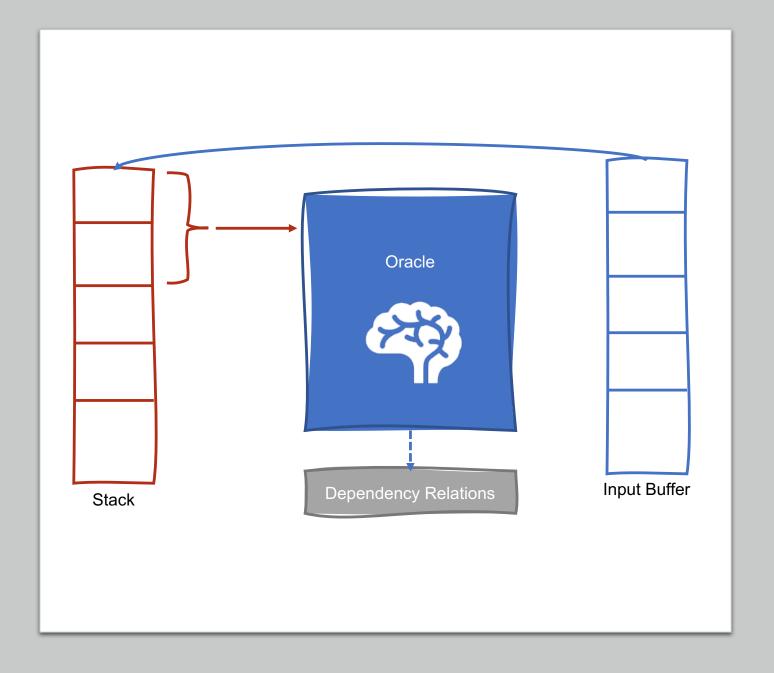
## Transition-Based Dependency Parsing

Natalie Parde UIC CS 421

## Transition-based Dependency Parsing

- Earliest transition-based approach: shift-reduce parsing
  - Input tokens are successively shifted onto a stack
  - The two top elements of the stack are matched against a set of possible relations provided by some knowledge source
  - When a match is found, a head-dependent relation between the matched elements is asserted
- Goal is to find a final parse that accounts for all words



### Transition-based Parsing

- We can build upon shift-reduce parsing by defining a set of transition operators to guide the parser's decisions
- Transition operators work by producing new configurations:
  - Stack
  - Input buffer of words
  - Set of relations representing a dependency tree

#### Transitionbased Parsing

#### Initial configuration:

- Stack contains the ROOT node
- Word list is initialized with all words in the sentence, in order
- Empty set of relations represents the parse

#### Final configuration:

- Stack should be empty
- Word list should be empty
- Set of relations represents the parse

#### **Operators**

- The operators used in transitionbased parsing then perform the following tasks:
  - Assign the current word as the head of some other word that has already been seen
  - Assign some other word that has already been seen as the head of the current word
  - Do nothing with the current word

#### **Operators**

- More formally, these operators are defined as:
  - LeftArc: Asserts a head-dependent relation between the word at the top of the stack and the word directly beneath it (the second word), and removes the second word from the stack
    - Cannot be applied when ROOT is the second element in the stack
    - Requires two elements on the stack
  - RightArc: Asserts a head-dependent relation between the second word and the word at the top of the stack, and removes the word at the top of the stack
    - Requires two elements on the stack
  - Shift: Removes a word from the front of the input buffer and pushes it onto the stack
- These operators implement the arc standard approach to transition-based parsing

# Arc Standard Approach to Transitionbased Parsing

#### Notable characteristics:

- Transition operators only assert relations between elements at the top of the stack
- Once an element has been assigned its head, it is removed from the stack
  - Not available for further processing!

#### Benefits:

- Reasonably effective
- Simple to implement

### Formal Algorithm: Arc Standard Approach

```
state ← {[root], [words], []}
while state not final:
    # Choose which transition operator to apply
    transition ← oracle(state)

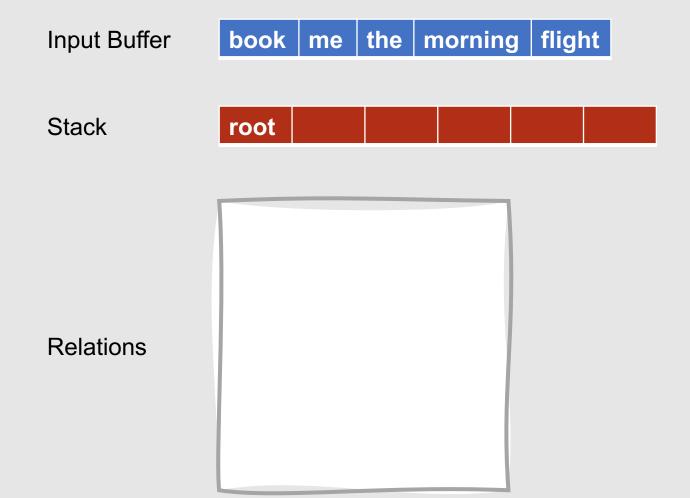
# Apply the operator and create a new state
    state ← apply(transition, state)
```

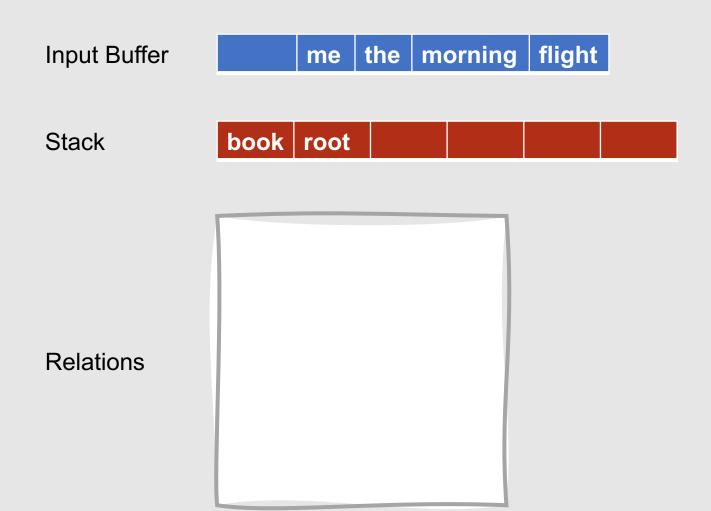
## When does the process end?

- When all words in the sentence have been consumed
- When the ROOT node is the only element remaining on the stack

## Is this another example of dynamic programming?

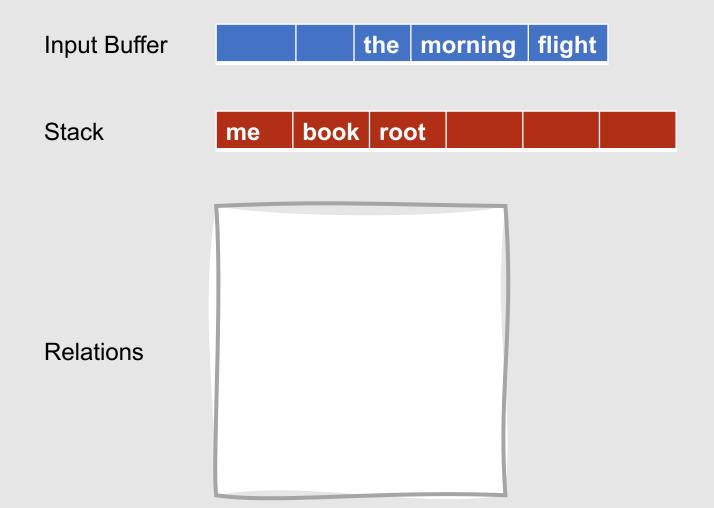
- No! 😺
- The arc standard approach is a greedy algorithm
  - Oracle provides a single choice at each step
  - Parser proceeds with that choice
    - No other options explored
    - No backtracking
  - Single parse returned at the end





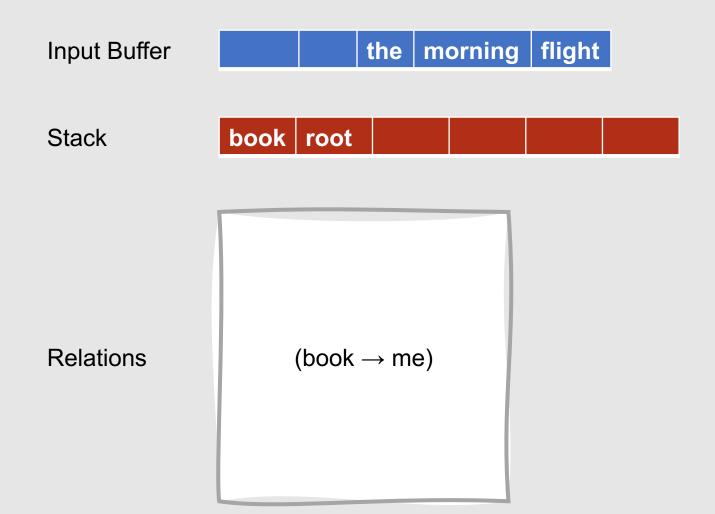
Only one item in the stack!

Shift **book** from the input buffer to the stack



Valid options: Shift, RightArc
Oracle selects Shift

Shift **me** from the input buffer to the stack



Valid options: Shift, RightArc, LeftArc

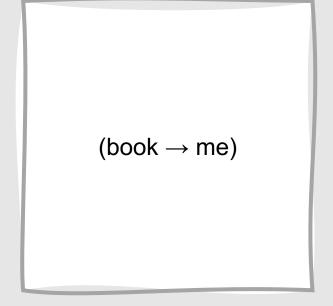
Oracle selects RightArc

Remove **me** from the stack

Add relation (book → me) to the set of relations



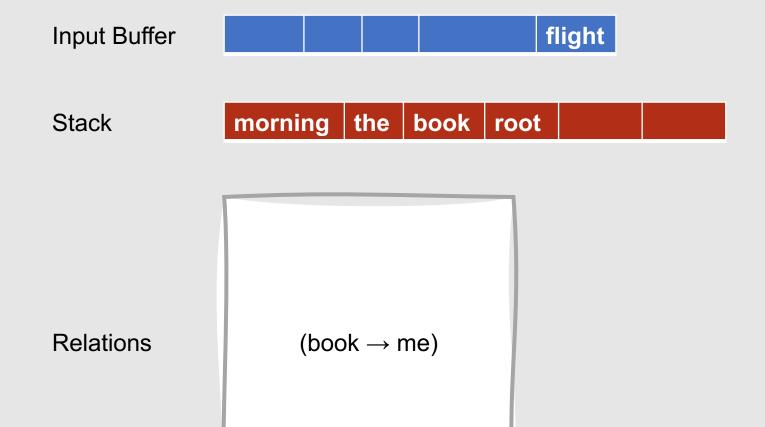
Relations



Valid options: Shift, RightArc

Oracle selects Shift

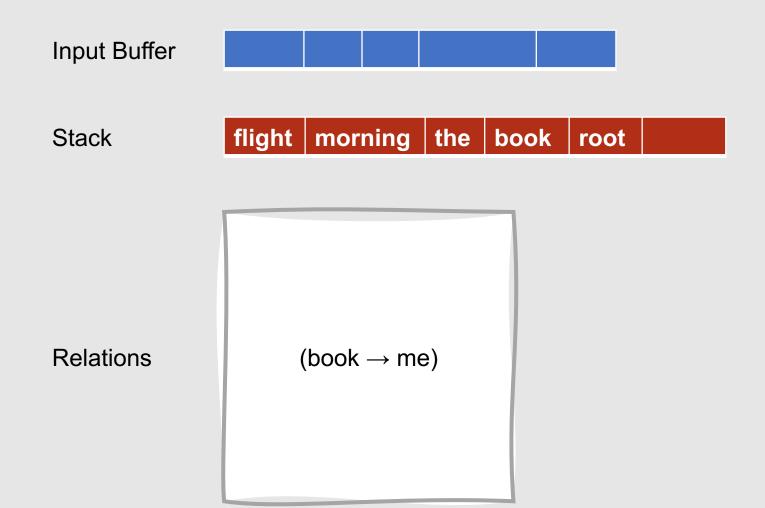
Shift the from the input buffer to the stack



Valid options: Shift, RightArc, LeftArc Oracle selects Shift

Shift morning from the input

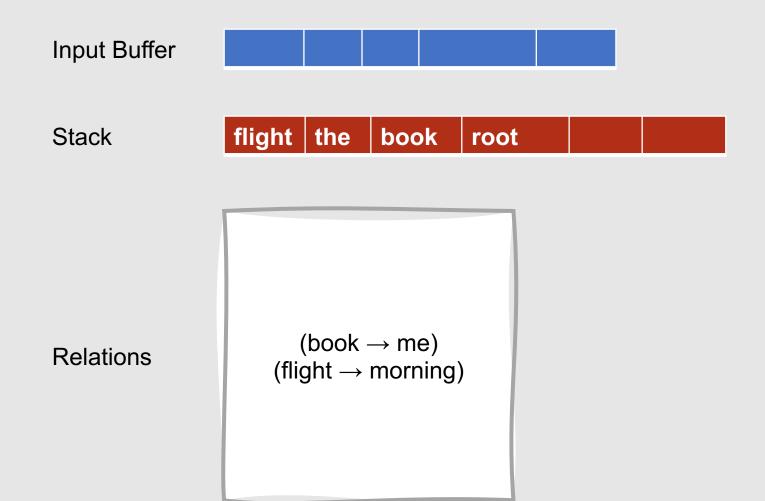
buffer to the stack



Valid options: Shift, RightArc, LeftArc Oracle selects Shift

Shift **flight** from the inn

Shift **flight** from the input buffer to the stack



Valid options: RightArc, LeftArc

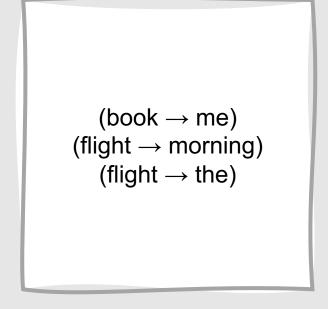
Oracle selects LeftArc

Remove **morning** from the stack

Add relation (flight → morning) to the set of relations



Relations

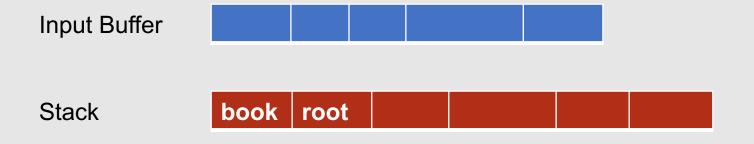


Valid options: RightArc, LeftArc

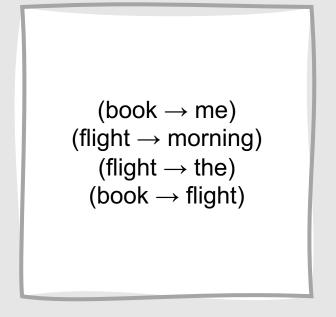
Oracle selects LeftArc

Remove the from the stack

Add relation (flight → the) to the set of relations



Relations

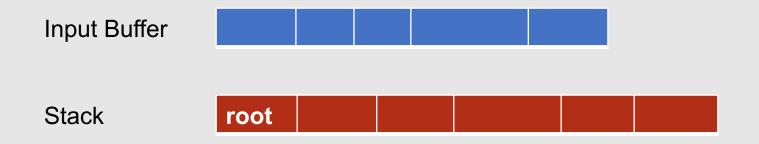


Valid options: RightArc, LeftArc

Oracle selects RightArc

Remove **flight** from the stack

Add relation (book → flight) to the set of relations



Relations

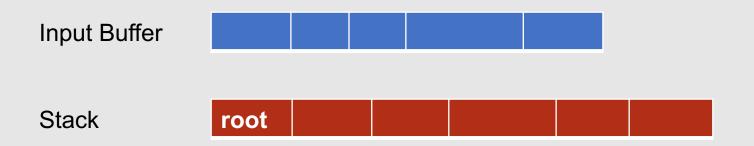
```
(book → me)
(flight → morning)
(flight → the)
(book → flight)
(root → book)
```

Valid options: RightArc

Oracle selects RightArc

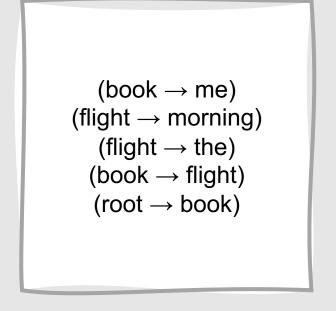
Remove **book** from the stack

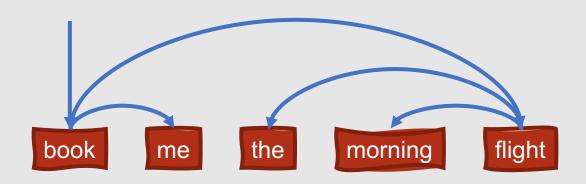
Add relation (root → book) to the set of relations



Valid options: None
State is final

Relations





## A few things worth noting....

- We assumed in the previous example that our oracle was always correct ...this is not necessarily (or perhaps not even likely) the case!
  - Incorrect choices lead to incorrect parses since the algorithm cannot perform any backtracking
- Alternate sequences may also lead to equally valid parses

### How do we get actual dependency labels?

- Parameterize LeftArc and RightArc
  - LeftArc(nsubj), RightArc(obj), etc.
- Of course, this makes the oracle's job more difficult (much larger set of operators from which to choose!)

```
\begin{array}{c} \text{iobj(book} \rightarrow \text{me)} \\ \text{compound(flight} \rightarrow \text{morning)} \\ \text{det(flight} \rightarrow \text{the)} \\ \text{obj(book} \rightarrow \text{flight)} \\ \text{root(root} \rightarrow \text{book)} \end{array}
```

### How does the oracle know what to choose?

- State of the art systems use supervised machine learning for this task
- This requires a training set of configurations labeled with correct transition operators
- The person designing the system needs to decide what types of features should be extracted from these configurations to best train the oracle (a machine learning model)
- The oracle will then learn which transitions to predict for previously-unseen configurations based on the extracted features and associated labels for configurations in the training set



What types of machine learning models are used as oracles?

#### Commonly:

- Logistic regression
- Support vector machines

#### Recently:

Neural networks