

# Non-projective Dependency-based Pre-Reordering with Recurrent

## Neural Network for Machine Translation

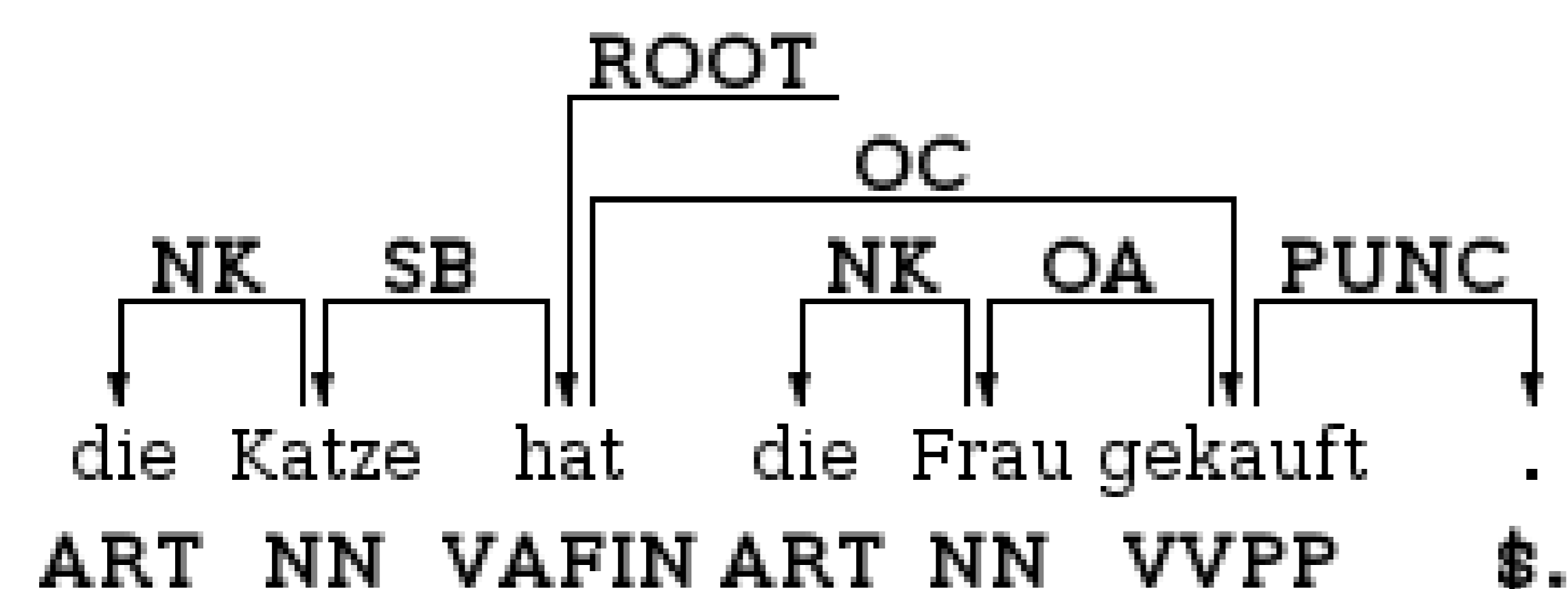
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Reorder sentences in a target-like word order

die Katze hat die Frau gekauft .  
die Frau hat gekauft die Katze .  
the woman has bought the cat .

Based (but not constrained) on  
the source dependency parse



Using recurrent neural network models:

*Base RNN-RM*

*Base GRU-RM*

*Fragment RNN-RM*

### Results

- Experiments on *de-en* Europarl, news2009, news2013
- All models improve over baseline (Moses)
- Base GRU-RM and Fragment GRU-RM equal or improve over Collins et al. (2005) rules
- Translation BLEU (Europarl):

Moses:	33.00		Base RNN-RM:	33.41	+0.41
Collins:	33.52	+0.52	<b>Base GRU-RM:</b>	<b>34.15</b>	<b>+1.15</b>
<b>"Oracle":</b>	<b>41.80</b>	<b>+8.80</b>	Fragment RNN-RM:	33.54	+0.54

Reordering transition system

State:

$(curNode, emittedNodes, lastAction)$

Actions:

$EMIT, UP, DOWN_{childNode}, RIGHT$

Fragment:

Sequence of actions between word emissions

### System

- Align parallel training corpus with giza++  
Symmetrize alignment with *grow-diag-final-and* heuristic
- Generate heuristic reference reordering of source training corpus  
Al-Onaizan and Papineni (2006) heuristic
- Parse source side with DeSR (Attardi and Ciaramita 2007)  
Parse tree may be non-projective
- Train one of these neural network models:  
*Base RNN-RM* or *Base GRU-RM*  
Compute permutation as sequence of words  
*Fragment RNN-RM* (two-level hierarchical RNN)  
Compute permutation as sequence of words and sequence transition actions between each word.  
all these model support *non-tree-local* reordering
- Reorder source training corpus with the neural network model  
Beam search
- Train *Moses* on the reordered parallel corpus