

Code Preparations

```
In [ ]: import nltk
        from nltk import grammar, parse
        from nltk.parse.generate import generate
        from platform import python_version
        python_version()
```

```
Out[ ]: '3.7.16'
```

```
In [ ]: print(nltk.__version__)

3.5
```

```
In [ ]: nltk.data.show_cfg('h2.fcfg')
```

```

% start S
# Grammar Rules
S[SEM=<?subj(?vp)>] -> DP[SEM=?subj] VP[SEM=?vp]
S[SEM=<?vp(?subj)>] -> NP[SEM=?subj] VP[SEM=?vp]
S[SEM=<?vp1(?subj) & ?vp2(?subj)>] -> NP[SEM=?subj] VP[SEM=?vp1] 'and' VP[SEM=?vp2]
DP[ SEM=<?X(?P)>] -> Det[SEM=?X] N[SEM=?P]
VP[SEM=?Q] -> 'is' A[SEM=?Q]
VP[SEM=?P] -> 'is' 'a' N[SEM=?P]
# This is included for testing.
VP[SEM=<\x.offend(x)>] -> 'offends'
# Transitive verb with individual object.
VP[ SEM=<?R(?n)>] -> TV[SEM=?R] NP[SEM=?n]
# Transitive verb with quantifier object.
# The object is given minimal scope.
VP[ SEM=<\m.?X(\n.(?R(n)(m)))>] -> TV[SEM=?R] DP[SEM=?X]
# Lexical Rules
A[SEM=<\n.exists c.(vowel(c) & char(n,c))>] -> 'vocalic'
A[SEM=<\n.exists c.((-vowel(c)) & char(n,c))>] -> 'consonantal'
A[SEM=<\n.exists c.(capital(n) & char(n,c))>] -> 'capitalized'
A[SEM=<\n. (n = 1) >] -> 'initial'
Det[SEM=<\P Q.all n.(P(n) -> Q(n))>] -> 'every'
Det[SEM=<\P Q.exists n.(P(n) & Q(n))>] -> 'a'
Det[SEM=<\P Q.exists n.(P(n) & Q(n))>] -> 'some'
Det[SEM=<\P Q.all n.(P(n) -> -Q(n))>] -> 'no'
N[SEM=<\n.char(n,leti)>] -> 'i'
N[SEM=<\n.char(n,lete)>] -> 'e'
N[SEM=<\n.char(n,letu)>] -> 'u'
N[SEM=<\n.char(n,letp)>] -> 'p'
N[SEM=<\n.char(n,lett)>] -> 't'
N[SEM=<\n.char(n,letk)>] -> 'k'
N[SEM=<\n.exists c.(glide(c) & char(n,c))>] -> 'glide'
N[SEM=<\n.exists c.char(n,c)>] -> 'letter'
N[SEM=<\n.exists c.(vowel(c) & char(n,c))>] -> 'vowel'
N[SEM=<\n.exists c.(-vowel(c) & char(n,c))>] -> 'consonant'
NP[SEM=<1>] -> 'letter' 'one'
NP[SEM=<2>] -> 'letter' 'two'
NP[SEM=<3>] -> 'letter' 'three'
NP[SEM=<4>] -> 'letter' 'four'
NP[SEM=<5>] -> 'letter' 'five'
TV[SEM=<\m n.le(m,n)>] -> 'follows'
TV[SEM=<\m n.le(n,m)>] -> 'precedes'

```

```

In [ ]: cp = nltk.load_parser('h2.fcfig', trace=0)
sent = 'letter two is vocalic'.split()
for tree in cp.parse(sent): print(tree)

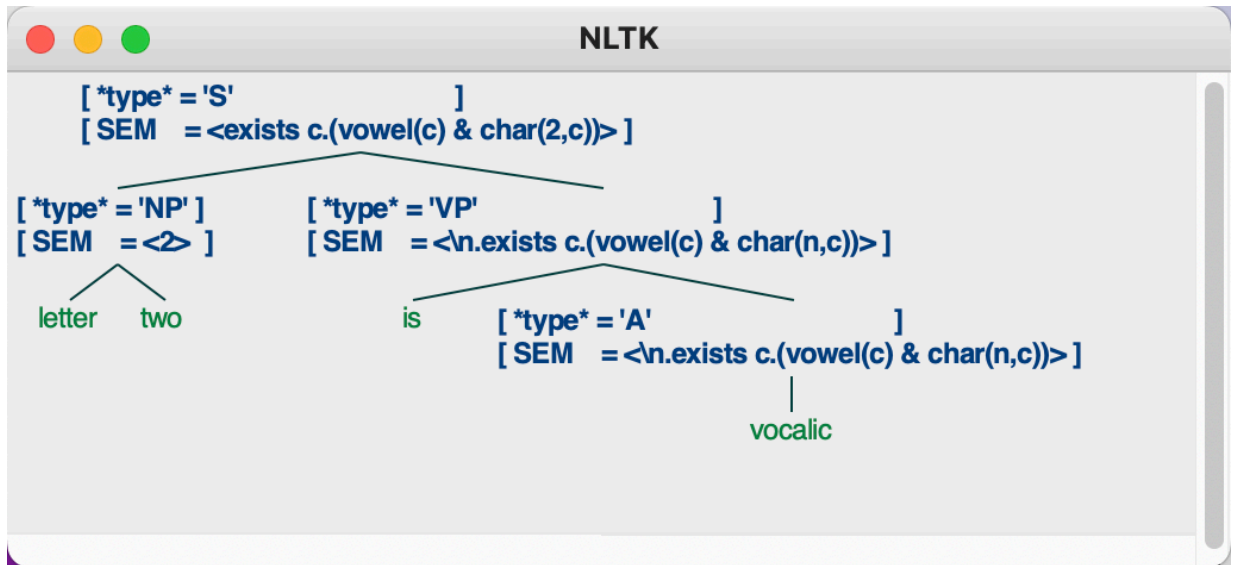
```

```

(S[SEM=<exists c.(vowel(c) & char(2,c))>]
  (NP[SEM=<2>] letter two)
  (VP[SEM=<\n.exists c.(vowel(c) & char(n,c))>]
    is
    (A[SEM=<\n.exists c.(vowel(c) & char(n,c))>] vocalic)))

```

```
In [ ]: tree.draw()
```



to use Angela Liu's implementation from mapping the word

```

In [ ]: from typing import Callable, List, Set

def to_model_str(word: str, special_rels: List[Callable[[str], str]]=[] ->
    """
    Creates the string form of the model for the input word. This string is
    By default, the function will only add the relations mapping i => i for
    mapping char => the set of tuples (i, word[i]). The `special_rels` funct
    be added to the valuation string.

    :param word: The word to create a model string for.
    :param special_rels: A list of functions that when called return a strin
    :returns: a string representing the model for word
    """
    n = len(word)
    model_str = []
    char = []
    for i in range(1, n+1):
        model_str.append(f'{i} => {i}')
        char.append((i, word[i-1]))
    model_str.append(f'char => {set(char)}'.lower())
    return '\n'.join(model_str + [rel(word) for rel in special_rels]).replac
# Angela Liu
import re

get_vowel = lambda w: f'vowel => {set(re.findall(r"[AEIOUaeiou]", w))}'.lower
get_cons = lambda w: f'cons => {}'.format(set(re.findall(r"^[AEIOUaeiou\W0-9]
follows = lambda w: f'le => {set([(i+1,j+1) for i in range(len(w)) for j in
get_capital = lambda w: f'capital => {set([m.span()[0] + 1 for m in re.findi
# Angela Liu

get_glide = lambda w: f'glide => {set(re.findall(r"[YWyw]", w))}'.lower()
# added

def emptysets(val:nltk.sem.evaluate.Valuation):
    val.update([(k,set()) for (k,v) in val.items() if v == 'set()'])

words = ['cat', 'mAtch', 'peRiLOuSy']
vals = [nltk.Valuation.fromstring(to_model_str(w, [get_vowel, get_cons, foll
for v in vals: emptysets(v)
models = [nltk.Model(val.domain, val) for val in vals]
for w, m in zip(words, models):
    print(f'{w}\n-----\n{m}\n')
# Angela Liu

cat
-----
Domain = {'3', 'a', 't', '2', 'c', '1'},
Valuation =
{'1': '1',
 '2': '2',
 '3': '3',
 'capital': set(),

```

```

'char': {('2', 'a'), ('3', 't'), ('1', 'c')},
'cons': {('t',), ('c',)},
'glide': set(),
'le': {('1', '2'), ('2', '3'), ('1', '3')},
'vowel': {('a',)}

```

mAtch

```

-----
Domain = {'h', '3', 'a', '5', 'm', 't', '2', '4', 'c', '1'},
Valuation =
{'1': '1',
 '2': '2',
 '3': '3',
 '4': '4',
 '5': '5',
 'capital': {('2',)},
 'char': {('4', 'c'), ('5', 'h'), ('3', 't'), ('1', 'm'), ('2', 'a')},
 'cons': {('m',), ('h',), ('t',), ('c',)},
 'glide': set(),
 'le': {('1', '2'),
        ('1', '3'),
        ('1', '4'),
        ('1', '5'),
        ('2', '3'),
        ('2', '4'),
        ('2', '5'),
        ('3', '4'),
        ('3', '5'),
        ('4', '5')},
 'vowel': {('a',)}

```

peRiLOuSy

```

-----
Domain = {'6', 'u', '3', '8', '5', 'r', 'o', 'y', 's', '2', '4', '9', '7', '1', 'e', 'l', 'i', 'p'},
Valuation =
{'1': '1',
 '2': '2',
 '3': '3',
 '4': '4',
 '5': '5',
 '6': '6',
 '7': '7',
 '8': '8',
 '9': '9',
 'capital': {('8',), ('6',), ('5',), ('3',)},
 'char': {('1', 'p'),
           ('2', 'e'),
           ('3', 'r'),
           ('4', 'i'),
           ('5', 'l'),
           ('6', 'o'),
           ('7', 'u'),

```

```

        ('8', 's'),
        ('9', 'y')},
'cons': {('r',), ('p',), ('l',), ('y',), ('s',))},
'glide': {('y',)},
'le': {('1', '2'),
        ('1', '3'),
        ('1', '4'),
        ('1', '5'),
        ('1', '6'),
        ('1', '7'),
        ('1', '8'),
        ('1', '9'),
        ('2', '3'),
        ('2', '4'),
        ('2', '5'),
        ('2', '6'),
        ('2', '7'),
        ('2', '8'),
        ('2', '9'),
        ('3', '4'),
        ('3', '5'),
        ('3', '6'),
        ('3', '7'),
        ('3', '8'),
        ('3', '9'),
        ('4', '5'),
        ('4', '6'),
        ('4', '7'),
        ('4', '8'),
        ('4', '9'),
        ('5', '6'),
        ('5', '7'),
        ('5', '8'),
        ('5', '9'),
        ('6', '7'),
        ('6', '8'),
        ('6', '9'),
        ('7', '8'),
        ('7', '9'),
        ('8', '9')},
'vowel': {('u',), ('o',), ('i',), ('e',))}

```

Start of the work:

Semantics of sentences about strings
 Computational Linguistics Spring 2023
 Problems Set 2

The text for this module is the NLTK book Chapter 9. Building Feature Based Grammars and Chapter 10. Analyzing the Meaning of Sentences

See also `lecture8_2023.ipynb` and `string_2023.ipynb`

The purpose of the assignment is to develop feature-based grammars that include logical semantics, and to evaluate the adequacy of the semantics by computing truth in logically constructed models. For instance, we want to be able to evaluate whether the sentence

every consonant is capitalized

is true or false as description of the word

CINEMA

or

Cinema

or

CINEmA.

In each problem n do these steps. The problem statement gives sentence s_n . See Chapters 9 and 10 for the methodology.

(i) Define a feature based grammar g_n that includes all the words in sentence s_n its lexicon. The feature grammars will usually add a word and/or construction to a base grammar which will be similar to `simple-sem.fcfcg`. This base grammar will be distributed. (It will be helpful to figure out how to add a lexical item or production to a grammar in Python. Discuss the method for this on the forum. Or you can define the grammar from scratch.)

(ii) Parse the sentence and display the tree.

(iii) Map s_n to a logical formula f_n by parsing with g_n and extracting the semantics that annotates the root.

(iv) Define a combination four words (serving as models) and the intuitive truth values of sentence s_n as a description of the word.

(v) Transform the four words into four models or valuations in the sense of Chapter 10. This can be done as in Lecture 8, or by using a function. Code for this may be shared

and discussed on the forum.

(vi) Evaluate formula fn in the four models to obtain four truth values. Compare them to the target truth values.

Work individually, except that code for mapping a word to a valuation may be shared. Post technical questions and requests for hints on the forum.

Notes The problems are selected so that quantifiers are used only in subject position. So it is not necessary (except perhaps in challenge problems) to apply the strategy from simple-sem.fcfcg to fit quantified NPs into object positions.

The words 'precedes' and 'follows' are interpreted in the sense of 'not necessarily immediately'.

Problems

1. letter two is consonantal

Base your analysis on 'letter two is vocalic', which is covered by the base grammar.

I added a grammar rule

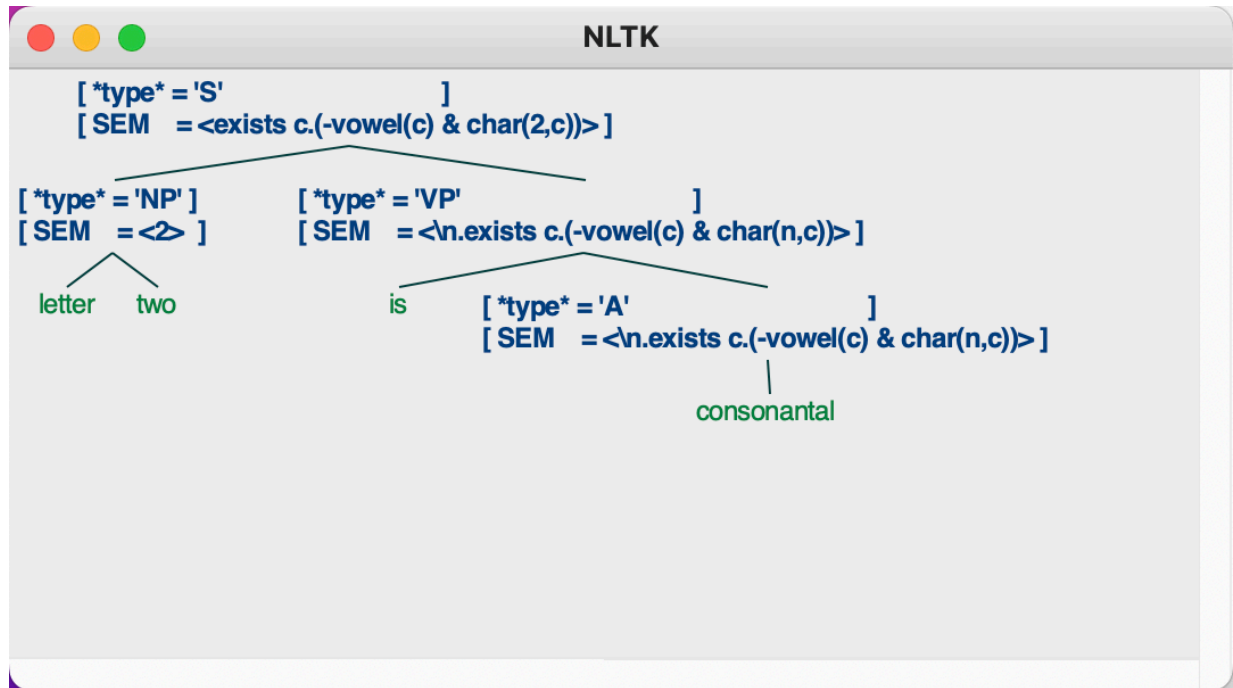
```
A[SEM=<\n.exists c.(consonant(c) & char(n,c))>] ->
'consonantal'
```

parse and display

```
In [ ]: g1 = nltk.load_parser('h2.fcfcg', trace=0)
s1 = 'letter two is consonantal'
s1_split = s1.split()
for tree in g1.parse(s1_split): print(tree)

(S[SEM=<exists c.(-vowel(c) & char(2,c))>]
 (NP[SEM=<2>] letter two)
 (VP[SEM=<\n.exists c.(-vowel(c) & char(n,c))>]
  is
  (A[SEM=<\n.exists c.(-vowel(c) & char(n,c))>] consonantal)))

In [ ]: tree.draw()
```

now the logical formula

```
In [ ]: t1=next(g1.parse(s1_split))
f1 = t1.label()['SEM']
print(f1)

exists c.(-vowel(c) & char(2,c))
```

define a sample four word

```
In [ ]: e1 = [('emu', True), ('bat', False), ('cat', False), ('aka', True)]
```

```
In [ ]: words = [e[0] for e in e1]
truths = [e[1] for e in e1]
vals = [nltk.Valuation.fromstring(to_model_str(w, [get_vowel, follows])) for
assignments = [nltk.Assignment(val.domain) for val in vals]
for val in vals: emptysets(val)
models = [nltk.Model(val.domain, val) for val in vals]
```

```
In [ ]: print(f'{s1}\n-----')
for w, a, m in zip(words, assignments, models):
    print(f'{w}\n{m.evaluate(str(f1),a)}\n-----')
```

letter two is consonantal

emu

True

bat

False

cat

False

aka

True

+ The answer is correct! emu & aka has the second letter being consonants

1. every vowel precedes letter three Add a production for 'precedes' to the grammar. Don't add it to the valuations. Instead, define the semantics of 'precedes' in terms of the primitive used for 'follows'.

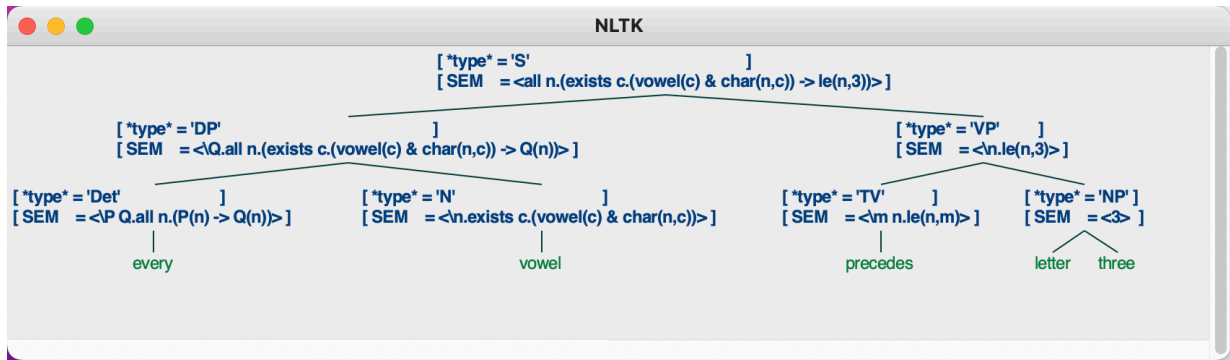
Add grammar

```
TV[SEM=<\m n.le(n,m)>] -> 'precedes'
N[SEM=<\n.exists c.(vowel(c) & char(n,c))>] -> 'vowel'
```

```
In [ ]: g2 = nltk.load_parser('h2.fcfig', trace=0,cache=False)
s2 = 'every vowel precedes letter three'
s2_split = s2.split()
for tree in g2.parse(s2_split): print(tree)

(S[SEM=<all n.(exists c.(vowel(c) & char(n,c)) -> le(n,3))>]
  (DP[SEM=<\Q.all n.(exists c.(vowel(c) & char(n,c)) -> Q(n))>]
    (Det[SEM=<\P Q.all n.(P(n) -> Q(n))>] every)
    (N[SEM=<\n.exists c.(vowel(c) & char(n,c))>] vowel))
  (VP[SEM=<\n.le(n,3)>]
    (TV[SEM=<\m n.le(n,m)>] precedes)
    (NP[SEM=<3>] letter three)))
```

```
In [ ]: tree.draw()
```



```
In [ ]: t2=next(g2.parse(s2_split))
f2 = t2.label()['SEM']
print(f2)
```

```
all n.(exists c.(vowel(c) & char(n,c)) -> le(n,3))
```

```
In [ ]: e2 = [('emmmu', False), ('aekjd', True), ('maqwr', True), ('aeiou', False)]
words = [e[0] for e in e2]
truths = [e[1] for e in e2]
vals = [nltk.Valuation.fromstring(to_model_str(w, [get_vowel, follows])) for
assignments = [nltk.Assignment(val.domain) for val in vals]
for val in vals: emptysets(val)
models = [nltk.Model(val.domain, val) for val in vals]
```

```
In [ ]: print(f'{s2}\n-----')
for w, a, m in zip(words, assignments, models):
    print(f'{w}\n{m.evaluate(str(f2),a)}\n-----')
```

```
every vowel precedes letter three
```

```
-----
```

```
emmmu
```

```
False
```

```
-----
```

```
aekjd
```

```
True
```

```
-----
```

```
maqwr
```

```
True
```

```
-----
```

```
aeiou
```

```
False
```

```
-----
```

+ The answer is correct! emmmu & aeiou is not correct

1. some vowel is capitalized

You need to add "capitalized" to the grammar, and to the valuations.

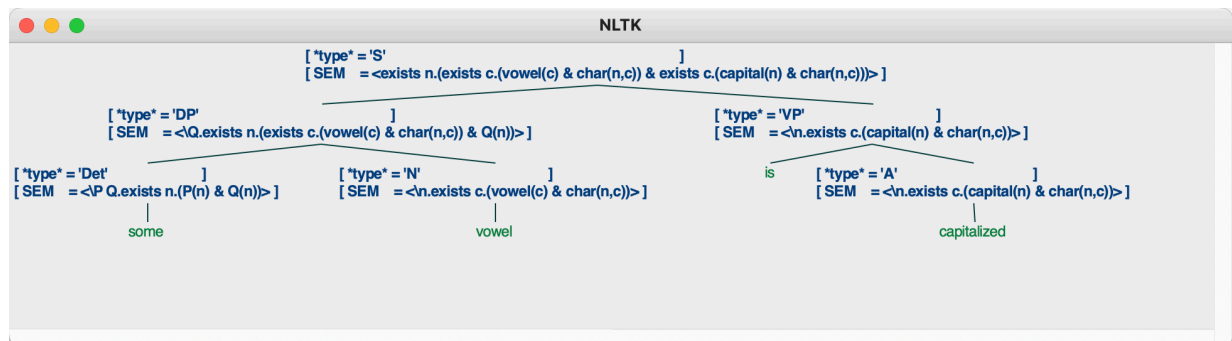
Add grammar

```
A[SEM=<\n.exists c.(capital(n) & char(n,c))>] ->
'capitalized'
```

```
In [ ]: g3 = nltk.load_parser('h2.fcfg', trace=0,cache=False)
s3 = 'some vowel is capitalized'
s3_split = s3.split()
for tree in g3.parse(s3_split): print(tree)

(S[SEM=<exists n.(exists c.(vowel(c) & char(n,c)) & exists c.(capital(n) & c
har(n,c)))>]
  (DP[SEM=<\Q.exists n.(exists c.(vowel(c) & char(n,c)) & Q(n))>]
    (Det[SEM=<\P Q.exists n.(P(n) & Q(n))>] some)
    (N[SEM=<\n.exists c.(vowel(c) & char(n,c))>] vowel))
  (VP[SEM=<\n.exists c.(capital(n) & char(n,c))>]
    is
    (A[SEM=<\n.exists c.(capital(n) & char(n,c))>] capitalized)))
```

```
In [ ]: tree.draw()
```



```
In [ ]: t3=next(g3.parse(s3_split))
f3 = t3.label()['SEM']
print(f3)

exists n.(exists c.(vowel(c) & char(n,c)) & exists c.(capital(n) & char(n,c)
))
```

```
In [ ]: e3 = [('emmmu',False),('aEkjd',True),('mAEOWr',True),('aeiou',False)]
words = [e[0] for e in e3]
truths = [e[1] for e in e3]
vals = [nltk.Valuation.fromstring(to_model_str(w, [get_vowel, get_capital]))
assignments = [nltk.Assignment(val.domain) for val in vals]
for val in vals: emptysets(val)
models = [nltk.Model(val.domain, val) for val in vals]
```

```
In [ ]: print(f'{s3}\n-----')
for w, a, m in zip(words, assignments, models):
    print(f'{w}\n{m.evaluate(str(f3),a,trace=None)}\n-----')
```

```

some vowel is capitalized
-----
emmmu
False
-----
aEkjd
True
-----
mAE0wr
True
-----
aeiou
False
-----

```

+ The answer is correct! aEkjd & mAE0wr has capitalized vowel(s)

1. no glide is capitalized

The glides are "y" and "w". Define "no", with the same semantic type as "every" and "some". Include a constant 'glide' in the valuations.

Add grammar

```

Det[SEM=<\P Q.all n.(P(n) -> -Q(n))>] -> 'no'
N[SEM=<\n.exists c.(glide(c) & char(n,c))>] -> 'glide'

```

Add in function

```

get_glide = lambda w: f'glide => {set(re.findall(r"[YWyw]",
w))}'.lower()

```

```

In [ ]: g4 = nltk.load_parser('h2.fcfig', trace=0,cache=False)
s4 = 'no glide is capitalized'
s4_split = s4.split()
for tree in g4.parse(s4_split): print(tree)

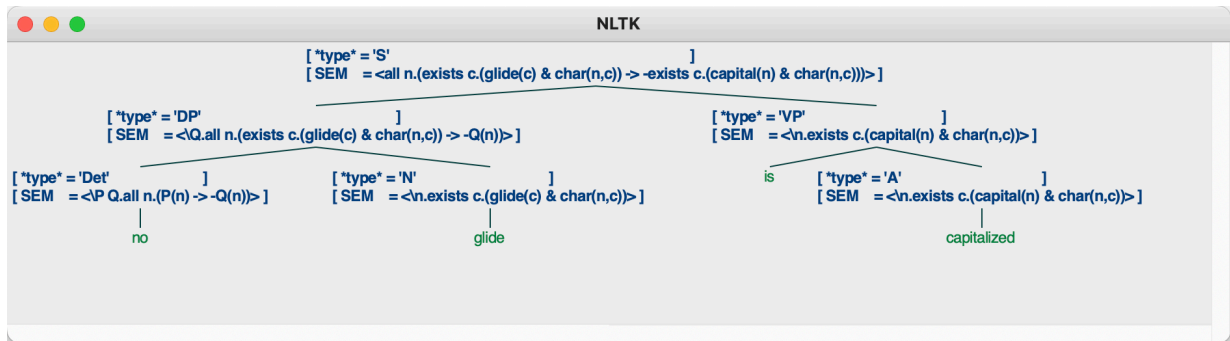
(S[SEM=<all n.(exists c.(glide(c) & char(n,c)) -> -exists c.(capital(n) & char(n,c)))>]
  (DP[SEM=<\Q.all n.(exists c.(glide(c) & char(n,c)) -> -Q(n))>]
    (Det[SEM=<\P Q.all n.(P(n) -> -Q(n))>] no)
    (N[SEM=<\n.exists c.(glide(c) & char(n,c))>] glide))
  (VP[SEM=<\n.exists c.(capital(n) & char(n,c))>]
    is
    (A[SEM=<\n.exists c.(capital(n) & char(n,c))>] capitalized)))

```

```

In [ ]: tree.draw()

```



```
In [ ]: t4=next(g4.parse(s4_split))
f4 = t4.label()['SEM']
print(f4)
```

```
all n.(exists c.(glide(c) & char(n,c)) -> -exists c.(capital(n) & char(n,c))
)
```

```
In [ ]: e4 = [( 'emwywu',True), ('aEWWWjd',False), ('mAEOWr',True), ('aeYWou',False)]
words = [e[0] for e in e4]
truths = [e[1] for e in e4]
vals = [nltk.Valuation.fromstring(to_model_str(w, [get_vowel, get_capital, get_initial]),
assignments = [nltk.Assignment(val.domain) for val in vals]
for val in vals: emptysets(val)
models = [nltk.Model(val.domain, val) for val in vals]
```

```
In [ ]: print(f'{s4}\n-----')
for w, a, m in zip(words, assignments, models):
    print(f'{w}\n{m.evaluate(str(f4),a,trace=None)}\n-----')
```

```
no glide is capitalized
```

```
-----
emwywu
True
-----
aEWWWjd
False
-----
mAEOWr
True
-----
aeYWou
False
-----
```

+ Correct!

1. letter one is initial and is a consonant

Define "initial", meaning 'is at the start of the word' in terms of the available primitives.

Add grammar

```

S[SEM=<?vp1(?subj) & ?vp2(?subj)>] -> NP[SEM=?subj] VP[SEM=?
vp1] 'and' VP[SEM=?vp2]
VP[SEM=?P] -> 'is' 'a' N[SEM=?P]
A[SEM=<\n. (n = 1) >] -> 'initial'

```

```

In [ ]: g5 = nltk.load_parser('h2.fcfig', trace=0,cache=False)
s5 = 'letter one is initial and is a consonant '
s5_split = s5.split()
for tree in g5.parse(s5_split): print(tree)

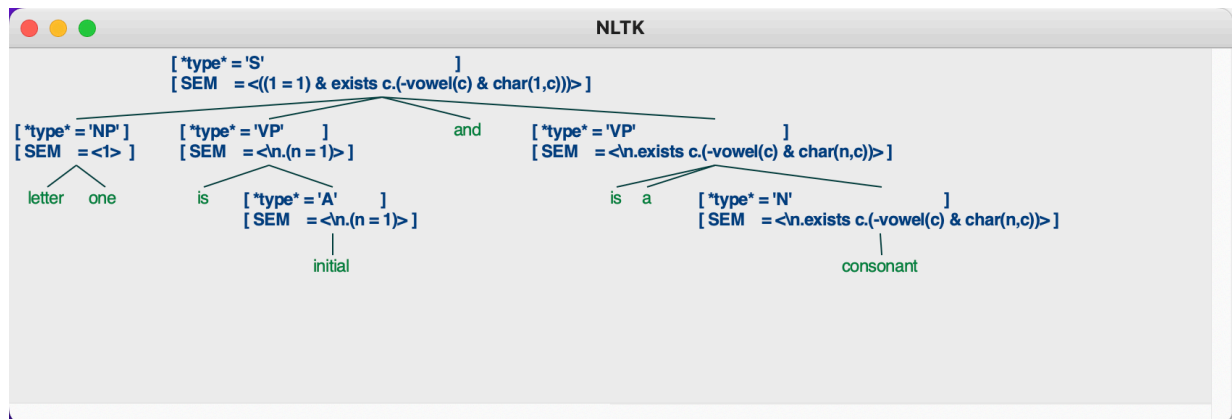
(S[SEM=<((1 = 1) & exists c.(-vowel(c) & char(1,c)))>]
  (NP[SEM=<1>] letter one)
  (VP[SEM=<\n.(n = 1)>] is (A[SEM=<\n.(n = 1)>] initial))
  and
  (VP[SEM=<\n.exists c.(-vowel(c) & char(n,c))>]
    is
    a
    (N[SEM=<\n.exists c.(-vowel(c) & char(n,c))>] consonant)))

```

```

In [ ]: tree.draw()

```



```

In [ ]: t5=next(g5.parse(s5_split))
f5 = t5.label()['SEM']
print(f5)

((1 = 1) & exists c.(-vowel(c) & char(1,c)))

```

```

In [ ]: e5 = [('emwywu',True),('aEWWWjd',False),('mAEOWr',True),('aeYWou',False)]
words = [e[0] for e in e5]
truths = [e[1] for e in e5]
vals = [nltk.Valuation.fromstring(to_model_str(w, [get_vowel])) for w in words]
assignments = [nltk.Assignment(val.domain) for val in vals]
for val in vals: emptysets(val)
models = [nltk.Model(val.domain, val) for val in vals]

```

```
In [ ]: print(f'{s5}\n-----')
        for w, a, m in zip(words, assignments, models):
            print(f'{w}\n{m.evaluate(str(f5),a,trace=0)}\n-----')
```

letter one is initial and is a consonant

emwywu

False

aEWWjd

False

mAEOwr

True

aeYWou

False

+ Correct!