

Let us place the languages of kernels in a hierarchy:

Relational

kernels

(klog) : Language

First

order logic

Generalised to Rational kernels

Graph kernels

Generalisation is Computation

Kernels induced by grammars

Relational subsequence kernels

String kernels

Tree kernels

Kernels obtained by virtue of grammars (content free)

Generalization to

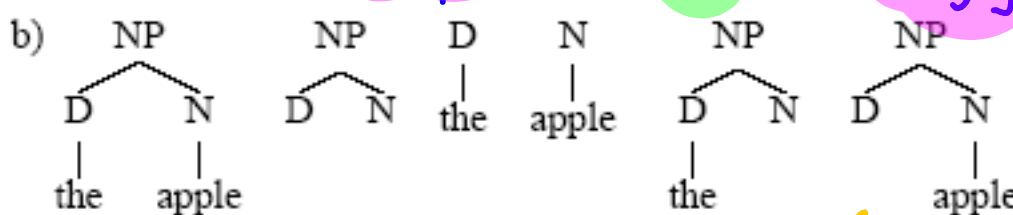
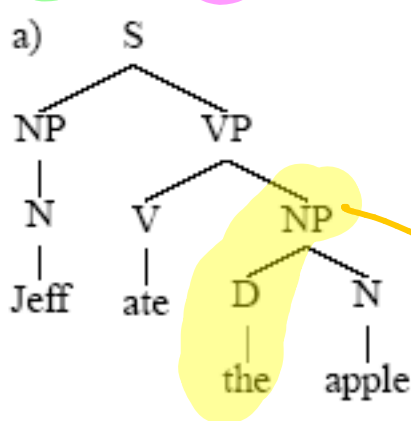
convolution kernels

$S \rightarrow NP VP$

$D \rightarrow the$

$NP \rightarrow N$
 $N \rightarrow apple$

$NP \rightarrow DN$
 $N \rightarrow Jeff$



Not a valid subtree

Each of these is a subtree indexed by some i_1, i_2, \dots, i_n

Figure 2: a) An example tree. b) The sub-trees of the NP covering *the apple*. The tree in (a) contains all of these sub-trees, and many others. We define a sub-tree to be any sub-graph which includes more than one node, with the restriction that entire (not partial) rule productions must be included. For example, the fragment [NP [D the]] is excluded because it contains only part of the production $NP \rightarrow DN$.

$V \rightarrow ate$

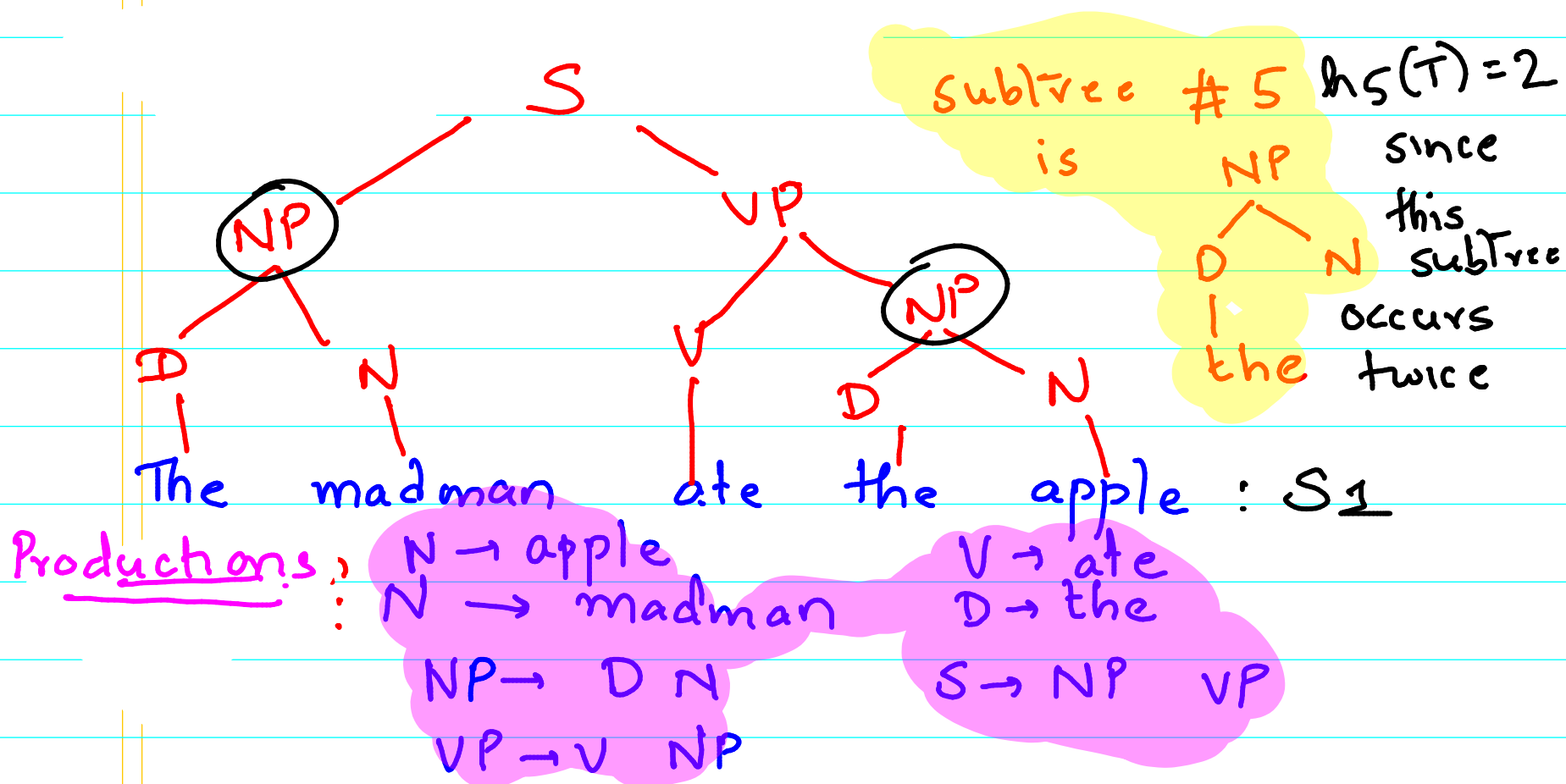
Terminals in pink
Nonterminals in pink.

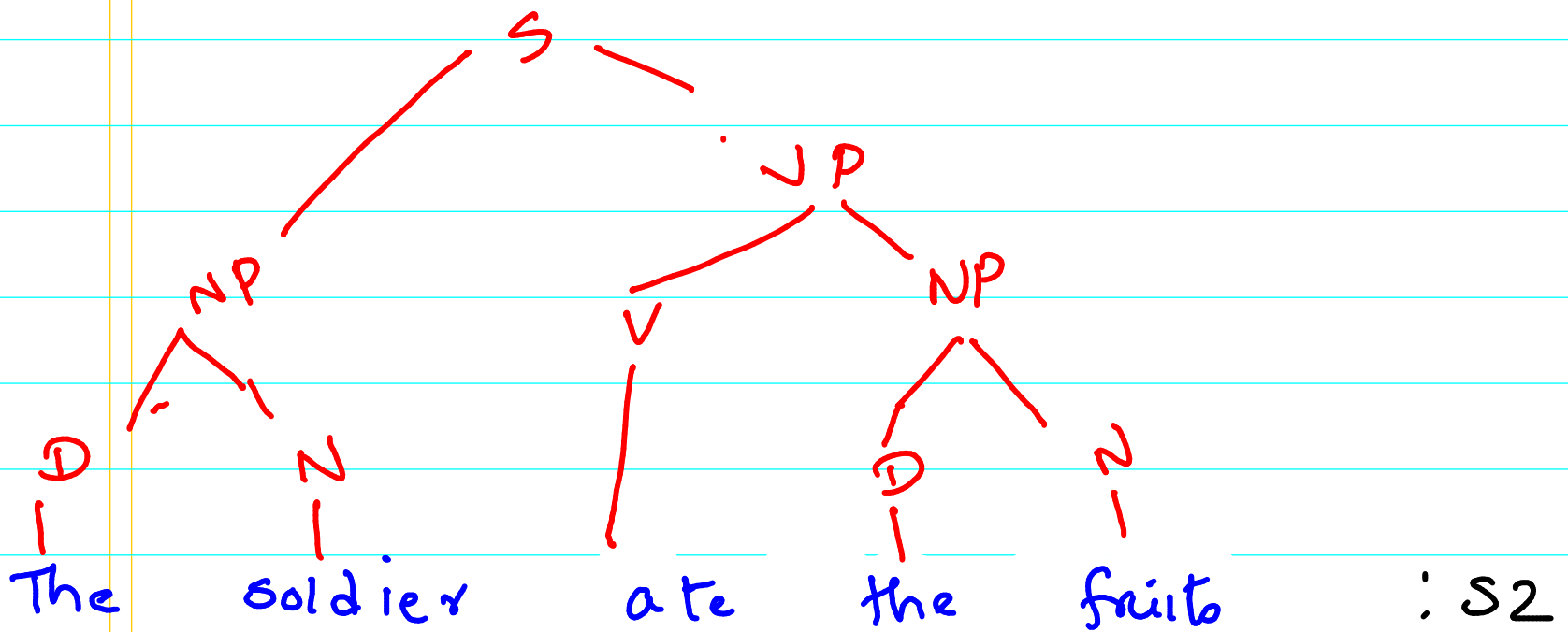
The parse tree (CFG) kernel

computes # of common subtrees between 2 trees

Conceptually we begin by enumerating all tree fragments that occur in the training data $1, \dots, n$. Note that this is done only implicitly. Each tree is represented by an n dimensional vector where the i 'th component counts the number of occurrences of the i 'th tree fragment. Let us define the function $h_i(T)$ to be the number of occurrences of the i 'th tree fragment in tree T , so that T is now represented as $\mathbf{h}(T) = (h_1(T), h_2(T), \dots, h_n(T))$.

$$\text{We want } k(T_1, T_2) = \sum_i h_i(T_1) h_i(T_2)$$



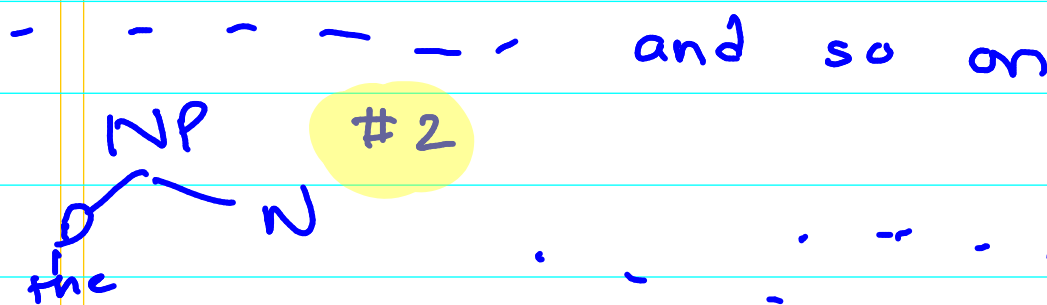
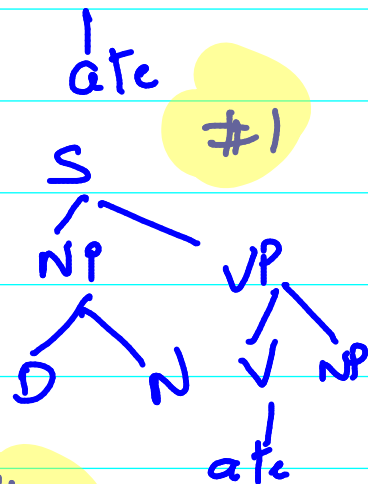
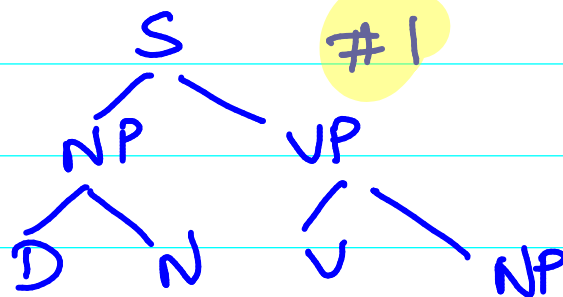
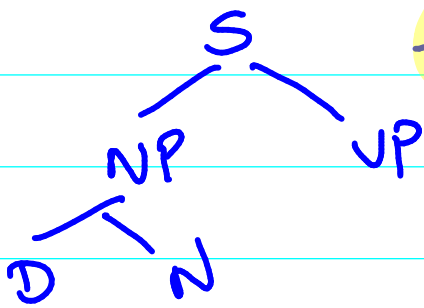
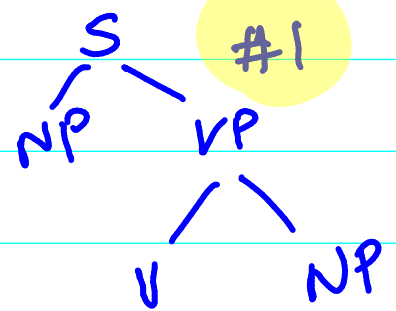
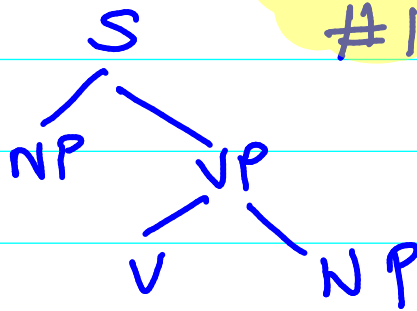
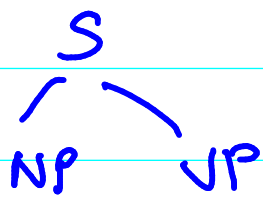


(we will assume that all productions of the form
 $N \rightarrow$ "a noun"
 $V \rightarrow$ "a verb" etc are implicitly present in some dictionaries.)

$K(S_1, S_2) = \sum_{\substack{T_1, T_2 \\ \text{valid}}} K(T_1, T_2) = \sum_{\text{subtrees } i} h_i(T_1) h_i(T_2)$

for the time being assume that S_1 & S_2 each have a single parse tree = 24 (?) verify

In practice you normalize $K(T_1, T_2) / (K(T_1, T_1) * K(T_2, T_2))^{1/2}$



To compute $K(\tau_1, \tau_2)$ efficiently

This decomposition is characteristic of convolution

(i) We realize that $h_i(\tau)$ is itself a composite function of a more basic indicator function

$$I_i(n, \tau) = 1 \text{ iff subtree } i \text{ is rooted at node } n \text{ in } \tau$$

$$= 0 \text{ o/w}$$

$$h_i(T_1) = \sum_{n_1 \in T_1} I_i(n_1, T_1)$$

$$h_i(T_2) = \sum_{n_2 \in T_2} I_i(n_2, T_2)$$

$$\textcircled{2} \quad K(T_1, T_2) = \sum_i h_i(T_1) h_i(T_2)$$

$$= \sum_i \sum_{n_1 \in T_1} \sum_{n_2 \in T_2} I_i(n_1, T_1) I_i(n_2, T_2)$$

$$= \sum_{n_1 \in T_1} \sum_{n_2 \in T_2} \sum_i I_i(n_1, T_1) I_i(n_2, T_2)$$

bringing summation over
subtrees inside since I would
try & get rid of it.

call $\sum_i I_i(n_1, T_1) I_i(n_2, T_2)$
as $H(n_1, n_2)$
& recurse on it

$H(n_1, n_2) = 0$ if the productions at
 n_1 & n_2 are different

(eg: $n_1: NP \rightarrow D N$

$n_2: VP \rightarrow V NP$

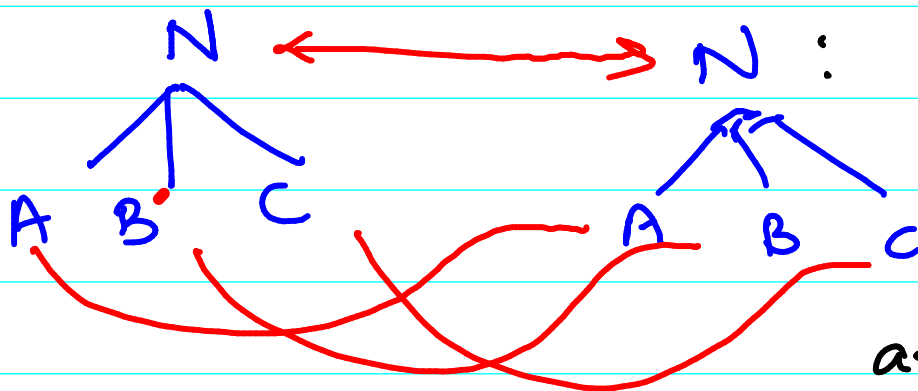
or even if $n_2: S \rightarrow NP VP$)

$H(n_1, n_2) = 1$ if $n_1: N \rightarrow t$
 $n_2: N \rightarrow t$

(where $N = \text{nonterminal}$ &
 $t = \text{terminal}$)

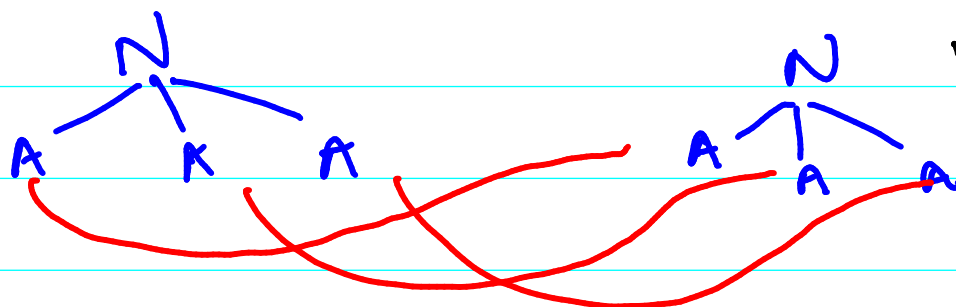
$$H(n_1, n_2) = \prod_{j=1}^{\# \text{children}(n)} (1 + H(\text{child}(n_1, j), \text{child}(n_2, j)))$$

order matters



The product comes because

I can consider expansions at one or more of A, B, C etc at that level



Since

$I_i(n_1, T)$

$\neq I_i(n_2, T)$

was considering if

i^{th} subtree was rooted at both n_1 & n_2 , the order of matches matters since the production from N

MUST be honored

Order imp because in $N \rightarrow ABC$ ABC is a sequence

Pls read

Convolution Kernels for Natural Language