
Advanced Natural Language Processing and Information Retrieval

LAB3: Kernel Methods for Reranking

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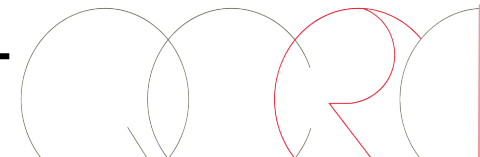
Preference Reranking slides at:

<http://disi.unitn.it/moschitti/teaching.html>

LAB3: Ranking with Tree Kernels

Download:

LAB3.zip



The Ranking SVM

[Herbrich et al. 1999, 2000; Joachims et al. 2002]

- The aim is to classify instance pairs as correctly ranked or incorrectly ranked
 - This turns an ordinal regression problem back into a binary classification problem

- We want a ranking function f such that

$$\mathbf{x}_i > \mathbf{x}_j \text{ iff } f(\mathbf{x}_i) > f(\mathbf{x}_j)$$

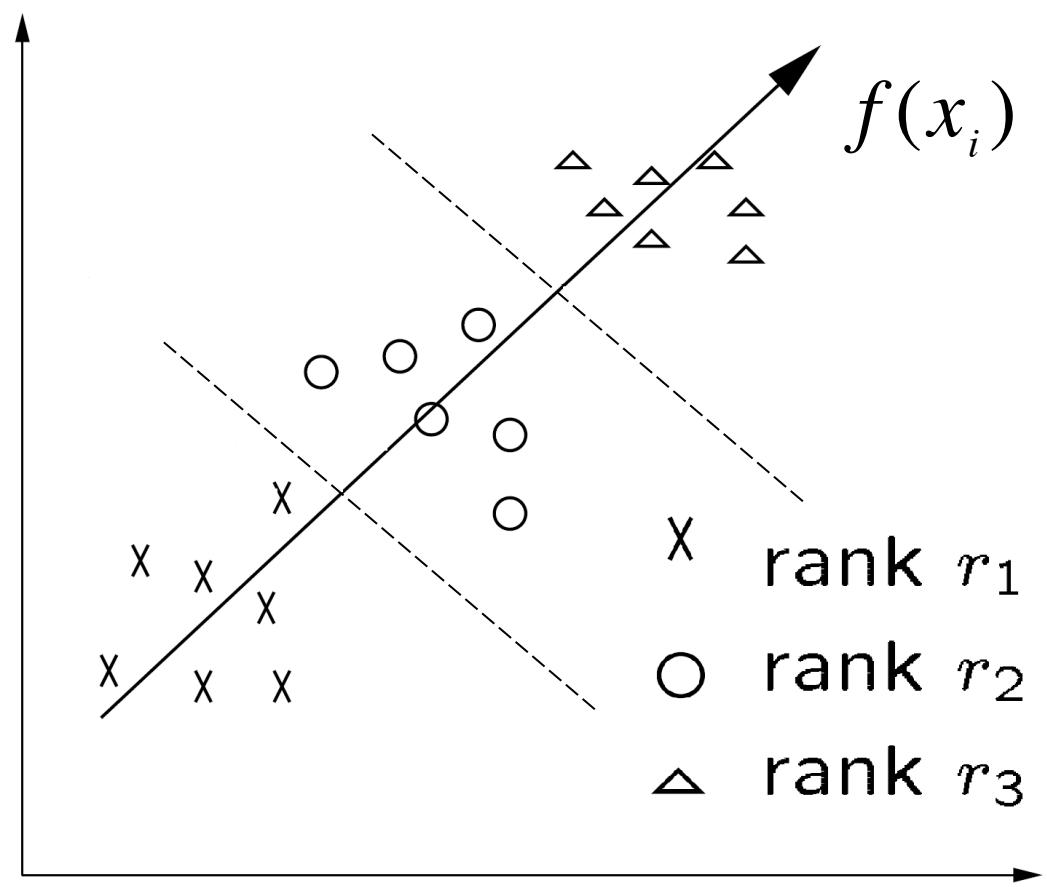
- ... or at least one that tries to do this with minimal error
- Suppose that f is a linear function

$$f(\mathbf{x}_i) = \mathbf{w} \cdot \mathbf{x}_i$$



The Ranking SVM

- Ranking Model: $f(\mathbf{x}_i)$



The Ranking SVM

- Then (combining the two equations on the last slide):

$$\mathbf{x}_i > \mathbf{x}_j \text{ iff } \mathbf{w} \cdot \mathbf{x}_i - \mathbf{w} \cdot \mathbf{x}_j > 0$$

$$\mathbf{x}_i > \mathbf{x}_j \text{ iff } \mathbf{w} \cdot (\mathbf{x}_i - \mathbf{x}_j) > 0$$

- Let us then create a new instance space from such pairs:

$$\mathbf{z}_k = \mathbf{x}_i - \mathbf{x}_k$$

$$y_k = +1, -1 \text{ as } \mathbf{x}_i \geq, < \mathbf{x}_k$$



Support Vector Ranking

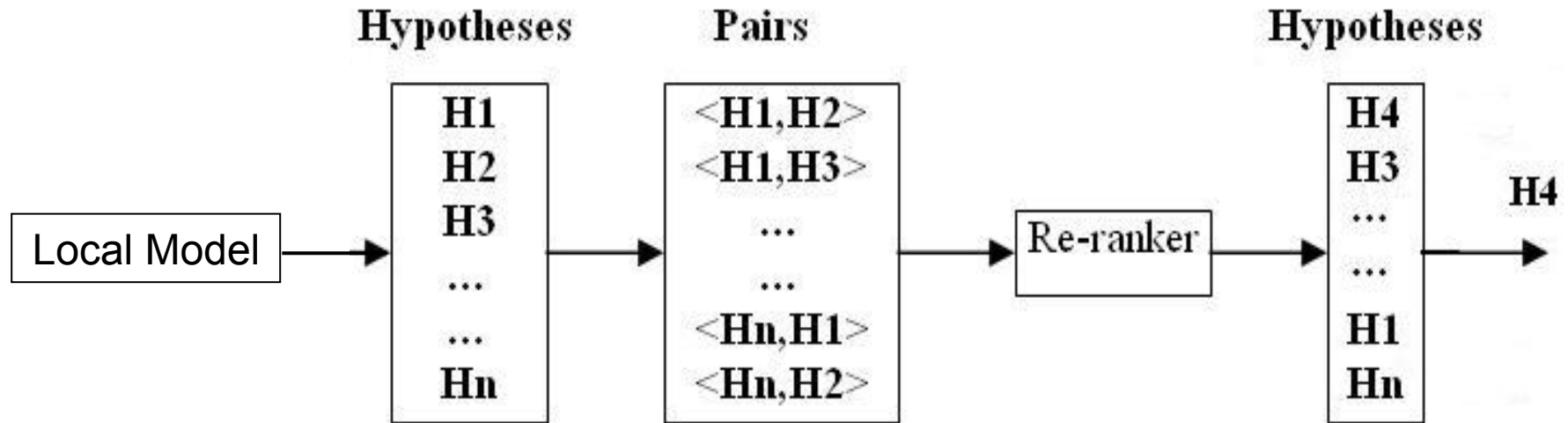
$$\begin{cases} \min & \frac{1}{2} \|\vec{w}\|^2 + C \sum_{i=1}^m \xi_i^2 \\ & y_k (\vec{w} \cdot (\vec{x}_i - \vec{x}_j) + b) \geq 1 - \xi_k, \quad \forall i, j = 1, \dots, m \\ & \xi_k \geq 0, \quad k = 1, \dots, m^2 \end{cases}$$

$y_k = 1$ if $\text{rank}(\vec{x}_i) > \text{rank}(\vec{x}_j)$, -1 otherwise, where $k = i \times m + j$

- Given two examples we build one example (x_i, x_j)



Framework of Preference Reranking

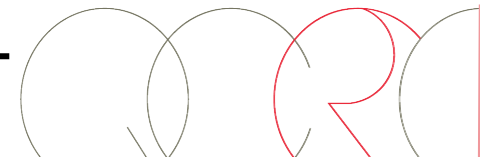


- The local model is a system providing the initial rank
- Preference reranking is superior to ranking with an instance classifier since it compares pairs of hypotheses



More formally

- Build a set of hypotheses: Q and A pairs
- These are used to build pairs of pairs $\langle (S^i, H^i), (S^j, H^j) \rangle$
 - positive instances if H^i is correct and H^j is not correct
- A binary classifier decides if H^i is more probable than H^j
- Each candidate annotation H^i is described by a structural representation
- This way kernels can exploit all dependencies between features and labels



Preference Reranking Kernel

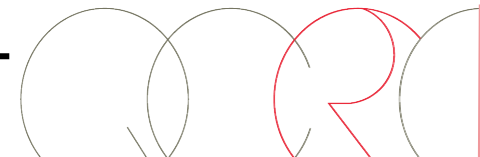
$H_1 > H_2$ and $H_3 > H_4$ then consider training vectors:

$\vec{Z}_1 = \phi(H_1) - \phi(H_2)$ and $\vec{Z}_2 = \phi(H_3) - \phi(H_4) \Rightarrow$ the dot product is:

$$\begin{aligned}\vec{Z}_1 \cdot \vec{Z}_2 &= (\phi(H_1) - \phi(H_2)) \cdot (\phi(H_3) - \phi(H_4)) = \\ &\phi(H_1) \cdot \phi(H_3) - \phi(H_1) \cdot \phi(H_4) - \phi(H_2) \cdot \phi(H_3) + \phi(H_2) \cdot \phi(H_4) \\ &= K(H_1, H_3) - K(H_1, H_4) - K(H_2, H_3) + K(H_2, H_4)\end{aligned}$$

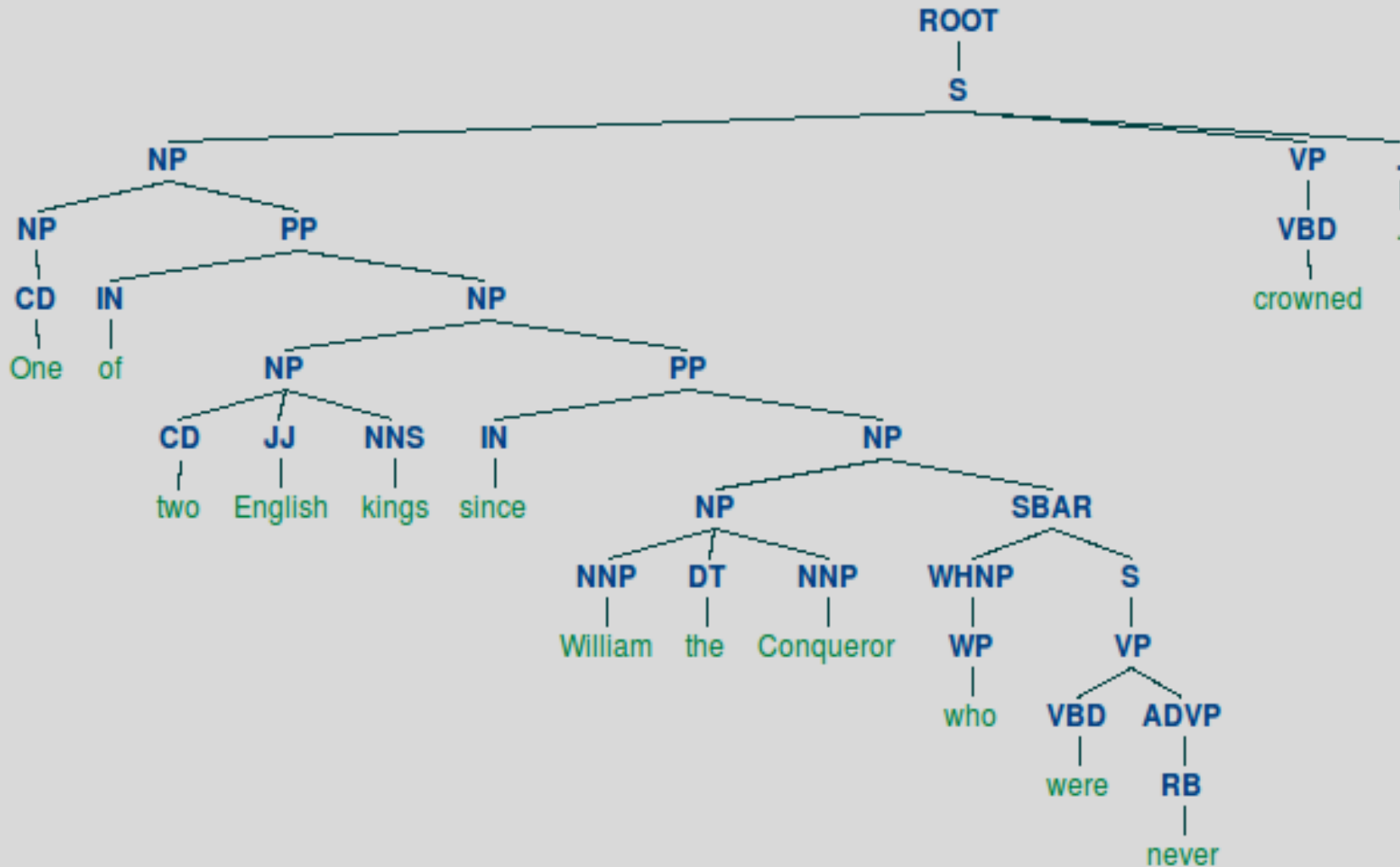
Let $H_i = \langle q_i, a_i \rangle$, $H_j = \langle q_j, a_j \rangle$

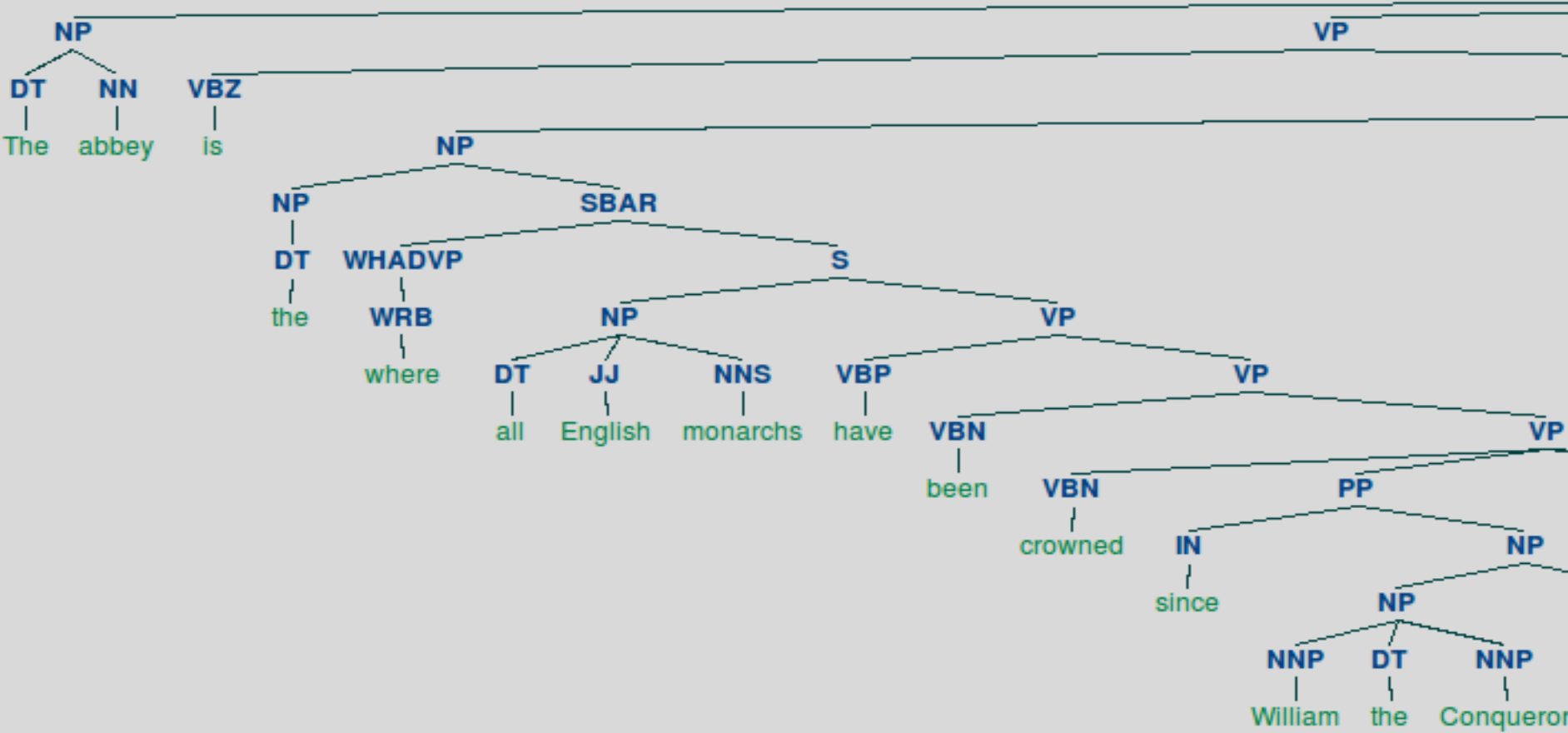
$$K(H_i, H_j) = PTK(q_i, q_j) + PTK(a_i, a_j)$$



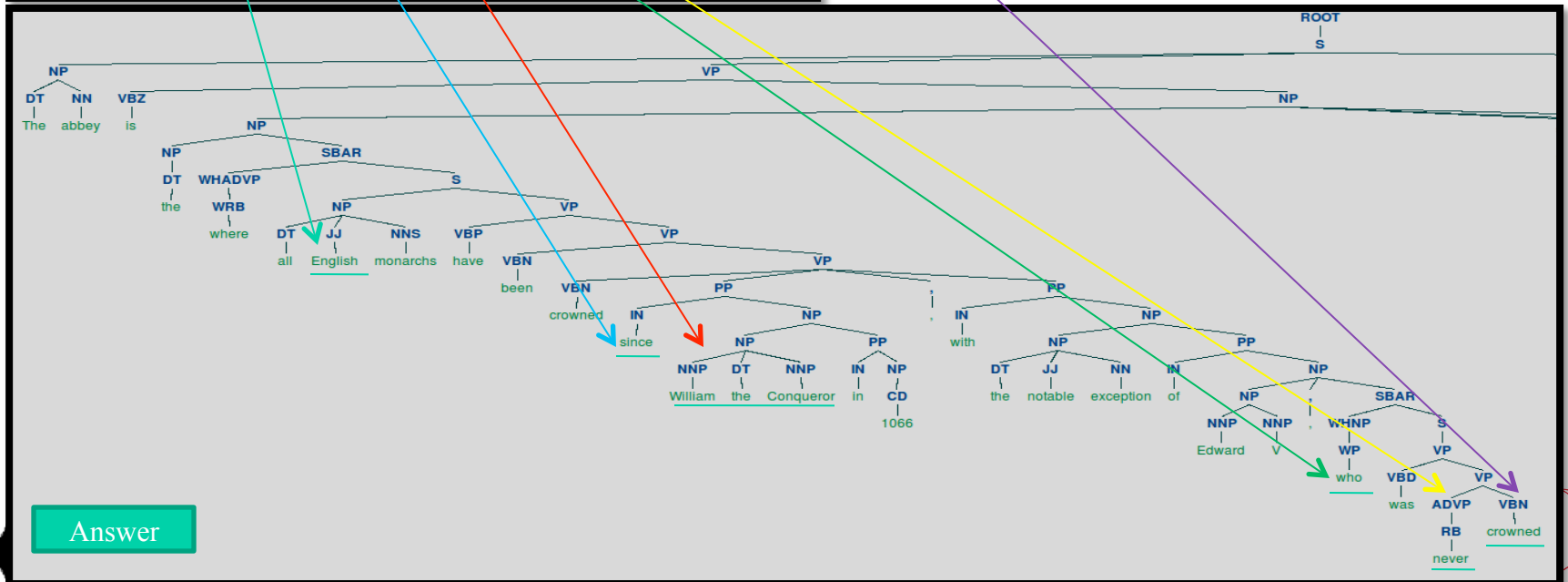
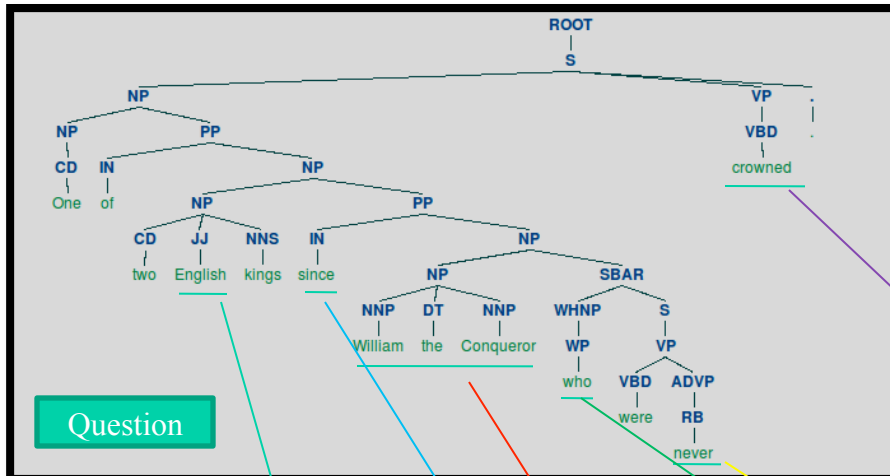


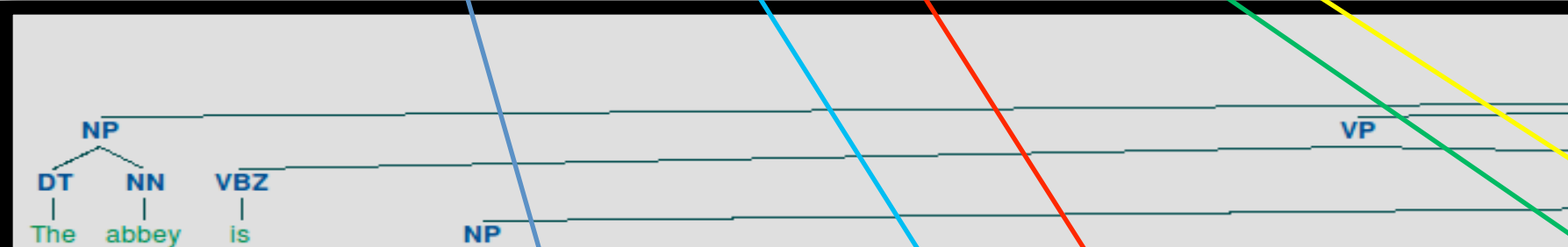
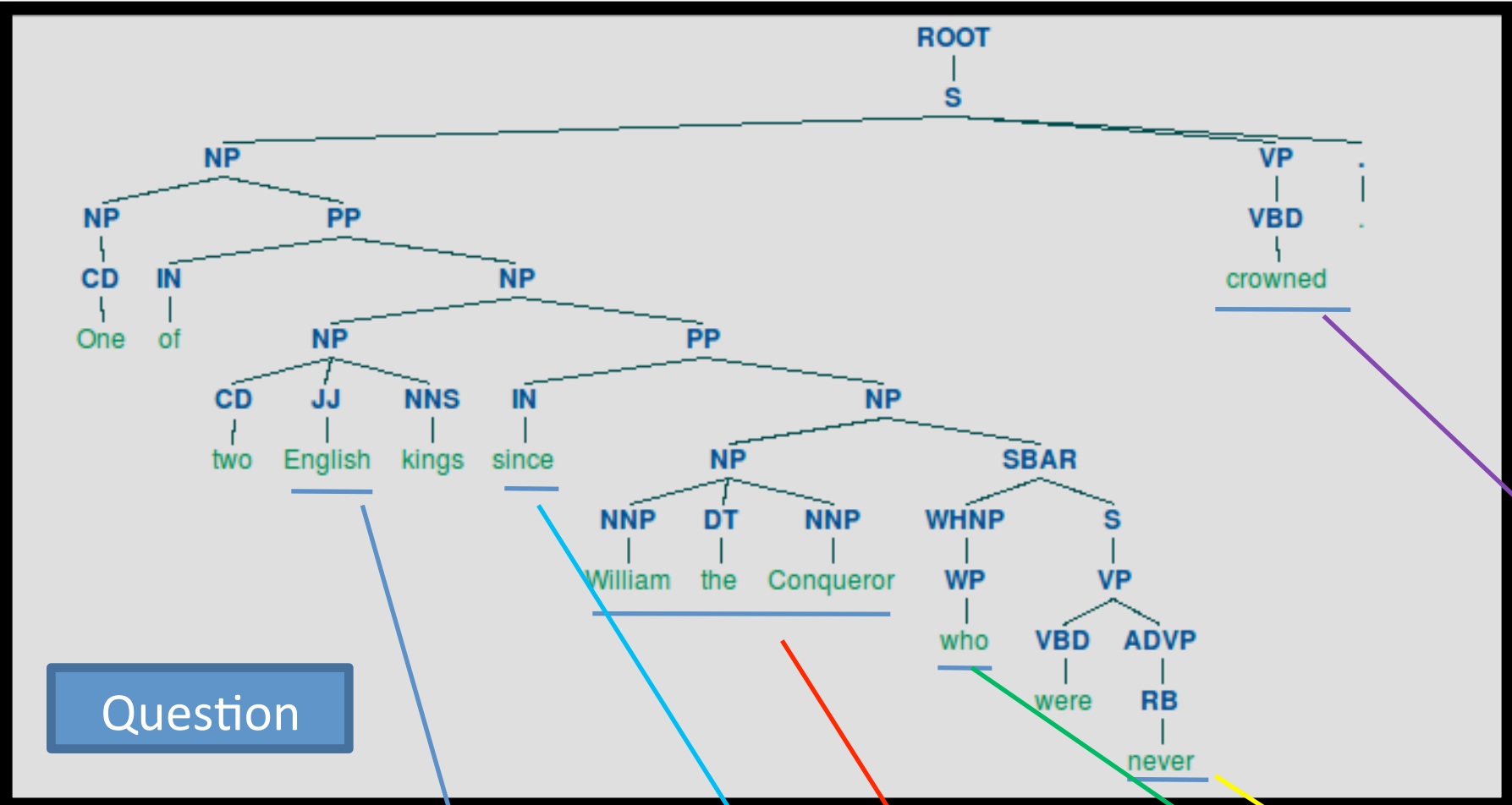
An example of Jeopardy! Question



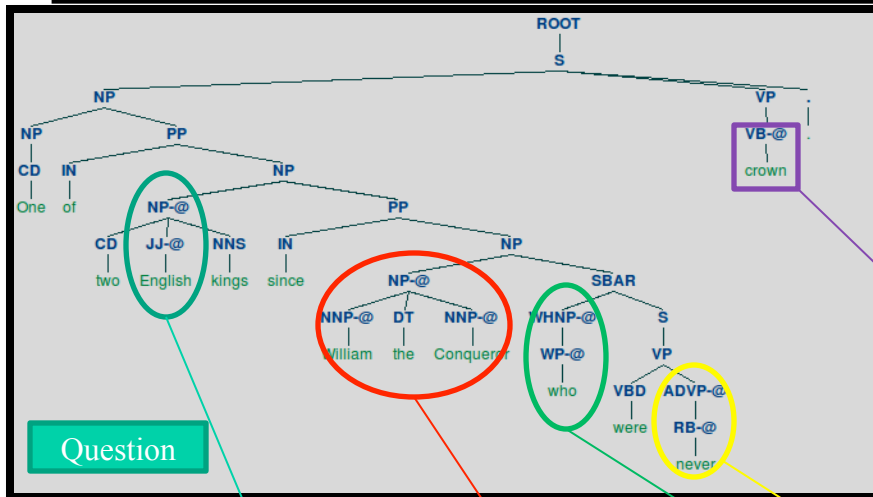


Adding Relational Links



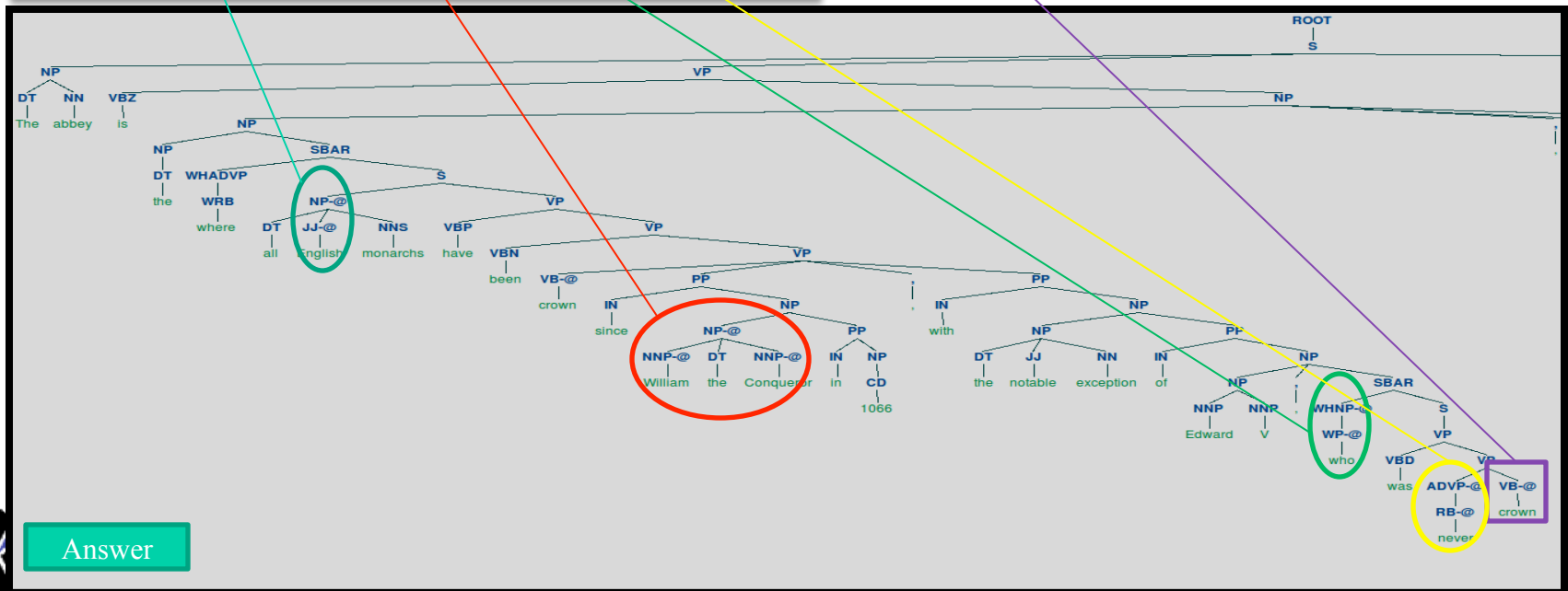


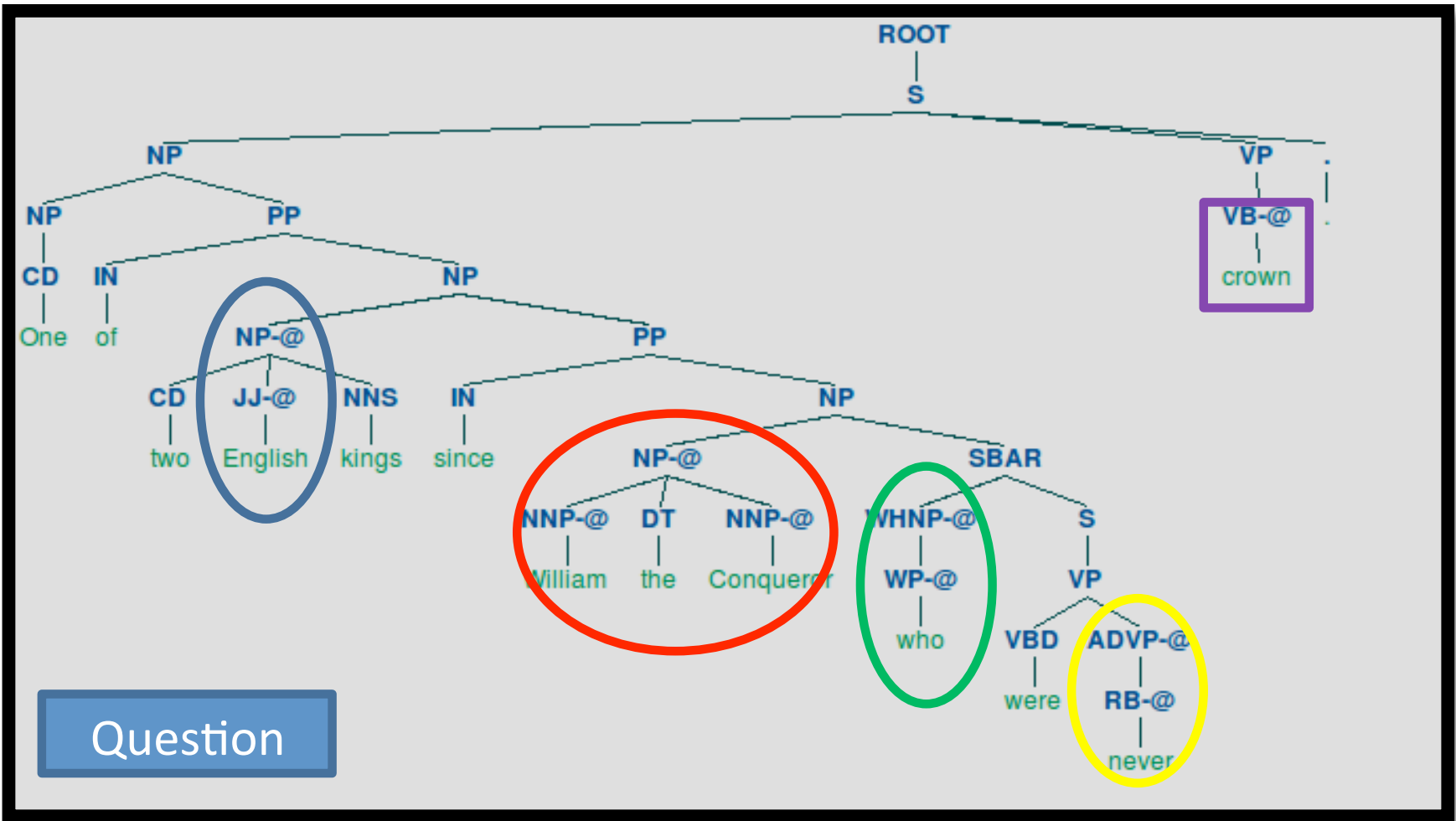
Links can be encoded marking tree nodes



Methodology:

- 1-Applying lemmatization or stemming to the leaves
- 2-Mark (with @ symbol) pre-terminal nodes and higher level nodes if the subtrees are shared in Q and A
- 3-Ignore stop words in the matching procedure



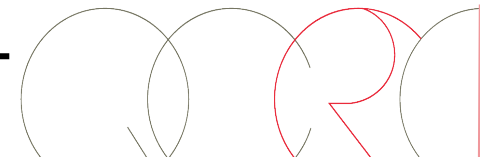


Question



Representation Issues

- Very large sentences
- The Jeopardy! cues can be constituted by more than one sentence
- The answer is typically composed by several sentences
- Too large structures cause inaccuracies in the kernel similarity and the learning algorithm loses some of its power



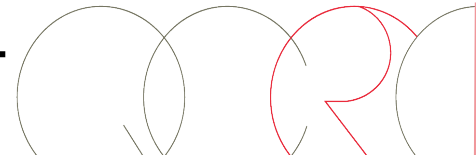
Running example from Answerbag

Question: Is movie theater popcorn vegan?

Answer:

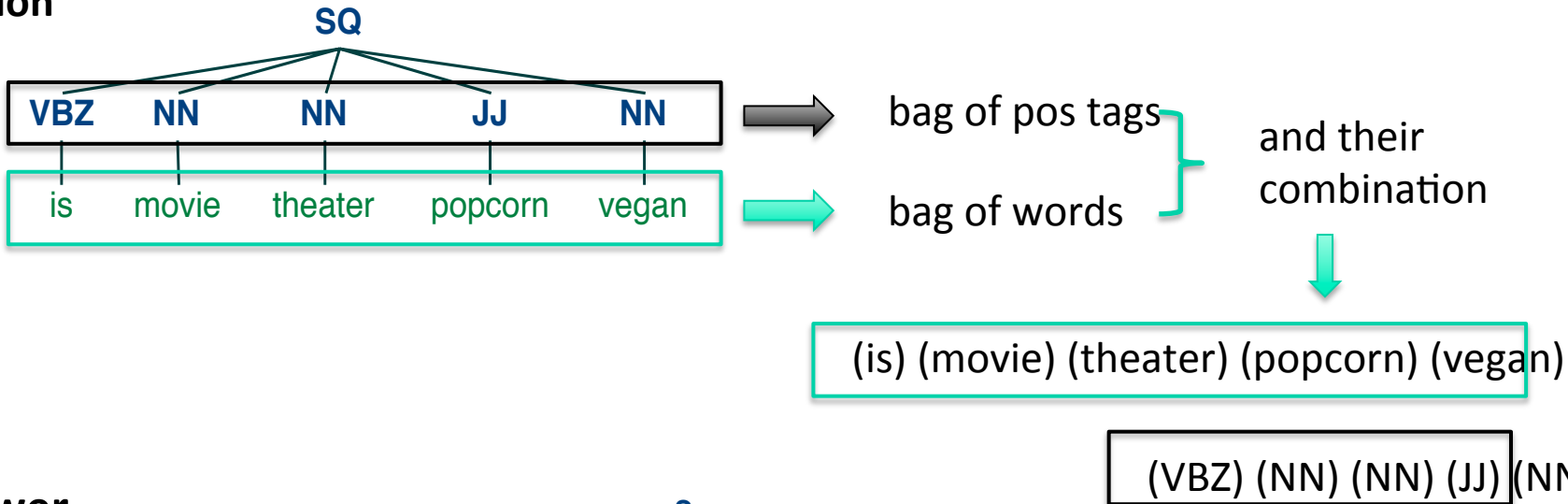
(01) Any movie theater popcorn that includes butter -- and therefore dairy products -- is not vegan.

(02) However, the popcorn kernels alone can be considered vegan if popped using canola, coconut or other plant oils which some theaters offer as an alternative to standard popcorn.

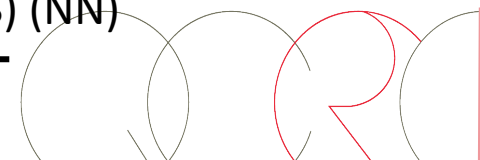
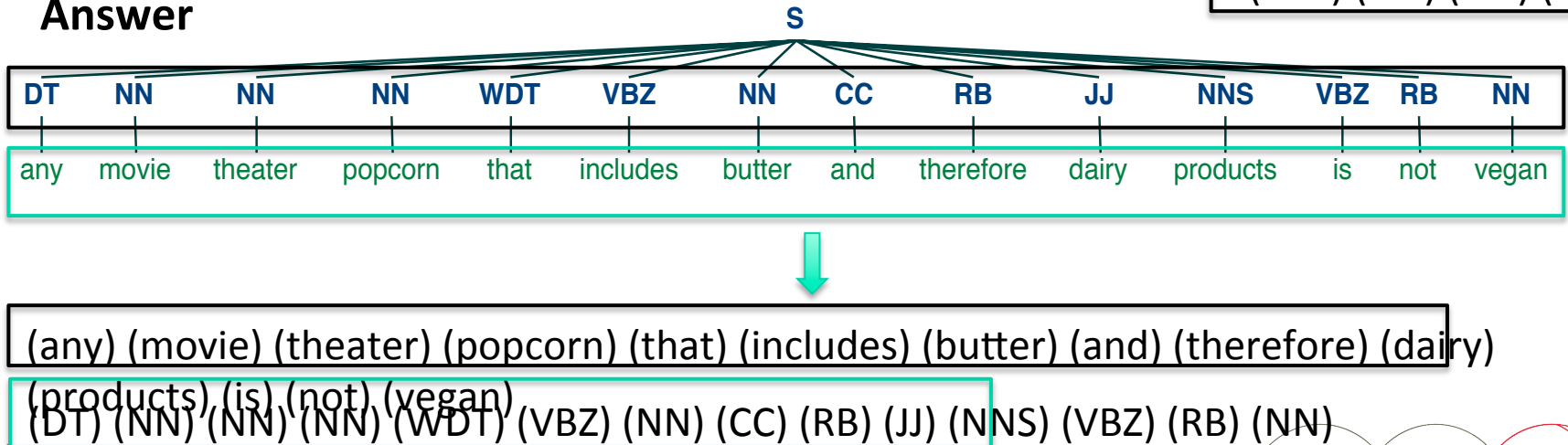


Shallow models for Reranking: [Severyn & Moschitti, SIGIR 2012]

Question

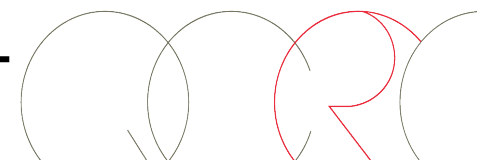
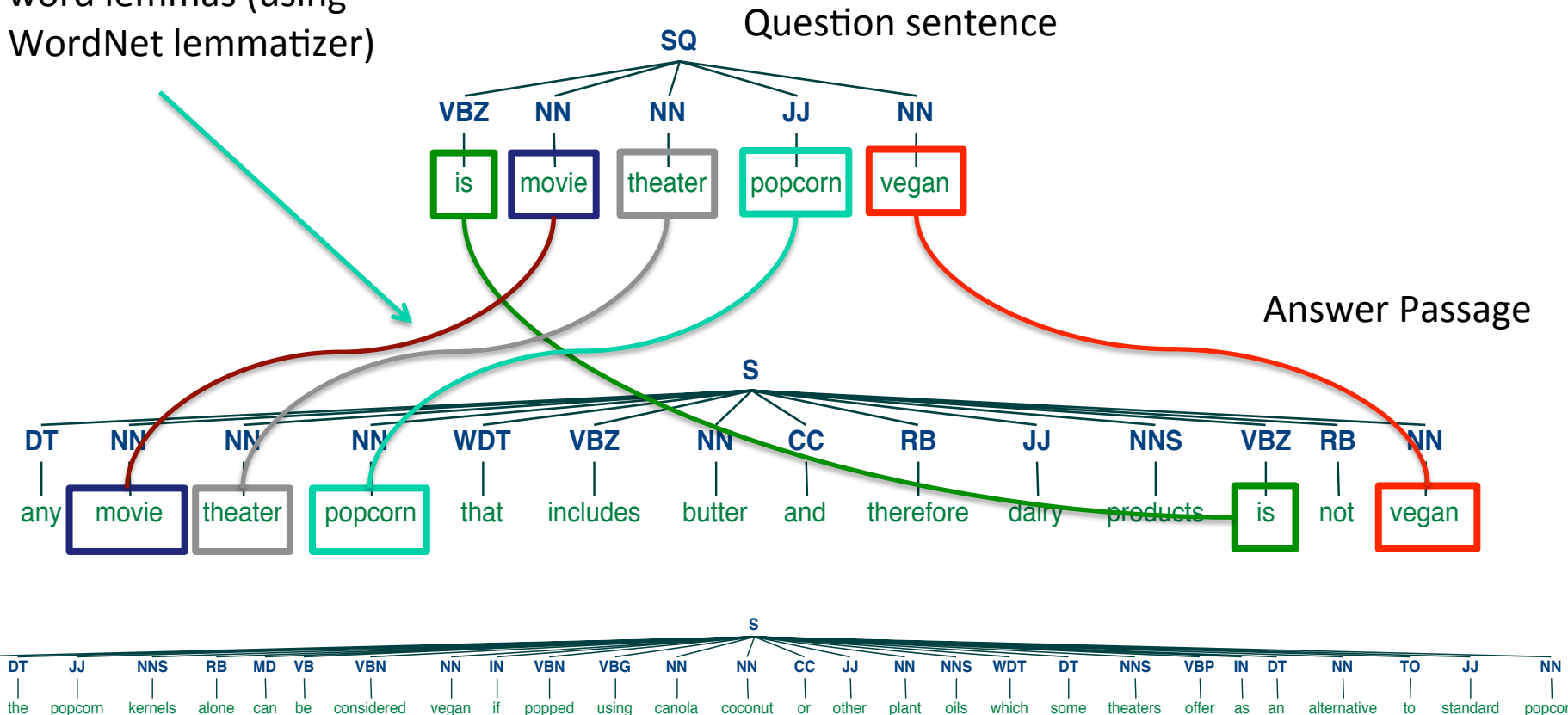


Answer



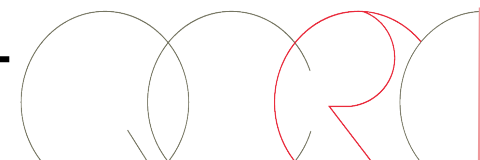
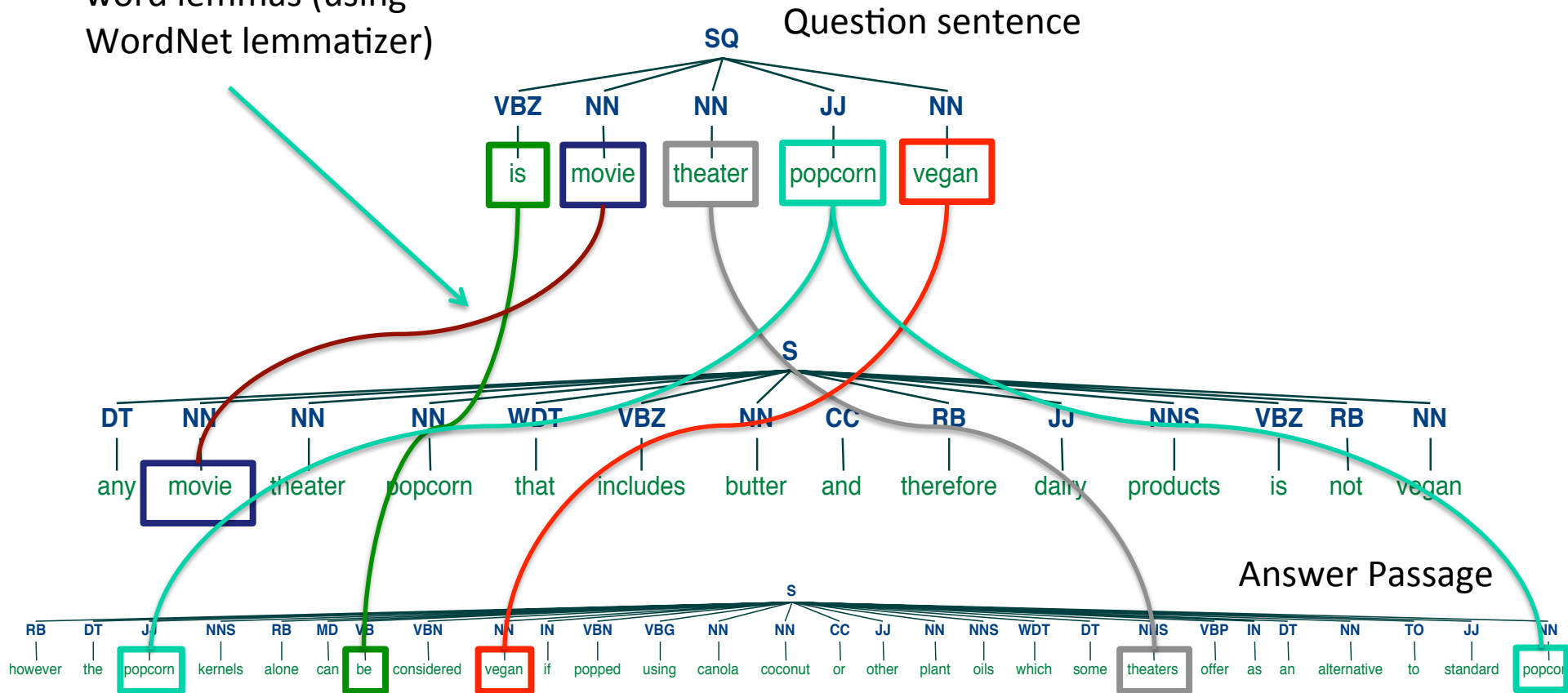
Linking question with the answer 01

Lexical matching is on word lemmas (using WordNet lemmatizer)

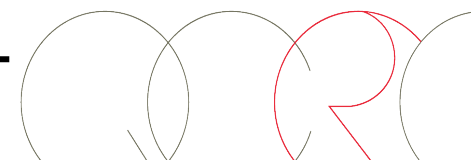
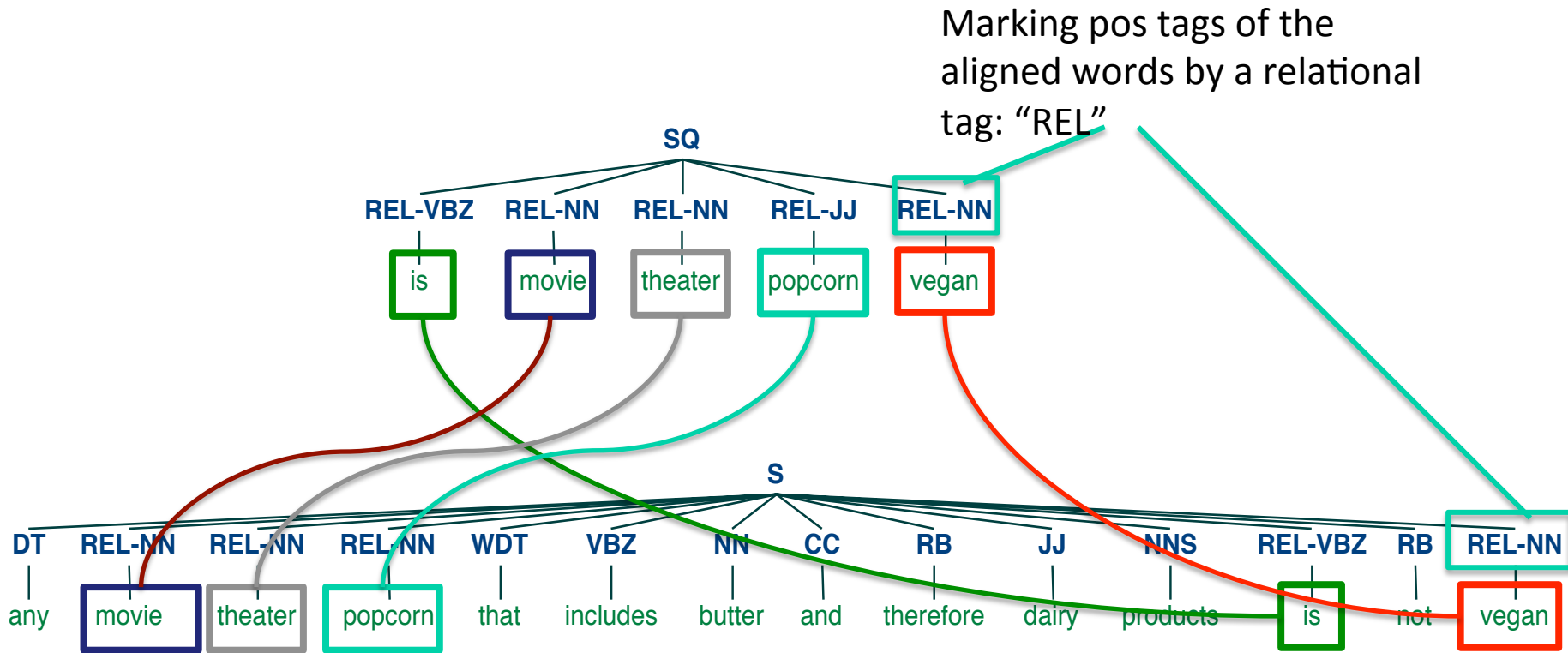


Linking question with the answer 02

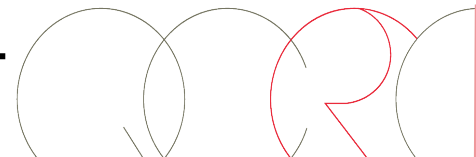
Lexical matching is on word lemmas (using WordNet lemmatizer)



Linking question and its answer passages using a relational tag

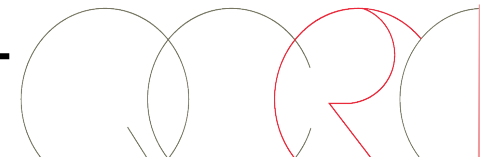


Let's start the LAB3: Ranking with Tree Kernels



SVM-light-TK and Ranking Data

- SVM-light-TK encodes STK, PTK and combination kernels in SVM-light [Joachims, 1999]
- <http://disi.unitn.it/moschitti/teaching.html>
 - Academic Year: 2015-2016
 - Download: LAB3.zip

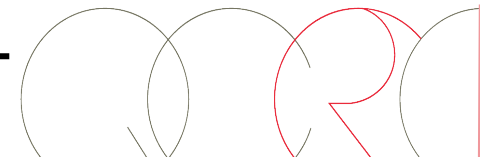


Compile the package

- Go under SVM directory
 - `cd SVM-Light-1.5-rer/`
- Type make to build the code
 - `make`
- Go back to the previous directory
 - `cd ..`

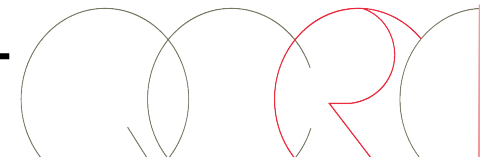
Generating examples for reranking

- **questions.5k.txt**, contains a set of questions
 - each line contains a unique id and the question itself separated by a tab, i.e., "\t"
- **answers.txt** -- contains a set of answers
 - each line contains a unique id and the answer passage itself separated by a tab, i.e. "\t"



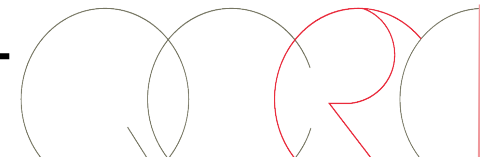
Training and testing files

- **results.*.15k**, a rank list for 1,000 questions (contained in questions.5k.txt)
 - (1) the id of the question
 - (2) the id of the passage, and
 - (3) its score from the search engine
- **results.train.15k, results.test.15k**
 - 1000 questions
 - 15 retrieved passages for each question
 - (**BOX** (the) (cell) (phone) (used) (tony) (stark) (the) (movie) (iron) (man) (was) (vx9400) (slider) (phone) (which) (was) (just) (one) (the) (mobile) (phones) (used) (the) (movie.))



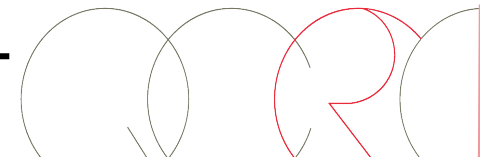
Building the reranker files

- Generate training examples for reranking
 - `python generate_reranking_pairs.py questions.5k.txt answers.txt results.train.15k`
 - `python2.7`
- Generate testing examples for reranking
 - `python generate_reranking_pairs.py -m test questions.5k.txt answers.txt results.test.15k`



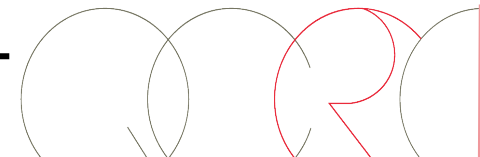
Retrieval results for a question

- 2500744 What kind of cell phone was used in the movie "Iron Man"?
- 2500744 The cell phone used by Tony Stark in the movie "Iron Man" was a LG VX9400 slider phone, which was just one of the LG mobile phones used in the movie.
- 2259459 The average person cannot trace a prepaid cell phone; however, the federal government and police force do have this capability. While they cannot determine a person's exact location, they can find what cell phone towers are being used and use this information to trace the phone.



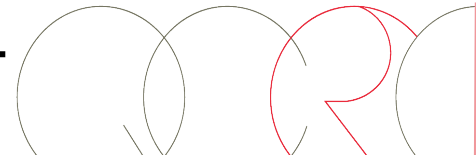
Generated learning files: svm.train

- +1 |BT| (BOX (the) (cell) (phone) (used) (tony) (stark) (the) (movie) (iron) (man) (was) (vx9400) (slider) (phone) (which) (was) (just) (one) (the) (mobile) (phones) (used) (the) (movie.))
|BT| (BOX (the) (average) (person) (cannot) (trace) (prepaid) (cell) (phone) (however) (the) (federal) (government) (and) (police) (force) (have) (this) (capability.) (while) (they) (cannot) (determine) (person) (exact) (location) (they) (can) (find) (what) (cell) (phone) (towers) (are) (being) (used) (and) (use) (this) (information) (trace) (the) (phone.))
|ET| 1:2.28489184 |BV| 1:0.65760440 |EV|



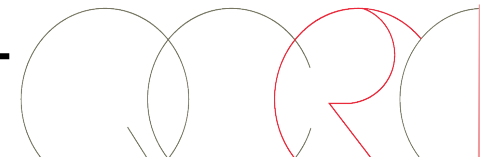
Generated test files: svm.test

- +1 |BT| (BOX (the) (cell) (phone) (used) (tony) (stark) (the) (movie) (iron) (man) (was) (vx9400) (slider) (phone) (which) (was) (just) (one) (the) (mobile) (phones) (used) (the) (movie.))
|BT| EMPTY |ET| 1:2.28489184 |BV| EMPTY |EV|



Training and testing the reranker

- `./SVM-Light-1.5-rer/svm_learn -t 5 -F 2 -C + -W R -V R -S 0 -N 1 svm.train model`
- SVM-TK options:
 - `-F 2`, tree kernel using words (bag of word)
 - `-C +`, combine contribution from trees and vectors
 - `-W R`, apply reranking on trees
 - `-V R`, apply reranking on vectors
 - `-S 0`, linear kernel on the feature vector
 - `-N 1`, normalization on tree, not on the feature vector



Testing and evaluating the reranker

- `./SVM-Light-1.5-rer/svm_classify svm.test model
pred`
- `python evReranker.py svm.test.res pred`