

Analytical Component:**Problem 1:**

$$\begin{aligned}
 \text{a) } P(\text{tags, words}) &= \text{PCFG of Parse Tree \#1} + \text{PCFG of Parse Tree \#2 (solved in problem 2b)} \\
 &= 0.006 + 0.0072 \\
 &= 0.0132
 \end{aligned}$$

b)

PCFG from problem

S → NP VP	[1.0]
NP → Adj NP	[0.3]
NP → PRP	[0.1]
NP → N	[0.6]
VP → V NP	[0.8]
VP → Aux V NP	[0.2]
PRP → they	[1.0]
N → potatoes	[1.0]
Adj → baking	[1.0]
V → baking	[0.5]
V → are	[0.5]
Aux → are	[1.0]

Transition Probabilities:

$$P(\text{Adj} | \text{Start}) = P(S \rightarrow \text{NP VP}) * P(\text{NP} \rightarrow \text{Adj NP}) = 1.0 * 0.3 = 0.3$$

$$P(\text{PRP} | \text{Start}) = P(S \rightarrow \text{NP VP}) * P(\text{NP} \rightarrow \text{PRP}) = 1.0 * 0.1 = 0.1$$

$$P(N | \text{Start}) = P(S \rightarrow \text{NP VP}) * P(\text{NP} \rightarrow N) = 1.0 * 0.6 = 0.6$$

$$P(\text{PRP} | \text{Adj}) = P(\text{NP} \rightarrow \text{Adj NP} \rightarrow \text{Adj PRP}) = 0.1$$

$$P(N | \text{Adj}) = P(\text{NP} \rightarrow \text{Adj NP} \rightarrow \text{Adj N}) = 0.6$$

$$P(\text{Adj} | \text{Adj}) = P(\text{NP} \rightarrow \text{Adj NP} \rightarrow \text{Adj Adj NP}) = 0.3$$

$$P(V | \text{PRP}) = P(\text{VP} \rightarrow V \text{ NP} \rightarrow V \text{ PRP}) = 0.8$$

$$P(\text{Aux} | \text{PRP}) = P(\text{NP VP} \rightarrow \text{PRP VP} \rightarrow \text{PRP Aux V NP}) = 1.0 * 0.2 = 0.2$$

$$P(V | N) = P(\text{NP VP} \rightarrow N \text{ VP} \rightarrow N V \text{ NP}) = 1.0 * 0.8 = 0.8$$

$$P(\text{Aux} | N) = P(\text{NP VP} \rightarrow N \text{ VP} \rightarrow N \text{ Aux V NP}) = 1.0 * 0.2 = 0.2$$

$$P(V | \text{Aux}) = P(\text{Aux} \rightarrow V \rightarrow \text{are}) = 1.0$$

$$P(\text{PRP} | V) = P(\text{VP} \rightarrow V \text{ NP} \rightarrow V \text{ PRP}) = 0.1$$

$$P(N | V) = P(\text{VP} \rightarrow V \text{ NP} \rightarrow V N) = 0.6$$

$$P(\text{Adj} | V) = P(\text{VP} \rightarrow V \text{ NP} \rightarrow V \text{ Adj NP}) = 0.3$$

Emission Probabilities:

$$P(\text{they} | \text{PRP}) = P(\text{PRP} \rightarrow \text{they}) = 1.0$$

$$P(\text{potatoes} | N) = P(N \rightarrow \text{potatoes}) = 1.0$$

$$P(\text{baking} | \text{Adj}) = P(\text{Adj} \rightarrow \text{baking}) = 1.0$$

$$P(\text{baking} | V) = P(V \rightarrow \text{baking}) = 0.5$$

$$P(\text{are} | V) = P(V \rightarrow \text{are}) = 0.5$$

$$P(\text{are} | \text{Aux}) = P(\text{Aux} \rightarrow \text{are}) = 1.0$$

- c) No, it is not possible to translate any PCFG into an HMM that produces the identical joint probability as the PCFG. PCFGs generate languages that can be expressed by context free grammar whereas HMMs generate languages that can be expressed by regular grammar. According to the Chomsky Hierarchy, the context free grammar is a generalization of the regular grammar. As such, a grammatical construct that can be described by regular grammar can be described by a context-free grammar, but that is not the case for the other way around. Hence, there are some PCFGs that cannot be translated to an HMM producing identical joint probabilities.

Problem 2:

a)

0 they 1 are 2 baking 3 potatoes 4

Chart 0:

S_0 S → • NP VP [0,0] init
 S_1 NP → • Adj NP [0,0] predict S_0
 S_2 NP → • PRP [0,0] predict S_0
 S_3 NP → • N [0,0] predict S_0
 S_4 Adj → • baking [0,0] predict S_1
 S_5 PRP → • they [0,0] predict S_2
 S_6 N → • potatoes [0,0] predict S_3

Chart 1:

S_7 PRP → they • [0,1] scan S_5
 S_8 NP → PRP • [0,1] complete S_2 with S_7
 S_9 S → NP • VP [0,1] complete S_0 with S_7
 S_10 VP → • V NP [1,1] predict S_9
 S_11 VP → • Aux V NP [1,1] predict S_9
 S_12 V → • baking [1,1] predict S_10
 S_13 V → • are [1,1] predict S_10
 S_14 Aux → • are [1,1] predict S_11

Chart 2:

S_15 V → are • [1,2] scan S_13
 S_16 Aux → are • [1,2] scan S_14
 S_17 VP → V • NP [1,2] complete S_10 with S_15
 S_18 VP → Aux • V NP [1,2] complete S_11 with S_16
 S_19 NP → • Adj NP [2,2] predict S_17
 S_20 NP → • PRP [2,2] predict S_17
 S_21 NP → • N [2,2] predict S_17
 S_22 V → • baking [2,2] predict S_18
 S_23 V → • are [2,2] predict S_18
 S_24 Adj → • baking [2,2] predict S_19
 S_25 PRP → • they [2,2] predict S_20
 S_26 N → • potatoes [2,2] predict S_21

Chart 3:

S_27 V → baking • [2,3] scan S_22
 S_28 Adj → baking • [2,3] scan S_24
 S_29 VP → Aux V • NP [1,3] complete S_18 with S_27
 S_30 NP → Adj • NP [2,3] complete S_19 with S_28
 S_31 NP → • Adj NP [3,3] predict S_29, predict S_30
 S_32 NP → • PRP [3,3] predict S_29, predict S_30
 S_33 NP → • N [3,3] predict S_29, predict S_30
 S_34 Adj → • baking [3,3] predict S_31
 S_35 PRP → • they [3,3] predict S_32
 S_36 N → • potatoes [3,3] predict S_33

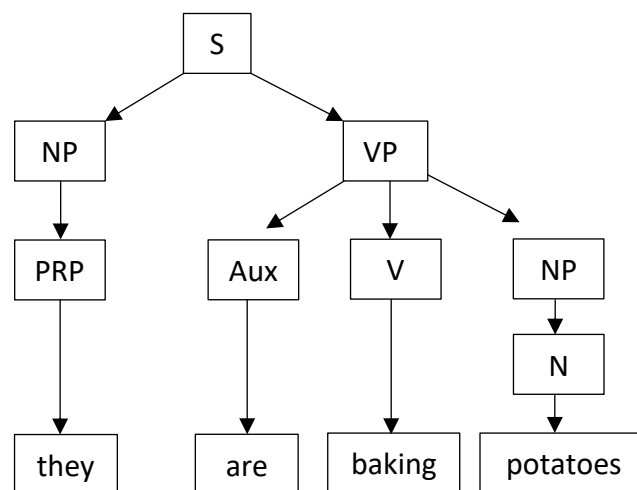
Chart 4:

S_37 N → potatoes • [3,4] scan S_36
 S_38 NP → N • [3,4] complete S_33 with S_37
 S_39 VP → Aux V NP • [1,4] complete S_29 with S_38
 S_40 NP → Adj NP • [2,4] complete S_30 with S_38
 S_41 S → NP VP • [0,4] complete S_9 with S_39
 S_42 VP → V NP • [1,4] complete S_17 with S_40
 S_43 S → NP VP • [0,4] complete S_9 with S_42

b)

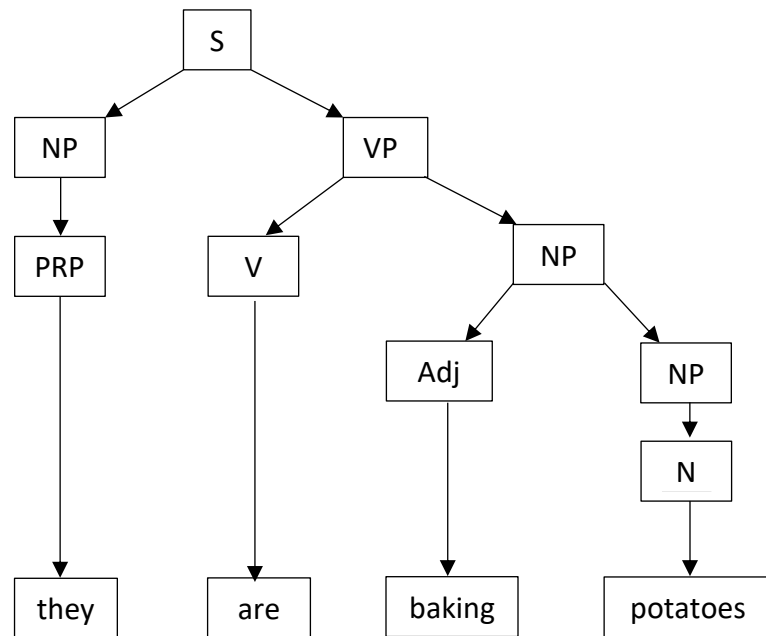
Final Parsing from part a:

Parse Tree #1



$$\text{PCFG} = 1.0 * 0.1 * 1.0 * 0.2 * 1.0 * 0.5 * 0.6 * 1.0 = 0.006$$

Parse Tree #2:



$$\text{PCFG} = 1.0 * 0.1 * 1.0 * 0.8 * 0.5 * 0.3 * 1.0 * 0.6 * 1.0 = 0.0072$$

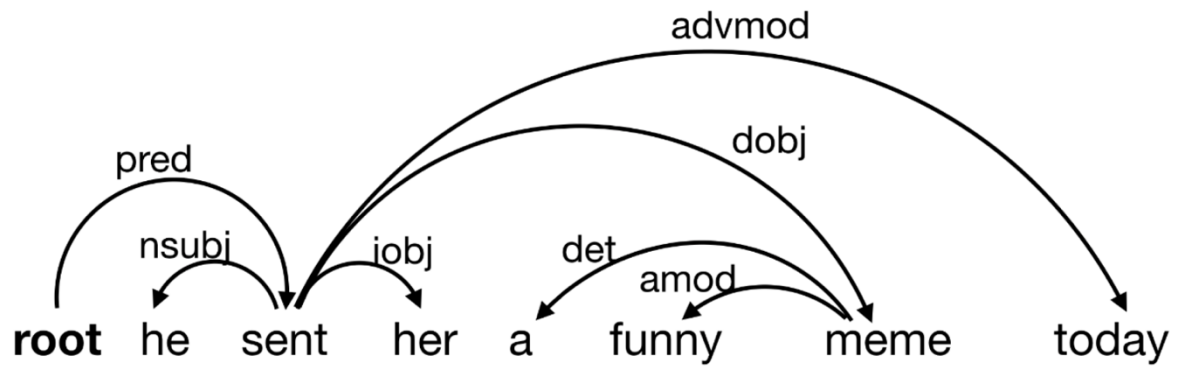
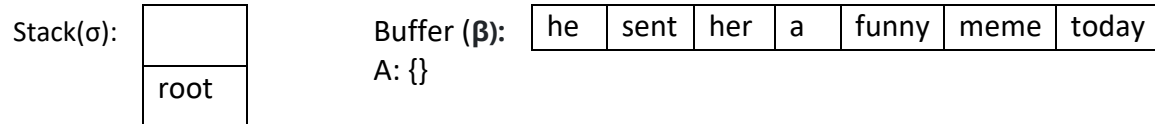
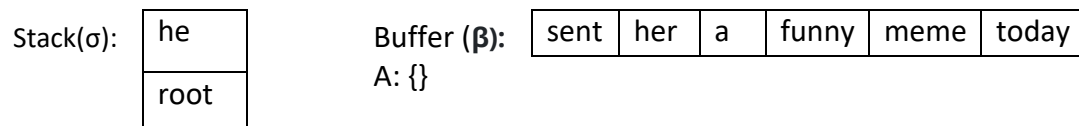
Problem 3:

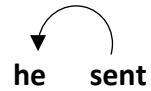
a)

- **Rules of the form $A \rightarrow B$:** For production rules that have a single non-terminal on the right-hand side (**B**), the general rule is to replace that non-terminal with a terminal. For instance, say we also have $B \rightarrow C$ where **C** is a terminal. Hence, in $A \rightarrow B$ we replace **B** with **C** and get rid of the $B \rightarrow C$ rule so that we get to have $A \rightarrow C$.
- **Rules of the form $A \rightarrow B C D E$:** For production rules that have three or more non-terminals on the right-hand side (**B C D E**), the general rule is to separate the non-terminals into sections, each consisting of exactly two non-terminals. Then create new production rules such that there's a single non-terminal on the left-side and maximum of two non-terminals on the right-hand side and use them to change the original rule. For instance, we can set the following rules $L \rightarrow BC$ and $R \rightarrow DE$ such that **L** and **R** are non terminals. Then, we can have the non-terminal **A** in this new rule: $A \rightarrow LR$.

CFG in CNF form:

$S \rightarrow NP VP$
 $NP \rightarrow Adj NP$
 $NP \rightarrow they$
 $NP \rightarrow potatoes$
 $VP \rightarrow V NP$
 $L \rightarrow Aux V$
 $VP \rightarrow L NP$
 $Adj \rightarrow baking$
 $V \rightarrow baking$
 $V \rightarrow are$
 $Aux \rightarrow are$

Problem 4)**1) Initial State****Next Transition: Shift****2) Next Transition: Left-Arc**

3) Next Transition: Shift

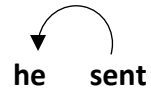
Stack(σ):

root

Buffer (β):

sent	her	a	funny	meme	today
------	-----	---	-------	------	-------

A:{sent, nsubj, he}

4) Next Transition: Right-Arc

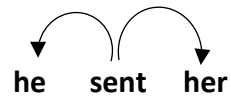
Stack(σ):

sent
root

Buffer (β):

her	a	funny	meme	today
-----	---	-------	------	-------

A:{sent, nsubj, he}

5) Next Transition: Shift

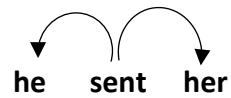
Stack(σ):

root

Buffer (β):

sent	a	funny	meme	today
------	---	-------	------	-------

A:{sent, nsubj, he}, {sent, iobj, her}

6) Next Transition: Shift

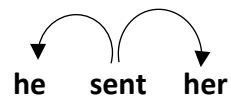
Stack(σ):

sent
root

Buffer (β):

a	funny	meme	today
---	-------	------	-------

A:{sent, nsubj, he}, {sent, iobj, her}

7) Next Transition: Shift

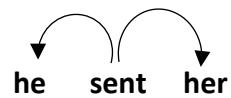
Stack(σ):

a
sent
root

Buffer (β):

funny	meme	today
-------	------	-------

A:{sent, nsubj, he}, {sent, iobj, her}

8) Next Transition: Left-Arc

Stack(σ):

funny
a
sent
root

Buffer (β):

meme	today
------	-------

A:{sent, nsubj, he}, {sent, iobj, her}

9) Next Transition: Left-Arc

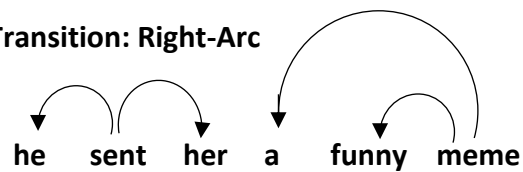
Stack(σ):

a
sent
root

Buffer (β):

meme	today
------	-------

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}

10) Next Transition: Right-Arc

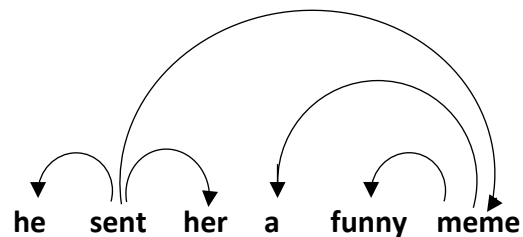
Stack(σ):

sent
root

Buffer (β):

meme	today
------	-------

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}, {meme, det, a}

11) Next Transition: Shift

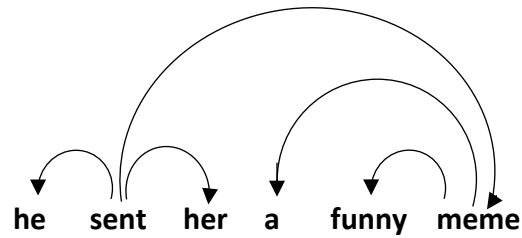
Stack(σ):

root

Buffer (β):

sent	today
------	-------

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}, {meme, det, a}, {sent, dobj, meme}

12) Next Transition: Right-Arc

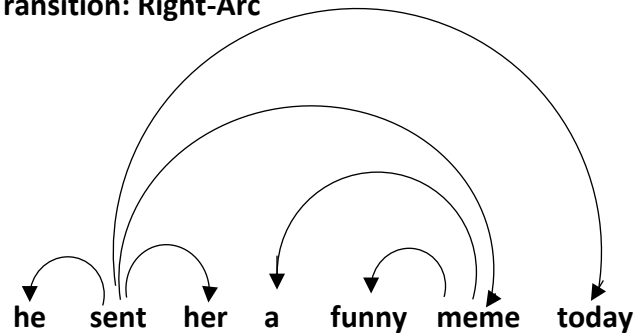
Stack(σ):

sent
root

Buffer (β):

today

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}, {meme, det, a}, {sent, dobj, meme}

13) Next Transition: Right-Arc

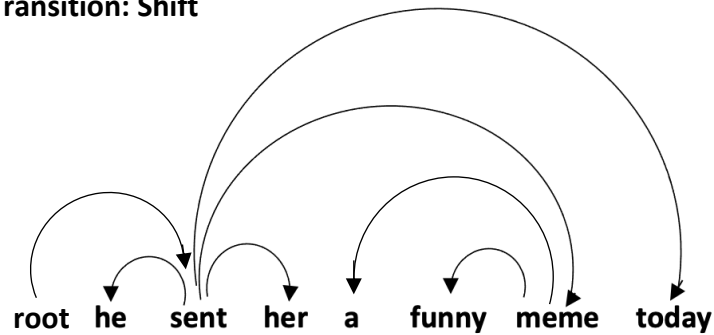
Stack(σ):

root

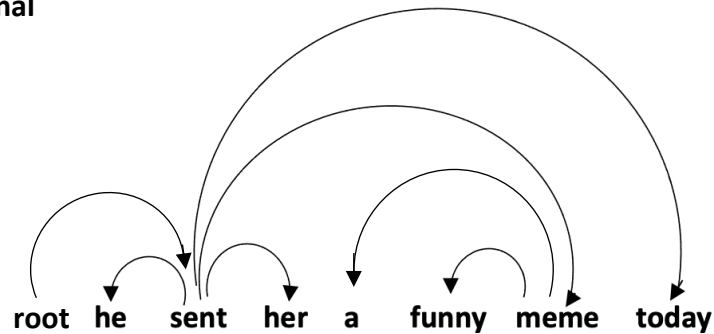
Buffer (β):

sent

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}, {meme, det, a}, {sent, dobj, meme}, {sent, advmod, today}

14) Next Transition: ShiftStack(σ): Buffer (β):

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}, {meme, det, a}, {sent, dobj, meme}, {sent, advmod, today}, {root, pred, sent}

15) TerminalStack(σ): Buffer (β):

A: {sent, nsubj, he}, {sent, iobj, her}, {meme, amod, funny}, {meme, det, a}, {sent, dobj, meme}, {sent, advmod, today}, {root, pred, sent}

Programming Component:**Part 1 - reading the grammar and getting started**

```
def verify_grammar(self):  
    """  
    Return True if the grammar is a valid PCFG in CNF.  
    Otherwise return False.  
    """  
    # TODO, Part 1  
  
    lhs_sum = []  
    lhs = self.lhs_to_rules #lists of rules(tuples) where tuple[0] is the lhs  
    #key : non terminal, v : list of its tuples  
  
    for non_terminal, list in lhs.items():  
        for tuple in list:  
            #check if non-terminal is uppercase  
            if tuple[0].isupper():  
                prob = tuple[2]  
                lhs_sum.append(prob)  
            #check if sum == 1  
        print(round(math.fsum(lhs_sum), 2))  
        if round(math.fsum(lhs_sum), 2) == 1.0:  
            lhs_sum = []  
        else:  
            return False  
  
    return True
```

Part 2 - Membership checking with CKY

```

def is_in_language(self, tokens):
    """
    Membership checking. Parse the input tokens and return True
    if
    the sentence is in the language described by the grammar.
    Otherwise
    return False
    """
    # TODO, part 2
    n = len(tokens) #number of words
    table = defaultdict() #initialization

    for i in range(0, n):
        for j in range(i + 1, n + 1):
            table[(i, j)] = defaultdict() #initialization

            if i + 1 == j: # diagonal tuple
                if self.grammar.rhs_to_rules.get((tokens[i],)):
                    k = (tokens[i],)
                    rules = self.grammar.rhs_to_rules[k]
                    #print("k = ", k)
                    #print("rules = ", rules)
                    for rule in rules:
                        table[(i, j)][rule[0]] = tokens[i]

    for length in range(2, n + 1):
        for i in range(0, n - length + 1):
            j = i + length
            for k in range(i + 1, j):
                for key in self.grammar.rhs_to_rules.keys():
                    for t1 in table[(i, k)]:
                        for t2 in table[(k, j)]:
                            if key == (t1, t2):
                                #print("key = ", key)
                                rules =
self.grammar.rhs_to_rules.get(key)
                                #print("rules = ", rules)
                                for r in rules:
                                    #print((t1, i, k), (t2, k,
j)))
                                    table[(i, j)][r[0]] = ((t1,
i, k), (t2, k, j))

    if self.grammar.startsymbol in table[(0, n)]:
        return True

```

```
else:  
    return False
```

Part 3 - Parsing with backpointers

```

def parse_with_backpointers(self, tokens):
    """
    Parse the input tokens and return a parse table and a
    probability table.
    """
    # TODO, part 3

    """
    TODO: Write the method parse_with_backpointers(self,
    tokens). You should modify your CKY implementation from part 2,
    but use
    (and return) specific data structures. The method should
    take a list of tokens as input and
    returns a) the parse table b) a probability table. Both
    objects should be constructed during
    parsing. They replace whatever table data structure you used
    in part 2.
    """

    table= defaultdict() #initialization
    probs = defaultdict() #initialization
    n = len(tokens) # number of words

    for i in range(0, n):
        for j in range(i+1, n+1):
            table[(i, j)] = defaultdict() #initialization
            probs[(i, j)] = defaultdict() #initialization

            if i + 1 == j: #diagonal tuple
                if self.grammar.rhs_to_rules.get((tokens[i],)):
                    k = (tokens[i],)
                    rules = self.grammar.rhs_to_rules[k]
                    for rule in rules:
                        table[(i, j)][rule[0]] = tokens[i]
                        probs[(i, j)][rule[0]] =
math.log(rule[2])

    for length in range(2, n + 1):
        for i in range(0, n - length + 1):
            j = i + length
            for k in range(i + 1, j):
                for key in self.grammar.rhs_to_rules.keys():
                    for t1 in table[(i, k)]:
                        for t2 in table[(k, j)]:
                            if key == (t1,t2):

```



```

        rules =
self.grammar.rhs_to_rules.get(key)
        p = probs[(i, k)][t1] +
probs[(k, j)][t2]

        for r in rules:
            log_prob = math.log(r[2]) +

p
            if r[0] in table[(i, j)]:
                #check for the highest

                if log_prob > probs[(i,
j)][r[0]]:
                    probs[(i, j)][r[0]]
                    table[(i, j)][r[0]]

            else:
                probs[(i, j)][r[0]] =
                table[(i, j)][r[0]] =

probability value
j)][r[0]]:
= log_prob
= ((t1, i, k), (t2, k, j))

log_prob
((t1, i, k), (t2, k, j))
    return table, probs

```

Part 4 - Retrieving a parse tree

```
def get_tree(chart, i,j,nt):  
    """  
    Return the parse-tree rooted in non-terminal nt and covering  
    span i,j.  
    """  
    # TODO: Part 4  
  
    backpointers = chart[(i,j)][nt]  
    #print(backpointers)  
    if type(backpointers) != str:  
        result = (nt,  
                  (get_tree(chart, i=backpointers[0][1],  
j=backpointers[0][2], nt=backpointers[0][0])),  
                  (get_tree(chart, i=backpointers[1][1],  
j=backpointers[1][2], nt=backpointers[1][0])))  
        return result  
    else:  
        result = (nt, backpointers)  
        return result
```

Part 5 - Evaluating the Parser

Since we were not required to write code in `evaluate_parser.py` and to only run it, this is what my parser implementation produces on the `atis3` test corpus:

Coverage: 67.24%, Average F-score (parsed sentences):
0.9526771952649747, Average F-score (all sentences):
0.6405932864712761