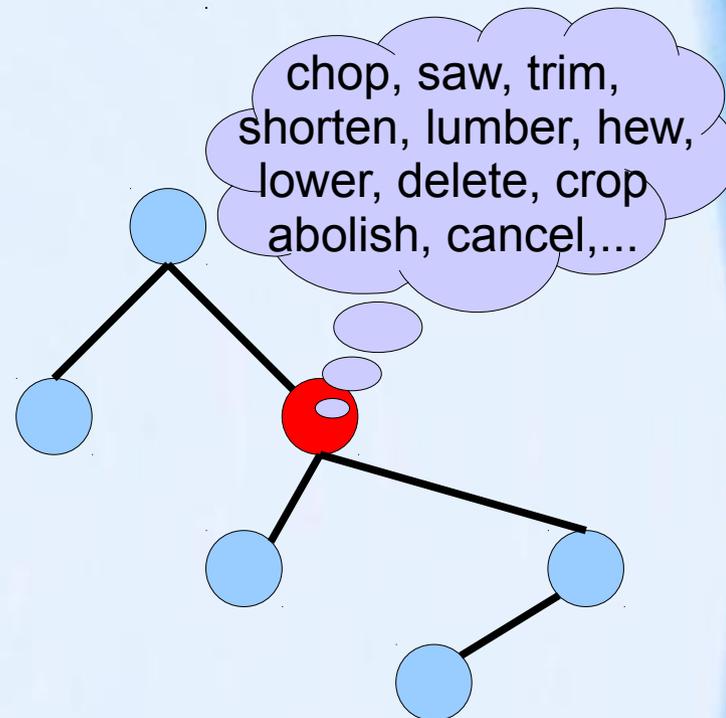
$$p(y|x) = \frac{1}{Z(x)} \exp \sum_i \lambda_i f_i(x, y)$$

- One MaxEnt model for each source lemma  
(same training data as for the Baseline Dict.)
- Interpolated with Baseline Dict. (due to pruning)
- Context features used (x = source context)
  - local tree context
  - local linear context
  - morphological & syntactic categories
  - ...

# Maximum Entropy Dictionary



TRANSFER



ANALYSIS

SYNTHESIS

He agreed with the unions to cut all overtime.

Dohodl se s odbory na zrušení všech přesčasů.

# Examples of Translation (2009)

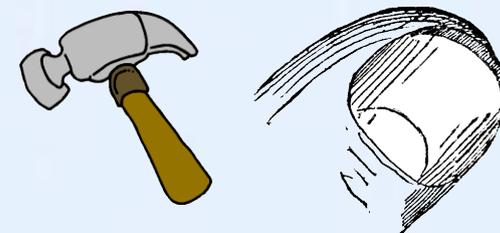
A miss by an inch  
is a miss by a mile.

Slečna palec je slečna miliónu.



I'd rather be a hammer  
than a nail.

Spíše bych byl kladivo než nehet.



A bird in the hand is worth  
two in the bush.

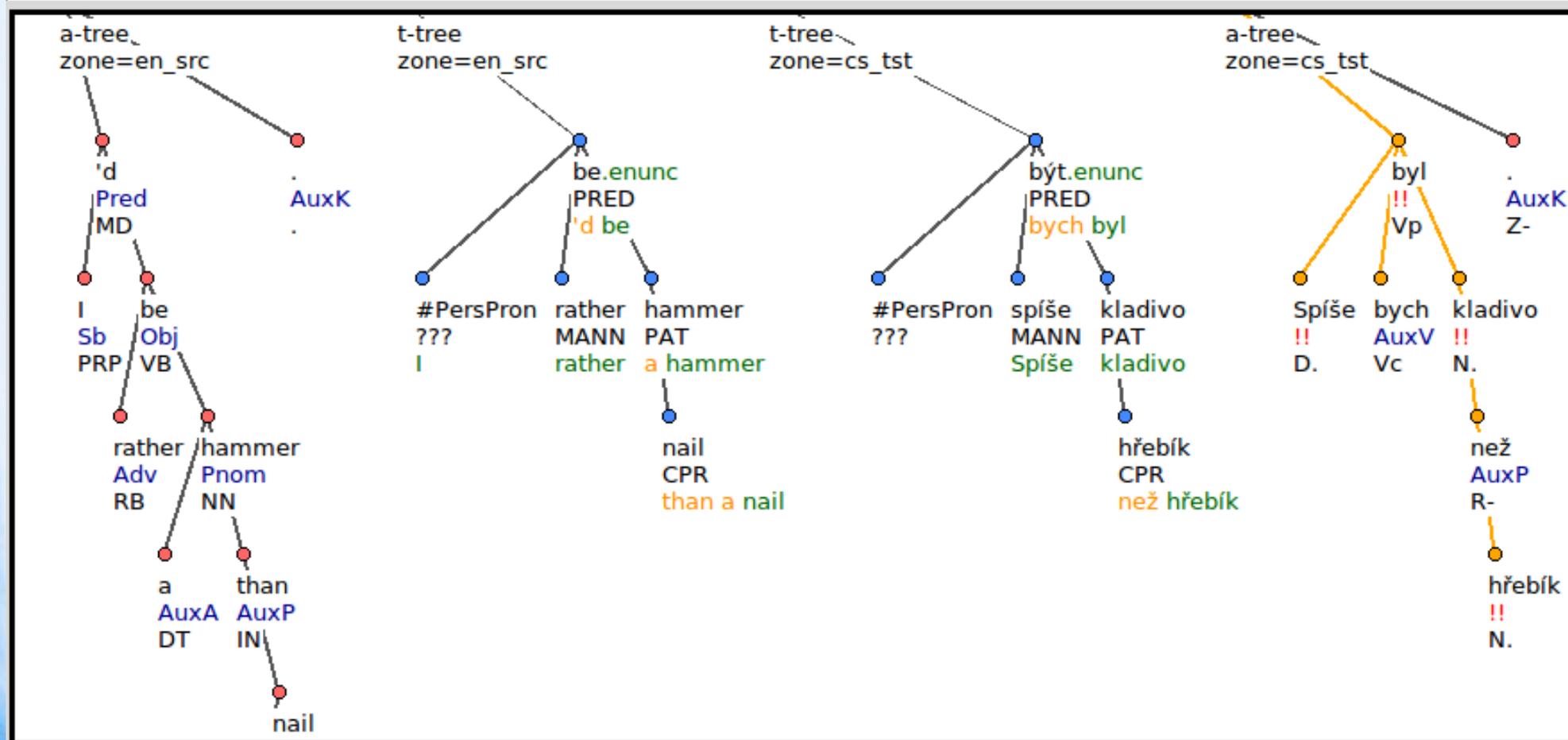
Pták v ruce je cenný  
dvakrát v Bushovi.



# Example of Translation (2011)

File Node Tree View Macros Setup Help

[cs\_tst] Spíše bych byl kladivo než hřebík .  
 [en\_src] I'd rather be a hammer than a nail.



The image displays four syntax trees illustrating the translation process:

- Tree 1 (a-tree, zone=en\_src):** Represents the English source sentence. The root node is 'd (Pred MD), which branches into 'I (Sb PRP) and 'be (Obj VB). 'be branches into 'rather (Adv RB) and 'hammer (Pnom NN). 'hammer branches into 'a (AuxA DT) and 'than (AuxP IN). 'than branches into 'nail (CPR).
- Tree 2 (t-tree, zone=en\_src):** Represents the English target sentence. The root node is 'be.enunc (PRED), which branches into '#PersPron (MANN) and 'rather (PAT). '#PersPron branches into 'I (CPR). 'rather branches into 'rather (CPR). 'hammer branches into 'a (CPR) and 'hammer (CPR).
- Tree 3 (t-tree, zone=cs\_tst):** Represents the Czech target sentence. The root node is 'být.enunc (PRED), which branches into '#PersPron (MANN) and 'spíše (PAT). '#PersPron branches into '??? (CPR). 'spíše branches into 'spíše (CPR). 'kladivo branches into 'kladivo (CPR).
- Tree 4 (a-tree, zone=cs\_tst):** Represents the Czech target sentence. The root node is 'byl (Vp), which branches into 'Spíše (AuxV D.), 'bych (AuxV Vc), and 'kladivo (N.). 'kladivo branches into 'než (AuxP R-) and 'hřebík (N.).

# Sample of MaxEnt Features

input\_label=nail

**output\_label=hřebík#N (metal nail)**

child_formeme_n:in+X=1	1.64483855116042
is_member=1	1.30042900630692
child_formeme_v:fin=1	1.04422203176176
next_node_tlemma=down	0.838961007712912
is_capitalized=1	0.792130821958927
<b>position=right</b>	<b>0.747785245407306</b>
tense_g=post	0.744919903760696
<b>voice_g=active</b>	<b>0.659489975893991</b>
prev_node_tlemma=drive	0.655357850937254
parent_capitalized=1	0.622953832124697
formeme=n:from+X	0.599348506643414
<b>prev_node_tlemma=hammer</b>	<b>0.592276691427986</b>
child_tlemma_few=1	0.553464629114697
child_tlemma_remove=1	0.546698831608057
sempos=n.denot	0.504719359514573
next_node_tlemma=and	0.502529618088752
formeme_g=v:until+fin	0.491064112122981
child_tlemma_rusty=1	0.428884558837039
tag_g=VBP	0.422967377093101
next_node_tlemma=screw	0.344701934524519
...	

**output\_label=nehet#N (fingernail or toenail)**

child_formeme_n:poss=1	1.32717038827268
child_tlemma_finger=1	1.07509772743853
child_formeme_n:of+X=1	0.982021327950337
position=left	0.886912864256063
prev_node_tlemma=black	0.770671304450658
child_tlemma_broken=1	0.761077744287099
child_formeme_v:attr=1	0.700099311992958
formeme=n:at+X	0.674547829214778
formeme_g=n:attr	0.673367412957367
child_tlemma_long=1	0.673158400394094
next_node_tlemma=file	0.600496248030202
child_tlemma_false=1	0.584236638145312
prev_node_tlemma=false	0.584236638145312
<b>number=sg</b>	<b>0.563056142428995</b>
formeme=n:obj	0.533943098032196
formeme=n:by+X	0.528852315800188
...	

# Cooperation is welcome

- Exploit English a-layer or t-layer for your project (e.g. Extra features/factors for Moses)
- Try your machine learning (CRF, SVM, NN,...) instead of our MaxEnt dictionary
- Suggest a better algorithm for the transfer (treelet → treelet)
- Use SMT for the deep-syntactic transfer (TectoMoses: linearize trees & project dependencies)

# Thank you

