50.040 Natural Language Processing (Summer 2020) Homework

Due

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Students with whom you have discussed (if any):

```
In [1]:
```

```
from google.colab import drive
drive.mount('/content/gdrive')
```

Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6bn6 qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect_uri=urn%3aietf%3awg%3aoauth%3a2.0%b&response_type=code&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonlyttps%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonlyttps%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly

Enter your authorization code:
......
Mounted at /content/gdrive

In [2]:

```
% pwd
% ls
% cd /content/gdrive/'My Drive'/'NLP'/'HW2'
% ls
```

gdrive/ sample_data/

/content/gdrive/My Drive/NLP/HW2 homework2.ipynb hw2.pdf imgs/ MACOSX/

In [3]:

```
import copy
from collections import Counter
from nltk.tree import Tree
from nltk import Nonterminal
from nltk.corpus import LazyCorpusLoader, BracketParseCorpusReader
from collections import defaultdict
import time
```

In [3]:

In [4]:

```
st = time.time()
```

In [5]:

```
import nltk
nltk.download('treebank')
```

[nltk_data] Downloading package treebank to /root/nltk_data...
[nltk_data] Unzipping corpora/treebank.zip.

```
Out[5]:
True
In [6]:
def set leave lower(tree string):
    if isinstance(tree string, Tree):
        tree = tree_string
    else:
    tree = Tree.fromstring(tree_string)
for idx, _ in enumerate(tree.leaves()):
        tree location = tree.leaf treeposition(idx)
        non terminal = tree[tree location[:-1]]
        non_terminal[0] = non_terminal[0].lower()
    return tree
def get_train_test_data():
    Load training and test set from nltk corpora
    train num = 3900
    test_index = range(10)
    treebank = LazyCorpusLoader('treebank/combined', BracketParseCorpusReader, r'wsj .*\.mrg')
    cnf_train = treebank.parsed_sents()[:train_num]
    cnf_test = [treebank.parsed_sents()[i+train_num] for i in test_index]
    #Convert to Chomsky norm form, remove auxiliary labels
    cnf_train = [convert2cnf(t) for t in cnf_train]
    cnf test = [convert2cnf(t) for t in cnf_test]
    return cnf train, cnf test
def convert2cnf(original_tree):
    Chomsky norm form
    tree = copy.deepcopy(original tree)
    #Remove cases like NP->DT, VP->NP
    tree.collapse unary(collapsePOS=True, collapseRoot=True)
    #Convert to Chomsky
    tree.chomsky_normal_form()
    tree = set_leave_lower(tree)
    return tree
In [7]:
### GET TRAIN/TEST DATA
cnf train, cnf test = get train test data()
In [8]:
cnf train[0].pprint()
  (NP-SBJ
    (NP (NNP pierre) (NNP vinken))
    (NP-SBJ|<,-ADJP-,>
      (,,)
      (NP-SBJ|<ADJP-,>
        (ADJP (NP (CD 61) (NNS years)) (JJ old))
        (, ,))))
  (S|<VP-.>
    (VP
      (MD will)
      (VP
        (VB join)
        (VP|<NP-PP-CLR-NP-TMP>
          (NP (DT the) (NN board))
           (VP|<PP-CLR-NP-TMP>
            (PP-CLR
               (IN as)
               (NP
```

To better understand PCFG, let's consider the first parse tree in the training data "cnf_train" as an example. Run the code we have provided for you and then writedown the roles of.productions(), .rhs(), .lhs(), .leaves()in the ipynb notebook.

```
In [9]:
rules = cnf train[0].productions()
print(rules, type(rules[0]))
[S -> NP-SBJ S|<VP-.>, NP-SBJ -> NP NP-SBJ|<,-ADJP-,>, NP -> NNP NNP, NNP -> 'pierre', NNP -> 'vin
ken', NP-SBJ|<,-ADJP-,> -> , NP-SBJ|<ADJP-,>, , -> ',', NP-SBJ|<ADJP-,> -> ADJP ,, ADJP -> NP JJ,
NP -> CD NNS, CD -> '61', NNS -> 'years', JJ -> 'old', , -> ',', S|<VP-.> -> VP ., VP -> MD VP, MD
-> 'will', VP -> VB VP|<NP-PP-CLR-NP-TMP>, VB -> 'join', VP|<NP-PP-CLR-NP-TMP> -> NP VP|<PP-CLR-NP
-TMP>, NP -> DT NN, DT -> 'the', NN -> 'board', VP|<PP-CLR-NP-TMP> -> PP-CLR NP-TMP, PP-CLR -> IN
NP, IN -> 'as', NP -> DT NP|<JJ-NN>, DT -> 'a', NP|<JJ-NN> -> JJ NN, JJ -> 'nonexecutive', NN -> '
director', NP-TMP -> NNP CD, NNP -> 'nov.', CD -> '29', . -> '.'] <class
'nltk.grammar.Production'>
In [10]:
rules[4].rhs(), type(rules[0].rhs()[0])
Out[10]:
(('vinken',), nltk.grammar.Nonterminal)
In [11]:
rules[10].rhs(), type(rules[10].rhs()[0])
Out[11]:
(('61',), str)
In [12]:
rules[0].lhs(), type(rules[0].lhs())
Out.[12]:
(S, nltk.grammar.Nonterminal)
In [13]:
print(cnf train[0].leaves())
['pierre', 'vinken', ',', '61', 'years', 'old', ',', 'will', 'join', 'the', 'board', 'as', 'a', 'n
onexecutive', 'director', 'nov.', '29', '.']
```

ANSWER HERE

- productions(): Returns the list of CFG rules that "explain" the tree. For each non-terminal node in the tree, tree.productions() will return a production with the parent node as LHS and the children as RHS.
- rhs(): Returns the right-hand side of a production its children nodes
- Ihs(): Returns the left-hand side of a production its parent node.
- leaves(): Returns the leaves of a production meaning the sentence formed using the set of rules

To count the number of unique rules, nonterminals and terminals, pleaseimplement functions **collect_rules**, **collect_nonterminals**, **collect_terminals**

```
In [14]:
```

```
def collect rules(train data):
    Collect the rules that appear in data.
       train data: list[Tree] --- list of Tree objects
    return:
       rules: list[nltk.grammar.Production] --- list of rules (Production objects)
        rules counts: Counter object --- a dictionary that maps one rule (nltk.Nonterminal) to its
number of
                                         occurences (int) in train data.
   rules = list()
    rules counts = Counter()
    ### YOUR CODE HERE (~ 2 lines)
    # print(train data[0].productions()
    [[rules.append(prod) for prod in rule.productions()] for rule in train data]
    rules counts = Counter(rules)
    # print(rules counts.most common(5))
    ### YOUR CODE HERE
    return rules, rules counts
def collect nonterminals(rules):
    collect nonterminals that appear in the rules
       rules: list[nltk.grammar.Production] --- list of rules (Production objects)
    return:
       nonterminals: set(nltk.Nonterminal) --- set of nonterminals
    nonterminals = list()
    ### YOUR CODE HERE (at least one line)
    for rule in rules:
     nonterminals.append(rule.lhs())
      for nonterm in rule.rhs():
       if (isinstance (nonterm, str)):
          continue
        else:
         nonterminals.append(nonterm)
    ### END OF YOUR CODE
    return set(nonterminals)
def collect terminals(rules):
    collect terminals that appear in the rules
      rules: list[nltk.grammar.Production] --- list of rules (Production objects)
       terminals: set of strings --- set of terminals
    terminals = list()
    ### YOUR CODE HERE (at least one line)
    # [[terminals.append(strings) for strings in rule.leaves()] for rule in rules]
    # print(terminals[0])
    # print(rules[0])
    for rule in rules:
```

```
for term in rule.rhs():
                                 if(isinstance(term,str)):
                                       terminals.append(term)
                                  else:
                                         continue
                 ### END OF YOUR CODE
                 return set(terminals)
  # terminals = collect_terminals(train_rules)
In [15]:
 train rules, train rules counts = collect rules(cnf train)
 nonterminals = collect nonterminals(train rules)
 terminals = collect terminals(train rules)
 In [16]:
 ### CORRECT ANSWER (19xxxx, 3xxxx, 1xxxx, 7xxx)
 len(train rules), len(set(train rules)), len(terminals), len(nonterminals)
Out[16]:
 (196646, 31656, 11367, 7869)
In [17]:
print(train rules counts.most common(5))
 [(, -> ', ', 4876), (DT -> 'the', 4726), (. -> '.', 3814), (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (PP -> IN NP, 3273), (S|<VP-.> -> VP ., (S|<VP-.> -> VP ., (S|<VP-.> -> VP .) (S|<VP-.> -> VP ., (
```

Implement the function **build_pcfg** which builds a dictionary that stores the terminal rules and non-terminal rules.

In [19]:

3003)]

```
def build_pcfg(rules_counts):
    Build a dictionary that stores the terminal rules and nonterminal rules.
       rules counts: Counter object --- a dictionary that maps one rule to its number of
occurences in train data.
   return:
        rules dict: dict(dict(dict)) --- a dictionary has a form like:
                    rules dict = {'terminals':{'NP':{'the':1000,'an':500}, 'ADJ':
{'nice':500,'good':100}},
                                  'nonterminals':{'S':{'NP@VP':1000},'NP':{'NP@NP':540}}}
    When building "rules dict", you need to use "lhs()", "rhs()" funtion and convert Nonterminal t
    All the keys in the dictionary are of type str.
    '@' is used as a special symbol to split left and right nonterminal strings.
    rules dict = dict()
    ### rules dict['terminals'] contains rules like "NP->'the'"
    ### rules dict['nonterminals'] contains rules like "S->NP@VP"
    rules dict['terminals'] = defaultdict(dict)
    rules_dict['nonterminals'] = defaultdict(dict)
    ### YOUR CODE HERE
    for rule in rules counts:
```

```
s = '0'
temp_list = []

if len(rule.rhs()) == 1:
    if isinstance(rule.rhs()[0], str):
        rules_dict['terminals'][str(rule.lhs())][rule.rhs()[0]] = rules_counts[rule]

else:
    for child in rule.rhs():
        temp_list.append(str(child))
    s = s.join(temp_list)
    # rules_dict['nonterminals'][str(rule.lhs())] = defaultdict(dict)
    rules_dict['nonterminals'][str(rule.lhs())][s] = rules_counts[rule]

### END OF YOUR CODE
return rules_dict
```

In [20]:

```
train_rules_dict = build_pcfg(train_rules_counts)
print(train_rules_dict['nonterminals']['ADJP'])
# print(train_rules_dict['nonterminals'].keys())
```

{'NP@JJ': 10, 'JJ@JJ': 8, 'JJ@S': 5, 'RB@JJ': 77, 'RB@VBN': 42, 'RB@JJR': 22, 'RB@DT': 2, 'JJ@PP': 42, 'CD@NN': 57, 'JJ@PRN': 1, 'QP@NN': 8, 'QP@-NONE-': 58, 'NP-ADV@JJR': 3, 'VBG@ADJP|<CC-VBG>': 1, 'RB@ADJP|<JJ-CC-JJ>': 2, 'ADJP@PP': 9, 'RB@ADJP|<RB-JJ>': 3, 'VBN@PP': 2, 'QP@JJR': 1, 'RBR@JJ': 22, '\$@ADJP|<JJ--NONE->': 7, 'ADVP-TMP+RB@VBN': 1, 'CD@NNS': 1, 'RB@ADJP|<JJ-PP>': 6, 'RB@ADJP|<RB -JJ-PP>': 1, 'JJ@ADJP|<CD-NN>': 1, 'ADVP@JJ': 4, 'ADJP@ADJP|<,-ADJP-,-ADJP>': 1, 'ADVP-TMP@JJ': 1, 'RBS@JJ': 18, 'JJR@ADJP|<CC-JJR>': 2, '\$@ADJP|<CD--NONE->': 28, 'ADVP-TMP+RB@ADJP|<RB-JJ>': 1, 'JJ @NP-TMP': 10, 'ADVP@VBN': 2, 'JJ@ADJP|<CC-JJ>': 18, 'JJ@ADJP|<CC-NNP>': 1, 'JJ@SBAR': 2, 'JJ@PP-LO C': 2, 'ADJP+JJ@ADJP|<,-CC-ADJP-,>': 1, 'NN@NN': 1, 'ADJP+JJ@ADJP|<CC-ADJP>': 1, 'RBS@ADJP|<RB-JJ> ': 1, 'NNP@JJ': 6, 'JJ@ADJP|<,-JJ-CC-JJ>': 3, 'JJR@JJ': 4, 'NNS@ADJP|<CC-NNS>': 1, 'NNP@ADJP|<,-JJ>': 8, 'JJ@ADJP|<,-JJ-,-JJ-CC-NN>': 1, 'JJ@NP-TMP+CD': 11, 'NNP@NNP': 3, 'ADVP@ADJP|<JJ-PP>': 1, "RB@ADJP|<``-JJ-CC-JJ-''>": 1, 'RB@RB': 4, 'JJ@ADJP|<,-CC-JJ>': 1, 'ADJP@ADJP|<CC-ADJP>': 1, 'RB@A DJP|<VBN-PRT+RP-PP>': 1, 'JJS@JJ': 3, 'VBN@PP-CLR': 1, 'RB@ADJP|<JJ-PP-LOC>': 1, 'JJ@JJR': 1, 'JJ@ ADJP|<CC-VBG>': 1, 'JJ@ADJP|<RB-SBAR+-NONE->': 1, 'NN@ADJP|<RB-SBAR+-NONE->': 1, 'QP@ADJP|<RB-JJ>' : 1, 'ADVP-TMP+RB@JJ': 1, 'ADJP@ADJP|<,-CC-ADJP>': 1, 'RB@ADJP|<RB-PP-S>': 1, 'RB@ADJP|<RB-S>': 1, 'NN@JJ': 1, 'ADJP@ADJP|<CC-ADJP+JJ>': 2, 'RB@ADJP|<RBR-JJ>': 2, 'NNS@ADJP|<PRN-VBN>': 1, 'RB@ADJP| <JJR-IN>': 1, 'JJR@VBN': 1, 'RB@VBG': 2, 'RBS@VBN': 1, 'JJR@PP': 1, 'RBR@ADJP|<JJ-PP>': 1, 'ADVP+RB@ADJP|<RB-JJR>': 1, 'DT@ADJP|<ADJP+JJ-CC-ADJP+JJ>': 1, "``@ADJP|<JJ-''-CC-``-JJ>": 1, 'JJS@ADJP|<JJ-S>': 1, 'JJ@RB': 1, '\$@ADJP|<CD-JJ>': 1, 'CD@ADJP|<CD-NN>': 8, 'JJ@PP-TMP': 2, 'JJ@J JS': 1, 'VB0JJR': 1, 'NP-ADV0JJ': 1, 'JJ0NP': 1, 'VBN0JJ': 1, 'NNP0ADJP|<NNP-,-JJ>': 1, 'ADJP+JJR@PP': 1, 'NN@ADJP|<CC-NN>': 1, '\$@JJ': 1, "``@ADJP|<RB-''-JJR>": 1, 'JJ@ADJP|<CC-RB>': 1, 'CD@JJ': 1, 'IN@NN': 1, 'NP@ADJP|<JJ-PP>': 1, '``@ADJP|<RB-VBN>': 1, 'NNP@ADJP|<,-NNP-JJ>': 1, 'AD JP@ADJP|<CC-ADJP+JJR>': 1, 'JJ@VBN': 1, 'VBN@ADJP|<CC-JJ>': 1}

Question 4

Estimate the probability of rule \$NP\rightarrow NNP@NNP\$

```
In [21]:
```

```
np_count = 0
for val in train_rules_dict['nonterminals']['NP'].values():
    np_count += val
# print(np_count)
print("The probability of the rule is: " + str(train_rules_dict['nonterminals']['NP']['NNP@NNP']/n
p_count))
# prob = train_rules_dict['nonterminals'][NP]
```

The probability of the rule is: 0.03950843529348353

Question 5

Find the terminal symbols in "cnf_test[0]" that never appeared in the PCFG we built.

In [22]:

```
# print(cnf_test[0].leaves())
# for dic in train_rules_dict['terminals'].values():
ret = cnf_test[0].leaves()
for terminal in cnf_test[0].leaves():
    for dic in train_rules_dict['terminals'].values():
        # print(dic)
    if terminal in dic:
        try:
        ret.remove(terminal)
        except ValueError:
        pass
print(ret)
```

['constitutional-law']

Question 6

We can use smoothing techniques to handle these cases. A simple smoothing method is as follows. We first create a new "unknown" terminal symbol \$unk\$.

Next, for each original non-terminal symbol \$A\in N\$, we add one new rule \$A \rightarrow unk\$ to the original PCFG.

The smoothed probabilities for all rules can then be estimated as: $\$q_{\text{smooth}}(A \rightarrow \theta) = \frac{(A \rightarrow \theta)}{(A \rightarrow \theta)}(A \rightarrow \theta) = \frac{(A \rightarrow \theta)}$

Implement the function smooth_rules_prob which returns the smoothed rule probabilities

In [24]:

```
def smooth_rules_prob(rules_counts):
   params:
       rules counts: dict(dict(dict)) --- a dictionary has a form like:
                      rules_counts = {'terminals':{'NP':{'the':1000,'an':500}, 'ADJ':
{'nice':500,'good':100}},
                                       'nonterminals':{'S':{'NP@VP':1000},'NP':{'NP@NP':540}}}
   return:
       rules prob: dict(dict(dict)) --- a dictionary that has a form like:
                               rules prob = {'terminals':{'NP':{'the':0.6,'an':0.3, '<unk>':0.1},
                                                            'ADJ':{'nice':0.6,'good':0.3,'<unk>':0.1
                                                           'S':{'<unk>':0.01}}}
                                              'nonterminals':{'S':{'NP@VP':0.99}}
   rules_prob = copy.deepcopy(rules_counts)
   unk = ' < unk > '
   ### Hint: don't forget to consider nonterminal symbols that don't appear in
rules counts['terminals'].keys()
   ### YOUR CODE HERE
   for dic in rules prob['terminals'].values():
     total = 0
     for val in dic.values():
       total += val
     for key, val in dic.items():
       # print(key,val)
       dic[key] = val/(total+1)
     dic[unk] = 1/(total+1)
      # print(dic[unk])
   for key,dic in rules_prob['nonterminals'].items():
     total = 0
     for val in dic.values():
       total += val
      for k, val in dic.items():
       # print(key,val)
       dic[k] = val/(total+1)
      if key not in rules prob['terminals']:
```

```
rules prob['terminals'][key] = {unk: (1/(total+1))}
    ### END OF YOUR CODE
    return rules_prob
In [25]:
s rules prob = smooth rules prob(train rules dict)
terminals.add('<unk>')
In [26]:
print(s_rules_prob['nonterminals']['S']['NP-SBJ@S|<VP-.>'])
print(s rules prob['nonterminals']['S']['NP-SBJ-1@S|<VP-.>'])
print(s_rules_prob['nonterminals']['NP']['NNP@NNP'])
print(s_rules_prob['terminals']['NP'])
0.1300172371337109
0.025240088648116228
0.039506305917861376
{'<unk>': 5.389673385792821e-05}
In [27]:
len(terminals)
Out[27]:
11368
```

CKY Algorithm

Similar to the Viterbi algorithm, the CKY algorithm is a dynamic-programming algorithm. Given a PCFG \$G=(N, \Sigma, \S, \R, \q)\$, we can use the CKY algorithm described in class to find the highest scoring parse tree for a sentence.

First, let us complete the CKY function from scratch using only Python built-in functions and the Numpy package.

The output should be two dictionaries \$\pi\$ and \$bp\$, which store the optimal probability and backpointer information respectively.

Given a sentence w_0 , w_1 , ..., w_{n-1} , v_i , $v_$

Question 7

Implement CKY function and run the test code to check your implementation.

```
In [28]:
```

```
{ 'mice : u.o, 'good ::u.o, \umk> ::u.i},
                                                                        'S':{'<unk>':0.01}}}
                                                          'nonterminals':{'S':{'NP@VP':0.99}}
    return:
       score: dict() --- score[(i,i+span)][root] represents the highest score for the parse
(sub) tree that has the root "root"
                          across words w_i, w_{i+1},..., w_{i+span-1}.
        back: dict() --- back[(i,i+span)][root] = (split , left child, right child); split: int;
                         left child: str; right child: str.
    score = defaultdict(dict)
    back = defaultdict(dict)
    sent len = len(sent)
    ### YOUR CODE HERE
    word set = set()
    for term in rules prob['terminals'].keys():
        for word in rules_prob['terminals'][term].keys():
           word set.add(word)
    split = 0
    for i in range(sent_len):
      for term key,term val in rules prob['terminals'].items():
        if sent[i] in word set and sent[i] in rules prob['terminals'][term key]:
            score[(i,i+1)][term key] = math.log(rules prob['terminals'][term key][sent[i]])
            back[(i,i+1)][term key] = (split, sent[i], None)
        elif sent[i] in word set and sent[i] not in rules prob['terminals'][term key]:
          continue
        else:
          score[(i,i+1)][term key] = math.log(rules prob['terminals'][term key]['<unk>'])
          back[(i,i+1)][term key] = (split,sent[i],None)
    for span in range(2,sent len+1):
      for begin in range(0,sent len-span+1):
        end=begin+span
        for split in range(begin+1,end):
          for nonterm_key in rules_prob['nonterminals'].keys():
            for combined child, probability in rules prob['nonterminals'][nonterm key].items():
              left child, right child = combined child.split('@')
              left childs = back[(begin, split)]
              right childs = back[(split,end)]
              if left child in left childs and right child in right childs:
                if nonterm key not in score[(begin,end)]:
                  score[(begin,end)][nonterm_key] = math.log(probability) + score[(begin,split)]
[left_child]+ score[(split,end)][right_child]
                  back[(begin,end)][nonterm_key] = (split,left_child,right_child)
                else:
                  temp = math.log(probability) + score[(begin,split)][left child]+ score[(split,end
)][right child]
                  if temp > score[(begin,end)][nonterm key]:
                      score[(begin,end)][nonterm_key] = temp
                      back[(begin,end)][nonterm key] = (split,left child,right child)
    ### END OF YOUR CODE
    return score, back
sent = cnf train[0].leaves()
score, back = CKY(sent, s rules prob)
print(score[(0, len(sent))]['S'])
4
```

T-- [001

```
params:
       back: dict() --- back[(i,i+span)][X] = (split , left_child, right_child); split:int;
left_child: str; right_child: str.
       root: tuple() --- (begin, end, nonterminal symbol), e.g., (0, 10, 'S
    return:
       tree: nltk.tree.Tree
   begin = root[0]
   end = root[1]
   root label = root[2]
   ### YOUR CODE HERE
   split, left_label, right label = back[(begin, end)][root label]
   if split == 0:
       tree = Tree(root label, [left label])
   else:
       left_child = build_tree(back, (begin, split, left_label), nonterminals)
       right_child = build_tree(back, (split, end, right_label), nonterminals)
       tree = Tree(root label, [left child, right child])
   ### END OF YOUR CODE
   return tree
```

```
In [32]:
build_tree(back, (0, len(sent), 'S'), nonterminals).pprint()
(S
  (NP-SBJ
    (NP (NNP pierre) (NNP vinken))
    (NP-SBJ|<,-NP-,>
      (, ,)
      (NP-SBJ|<NP-,>
        (NP (CD 61) (NP | < NNS-JJ> (NNS years) (JJ old)))
        (, ,))))
  (S|<VP-.>
    (VP
      (MD will)
      (VP
        (VB join)
        (VP|<NP-PP-CLR-NP-TMP>
          (NP (DT the) (NN board))
          (VP|<PP-CLR-NP-TMP>
            (PP-CLR
              (IN as)
              (NP
                 (DT a)
                 (NP|<JJ-NN> (JJ nonexecutive) (NN director))))
            (NP-TMP (NNP nov.) (CD 29))))))
    (. .)))
```

In [33]:

```
def set leave index(tree):
   Label the leaves of the tree with indexes
       tree: original tree, nltk.tree.Tree
   Return:
      tree: preprocessed tree, nltk.tree.Tree
             _ in enumerate(tree.leaves()):
   for idx,
       tree_location = tree.leaf_treeposition(idx)
       non terminal = tree[tree location[:-1]]
       non_terminal[0] = non_terminal[0] + "_" + str(idx)
   return tree
def get_nonterminal_bracket(tree):
   Obtain the constituent brackets of a tree
      tree: tree, nltk.tree.Tree
   Return:
       nonterminal brackets: constituent brackets, set
   nonterminal brackets = set()
   for tr in tree.subtrees():
       label = tr.label()
       #print(tr.leaves())
       if len(tr.leaves()) == 0:
           continue
       start = tr.leaves()[0].split(' ')[-1]
       end = tr.leaves()[-1].split(' ')[-1]
       if start != end:
           nonterminal brackets.add(label+'-('+start+':'+end+')')
   return nonterminal brackets
def word2lower(w, terminals):
   Map an unknow word to "unk"
   return w.lower() if w in terminals else '<unk>'
```

In [34]:

```
correct count = 0
pred count = 0
gold count = 0
for i, t in enumerate(cnf test):
   #Protect the original tree
   t = copy.deepcopy(t)
    sent = t.leaves()
   #Map the unknow words to "unk"
    sent = [word2lower(w.lower(), terminals) for w in sent]
    #CKY algorithm
    score, back = CKY(sent, s_rules_prob)
    candidate_tree = build_tree(back, (0, len(sent), 'S'), nonterminals)
    #Extract constituents from the gold tree and predicted tree
    pred_tree = set_leave_index(candidate_tree)
    pred brackets = get nonterminal bracket(pred tree)
    #Count correct constituents
    pred count += len(pred brackets)
    gold tree = set leave index(t)
    gold brackets = get nonterminal bracket(gold tree)
    gold count += len(gold brackets)
    current correct num = len(pred brackets.intersection(gold brackets))
    correct count += current correct num
    print('#'*20)
    print('Test Tree:', i+1)
    print (!Constituent number in the predicted tree.! len(pred brackets))
```

```
print( constituent number in the gold tree:', len(gold_brackets))
    print('Correct constituent number:', current correct num)
recall = correct_count/gold_count
precision = correct count/pred count
f1 = 2*recall*precision/(recall+precision)
###################
Test Tree: 1
Constituent number in the predicted tree: 20
Constituent number in the gold tree: 20
Correct constituent number: 14
##################
Test Tree: 2
Constituent number in the predicted tree: 54
Constituent number in the gold tree: 54
Correct constituent number: 28
####################
Test Tree: 3
Constituent number in the predicted tree: 30
Constituent number in the gold tree: 30
Correct constituent number: 23
####################
Test Tree: 4
Constituent number in the predicted tree: 17
Constituent number in the gold tree: 17
Correct constituent number: 16
##################
Test Tree: 5
Constituent number in the predicted tree: 32
Constituent number in the gold tree: 32
Correct constituent number: 26
####################
Test Tree: 6
Constituent number in the predicted tree: 40
Constituent number in the gold tree: 40
Correct constituent number: 18
####################
Test Tree: 7
Constituent number in the predicted tree: 22
Constituent number in the gold tree: 22
Correct constituent number: 7
#####################
Test Tree: 8
Constituent number in the predicted tree: 18
Constituent number in the gold tree: 18
Correct constituent number: 6
#####################
Test Tree: 9
Constituent number in the predicted tree: 28
Constituent number in the gold tree: 28
Correct constituent number: 16
####################
Test Tree: 10
Constituent number in the predicted tree: 40
Constituent number in the gold tree: 40
Correct constituent number: 8
In [35]:
print('Overall precision: {:.3f}, recall: {:.3f}, f1: {:.3f}'.format(precision, recall, f1))
Overall precision: 0.538, recall: 0.538, f1: 0.538
In [36]:
print('Overall precision: {:.3f}, recall: {:.3f}, f1: {:.3f}'.format(precision, recall, f1))
Overall precision: 0.538, recall: 0.538, f1: 0.538
In [37]:
```

```
et=time.time()
print(et - st)

874.4197461605072

In []:
```