

# Language and Computers (Ling 384)

Topic 4: Writer's Aids (Spelling and Grammar Correction)

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\*This course was created by Markus Dickinson, Detmar Meurers and Chris Brew.

## 1 Introduction

### Who cares about spelling?

Aoccdmrig to a rscheearch at Cmabrigde Uinervtisy, it deosn't mtttaer in waht oredr the ltteers in a wrod are, the olny iprmoentn tihng is taht the frist and lsat ltteer be at the rghit pclae. The rset can be a toatl mses and you can siltll raed it wouthit porbelm. Tihis is bcuseae the huamn mnid deos not raed ervey lteer by istlef, but the wrod as a wlohe.

(See <http://www.mrc-cbu.cam.ac.uk/personal/matt.davis/Cmabrigde/> for the story behind this supposed research report.)

A dootcr has aimtted the magltheuansr of a tageene ceacnr pintaet who deid aetfr a hatospil durg blendur.

### Why people care about spelling

- People want to appear to be educated.
- Misspellings can cause misunderstandings and real-life problems:
  - For example:
    - \* Did you see her god yesterday? It's a big golden retriever.
    - \* This will be a fee [free] concert.
  - 1991 Bell Atlantic & Pacific Bell telephone network outages were partly caused by a typographical error:  
A *6* in a line of computer code was supposed to be a *D*. "That one error caused the equipment and software to fail under an avalanche of computer-generated messages." (Wall Street Journal, Nov. 25, 1991)

### Why people care about spelling (cont.)

- Standard spelling makes it easy to organize words and text:
  - e.g., Without standard spelling, how would you look up things in a lexicon or thesaurus?
  - e.g., Optical character recognition software can use knowledge about standard spelling to recognize scanned words even for barely legible input.
- Standard spelling makes it possible to provide a single text, which is accessible to a wide range of readers (different backgrounds, speaking different dialects, etc.).
- Using standard spelling is associated with being well-educated, i.e., is used to make a good impression in social interaction.

### How are spell checkers used?

- **interactive spelling checkers** = spell checker detects errors as you type.
  - It may or may not make suggestions for correction.
  - Requires a “real-time” response (i.e., must be fast)
  - It is up to the human to decide if the spell checker is right or wrong.
  - If there are a list of choices, we may not require 100% accuracy in the corrected word
- **automatic spelling correctors** = spell checker runs on a whole document, finds errors, and corrects them
  - A much more difficult task.
  - A human may or may not proofread the results later.

### Detection vs. Correction

- There are two distinct tasks:
  - **error detection** = simply find the misspelled words
  - **error correction** = correct the misspelled words
- e.g., It might be easy to tell that *ater* is a misspelled word, but what is the correct word? *water?* *later?* *after?*

⇒ Depends on what we want to do with our results as to what we want to do.

Note, though, that detection is a prerequisite for correction.

## 2 Error causes

### What causes errors?

- Keyboard mistypings
- Phonetic errors
- Knowledge problems

### 2.1 Keyboard mistypings

#### Keyboard mistypings

Space bar issues

- **run-on** errors = two separate words become one
  - e.g., *the fuzz* becomes *thefuzz*

- **split** errors = one word becomes two separate words
  - e.g., *equalization* becomes *equali zation*

Note that the resulting items might still be words!

- e.g., *a tollway* becomes *atoll way*

### Keyboard mistypings (cont.)

Keyboard proximity

- e.g., *Jack* becomes *Hack* since *h* and *j* are next to each other on a typical American keyboard

Physical similarity

- similarity of shape, e.g., mistaking two physically similar letters when typing up something handwritten
  - e.g., *tight* for *fight*

### 2.2 Phonetic errors

#### Phonetic errors

**phonetic errors** = errors based on the sounds of a language (not necessarily on the letters)

- **homophones** = two words which sound the same
  - e.g., *red/read* (past tense), *cite/site/sight*, *they're/their/there*
- **Spoonerisms** = switching two letters/sounds around
  - e.g., *Pleating and humming*

#### Phonetic errors (cont.)

- letter substitution: replacing a letter (or sequence of letters) with a similar-sounding one
  - e.g., *John cracked his nuckles.* instead of *John cracked his knuckles.*
  - e.g., *I study sikologee.*
- word replacement: replacing one word with some similar-sounding word
  - e.g., *John battled me on the back.* instead of *John patted me on the back.*

### More examples for phonetic errors

- (1) a. death in Venice  
b. deaf in Venice
- (2) a. give them an ice bucket  
b. give them a nice bucket
- (3) a. the stuffy nose  
b. the stuff he knows
- (4) a. the biggest hurdle  
b. the biggest turtle
- (5) a. a Coke and a danish  
b. a coconut danish

### 2.3 Knowledge problems

#### Knowledge problems

- not knowing a word and guessing its spelling (can be phonetic)
  - e.g., *sientist*
- not knowing a rule and guessing it
  - e.g., Do we double a consonant for *ing* words? *jog* → *joging* *joke* → *jokking*

## 3 Difficult issues

### What makes spelling correction difficult?

- **Tokenization:** What is a word?
- **Inflection:** How are some words related?
- **Productivity of language:** How many words are there?

How we handle these issues determines how we build a dictionary.

### 3.1 Tokenization

#### Tokenization

Intuitively a “word” is simply whatever is between two spaces, but this is not always so clear.

- contractions = two words combined into one
  - e.g., *can't*, *he's*, *John's [car]* (vs. *his car*)

- multi-token words = (arguably) a single word with a space in it
  - e.g., *New York*, *in spite of*, *deja vu*
- hyphens (note: can be ambiguous if a hyphen ends a line)
  - Some are always a single word: *e-mail*, *co-operate*
  - Others are two words combined into one: *Columbus-based*, *sound-change*
- Abbreviations: may stand for multiple words
  - e.g., *etc.* = *et cetera*, *ATM* = *Automated Teller Machine*

### 3.2 Inflection

#### Inflection

- A word in English may appear in various guises due to word **inflections** = word endings which are fairly systematic for a given part of speech
  - plural noun ending: *the boy* + *s* → *the boys*
  - past tense verb ending: *walk* + *ed* → *walked*
- This can make spell-checking hard:
  - There are exceptions to the rules: *mans*, *runned*
  - There are words which look like they have a given ending, but they don't: *Hans*, *deed*

### 3.3 Productivity

#### Productivity

- part of speech change: nouns can be verbified
  - *tabled* is a new verb coined after the noun *table*
  - Calvin: I like to verb words. Hobbes: What? Calvin: I take nouns and adjectives and use them as verbs. Remember when “access” was a thing? Now, it's something you do. It got verbed. Verbing weirds language. Hobbes: Maybe we can eventually make language a complete impediment to understanding.
- morphological productivity: prefixes and suffixes can be added
  - e.g., I can speak of *un-email-able* for someone who you can't reach by email.
- words entering and exiting the lexicon, e.g.:
  - *thou*, or *spleet 'split'* (*Hamlet III.2.10*) are on their way out
  - *SARS* and *wifi* (or *wi-fi*) have entered recently

### Techniques used for spell checking

- Non-word error detection
- Isolated-word error correction
- Context-dependent word error detection and correction → grammar correction.

## 4 Non-word error detection

### Non-word error detection

- **non-word error detection** is essentially the same thing as **word recognition** = splitting up “words” into true words and non-words.
- How is non-word error detection done?
  - using a dictionary (construction and lookup)
  - n-gram analysis

### 4.1 Dictionaries

#### Dictionaries

Intuition:

- Have a complete list of words and check the input words against this list.
- If it's not in the dictionary, it's not a word.

Two aspects:

- **Dictionary construction** = build the dictionary (what do you put in it?)
- **Dictionary lookup** = lookup a potential word in the dictionary (how do you do this quickly?)

#### Dictionary construction

- Do we include inflected words? i.e., words with prefixes and suffixes already attached.
  - Pro: lookup can be faster
  - Con: takes much more space, doesn't account for new formations (e.g. *google* → *googled*)
- Want the dictionary to have only the word relevant for the user → **domain-specificity**
  - e.g., For most people *memoize* is a misspelled word, but in computer science this is a technical term and spelled correctly.

- Foreign words, hyphenations, derived words, proper nouns, and new words will always be problems for dictionaries since we cannot predict these words until humans have made them words.
- Dictionary should probably be dialectally consistent.
  - e.g., include only *color* or *colour* but not both

### Dictionary lookup

Several issues arise when trying to look up a word:

- Have to make lookup fast by using efficient lookup techniques, such as a hash table
- Have to strip off prefixes and suffixes if the word isn't an entry by itself.
  - *running* → *run*
  - *antireligious* → *religious* both

### 4.2 N-gram analysis

#### N-gram analysis

- An **n-gram** here is a string of  $n$  letters.

<i>a</i>	1-gram (unigram)
<i>at</i>	2-gram (bigram)
<i>ate</i>	3-gram (trigram)
<i>late</i>	4-gram
⋮	⋮

- We can use this n-gram information to define what the possible strings in a language are.
  - e.g., *po* is a possible English string, whereas *kvt* is not.

#### How do we store and use n-gram information?

- Store the number of times an n-gram appears (like in Language Identification). But, maybe we just want to know if an n-gram is possible.
- We could have a list of possible and impossible n-grams (1 = possible, 0 = impossible):

<i>po</i>	1
<i>kvt</i>	0
<i>police</i>	1
<i>asdf</i>	0

- Any word which has a 0 for any substring is a misspelled word.

- Problems with such an approach:
  - Information is repeated (*po* is in *police*)
  - Requires a lot of computer storage space
  - Inefficient (slow) when looking up every string

### Bigram array

- Instead, we can define a **bigram array** = information stored in a tabular fashion.
- An example, for the letters *k, l, m*, with examples in parentheses

	...	k	l	m	...
⋮					
k		0	1 ( <i>tackle</i> )	1 ( <i>Hackman</i> )	
l		1 ( <i>elk</i> )	1 ( <i>hello</i> )	1 ( <i>alms</i> )	
m		0	0	1 ( <i>hammer</i> )	
⋮					

- The first letter of the bigram is given by the vertical letters (i.e., down the side), the second by the horizontal ones (i.e., across the top).
- This is a **non-positional bigram array** = the array 1's and 0's apply for a string found anywhere within a word (beginning, 4th character, ending, etc.).

### Positional bigram array

- To store information specific to the beginning, the end, or some other position in a word, we can use a **positional bigram array** = the array only applies for a given position in a word.
- Here's the same array as before, but now only applied to word endings:

	...	k	l	m	...
⋮					
k		0	0	0	
l		1 ( <i>elk</i> )	1 ( <i>hall</i> )	1 ( <i>elm</i> )	
m		0	0	0	
⋮					

## 5 Isolated-word error correction

### Isolated-word error correction

- Having discussed how errors can be detected, we want to know how to correct these misspelled words:
  - The most common method is **isolated-word error correction** = correcting words without taking context into account.

- Note: This technique can only handle errors that result in non-words.

- Knowledge about what is a typical error helps in finding correct word.

### Knowledge about typical errors

- word length effects: most misspellings are within two characters in length of original
  - When searching for the correct spelling, we do not usually need to look at words with greater length differences.
- first-position error effects: the first letter of a word is rarely erroneous
  - When searching for the correct spelling, the process is sped up by being able to look only at words with the same first letter.

### Isolated-word error correction methods

- Many different methods are used; we will briefly look at four methods:
  - rule-based methods
  - similarity key techniques
  - minimum edit distance
  - probabilistic methods
- The methods play a role in one of the three basic steps:
  1. Detection of an error (discussed above)
  2. Generation of candidate corrections
    - rule-based methods
    - similarity key techniques
  3. Ranking of candidate corrections
    - probabilistic methods
    - minimum edit distance

### 5.1 Rule-based methods

#### Rule-based methods

One can generate correct spellings by writing rules:

- Common misspelling rewritten as correct word:
  - e.g., *hte* → *the*
- Rules

- based on inflections:
  - \* e.g.,  $V+CC+ing \rightarrow V+CC+ing$  (where V = vowel and C = consonant)
- based on other common spelling errors (such as keyboard effects or common transpositions):
  - \* e.g.,  $CsC \rightarrow CaC$
  - \* e.g.,  $Cie \rightarrow Cei$

## 5.2 Similarity key techniques

### Similarity key techniques

- Problem: How can we find a list of possible corrections?
- Solution: Store words in different boxes in a way that puts the similar words together.
- Example:
  1. Start by storing words by their first letter (first letter effect),
    - e.g., *punc* starts with the code P.
  2. Then assign numbers to each letter
    - e.g., 0 for vowels, 1 for *b, p, f, v* (all bilabials), and so forth, e.g., *punc*  $\rightarrow$  P052
  3. Then throw out all zeros and repeated letters,
    - e.g., P052  $\rightarrow$  P52.
  4. Look for real words within the same box,
    - e.g., *punk* is also in the P52 box.

## 5.3 Probabilistic methods

### Probabilistic methods

Two main probabilities are taken into account:

- **transition probabilities** = probability (chance) of going from one letter to the next.
  - e.g., What is the chance that *a* will follow *p* in English? That *u* will follow *q*?
- **confusion probabilities** = probability of one letter being mistaken (substituted) for another (can be derived from a confusion matrix)
  - e.g., What is the chance that *q* is confused with *p*?

Useful to combine probabilistic techniques with dictionary methods

## Confusion probabilities

- For the various reasons discussed above (keyboard layout, phonetic similarity, etc.) people type other letters than the ones they intended.
- It is impossible to fully investigate all possible error causes and how they interact, but we can learn from watching how often people make errors and where.
- One way of doing so is to build a **confusion matrix** = a table indicating how often one letter is mistyped for another

		correct				
		...	r	s	t	...
typed	r		n/a	12	22	
	s		14	n/a	15	
	t		11	37	n/a	
	⋮					

(cf. Kernighan et al 1999)

## 5.4 Minimum edit distance

### How is a mistyped word related to the intended?

Types of errors

- **insertion** = a letter is added to a word
- **deletion** = a letter is deleted from a word
- **substitution** = a letter is put in place of another one
- **transposition** = two adjacent letters are switched

Note that the first two alter the length of the word, whereas the second two maintain the same length.

General properties

- **single-error misspellings** = only one instance of an error
- **multi-error misspellings** = multiple instances of errors (harder to identify)

### Minimum edit distance

- In order to rank possible spelling corrections, it can be useful to calculate the **minimum edit distance** = minimum number of operations it would take to convert one word into another.
- For example, we can take the following five steps to convert *junk* to *haiku*:

1. *junk* → *juk* (deletion)
2. *juk* → *huk* (substitution)
3. *huk* → *hku* (transposition)
4. *hku* → *hiku* (insertion)
5. *hiku* → *haiku* (insertion)

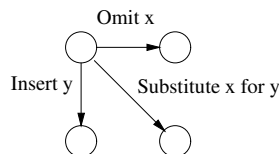
- But is this the minimal number of steps needed?

### Computing edit distances

- To be able to compute the edit distance of two words at all, we need to ensure there is a finite number of steps.
- This can be accomplished by
  - requiring that letters cannot be changed back and forth a potentially infinite number of times, i.e., we
  - limit the number of changes to the size of the material we are presented with, the two words.
- Idea: Never deal with a character in either word more than once.
- Result:
  - In the worst case, we delete each character in the first word and then insert each character of the second word.
  - The worst case edit distance for two words is  $length(word1) + length(word2)$

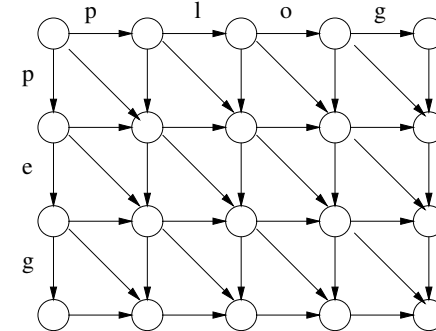
### Computing edit distances

- To calculate minimum edit distance, we set up a **directed, acyclic graph**, a set of nodes (circles) and arcs (arrows).
- Horizontal arcs correspond to deletions, vertical arcs correspond to insertions, and diagonal arcs correspond to substitutions (and a letter can be “substituted” for itself).



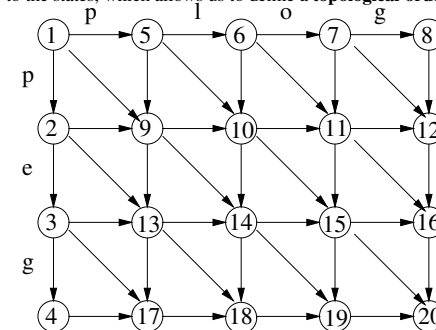
### Computing edit distances

- Say, the user types in *plog*.
- We want to calculate how far away *peg* is (one of the possible corrections). In other words, we want to calculate the minimum edit distance (or minimum edit cost) from *plog* to *peg*.
- As the first step, we draw the following directed graph:



### Computing edit distances

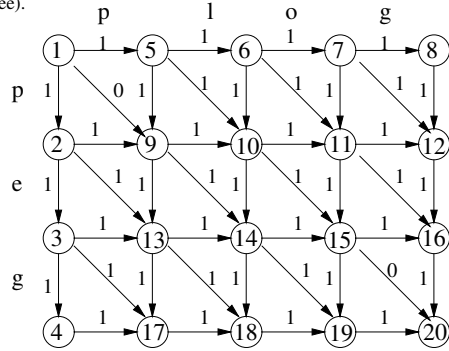
- The graph is **acyclic** = for any given node, it is impossible to return to that node by following the arcs.
- We can add identifiers to the states, which allows us to define a **topological order**:



### Computing edit distances

- We need to add the costs involved to the arcs.

- In the simplest case, the cost of deletion, insertion, and substitution is 1 each (and substitution with the same character is free).



- Instead of assuming the same cost for all operations, in reality one will use different costs, e.g., for the first character or based on the confusion probability.

### Computing edit distances

We want to find the path from the start (1) to the end (20) with the least cost.

- The simple but dumb way of doing it:
  - Follow every path from start (1) to finish (20) and see how many changes we have to make.
  - But this is very inefficient! There are 131 different paths to check.

### Computing edit distances

- The smart way to compute the least cost uses **dynamic programming** = a program designed to make use of results computed earlier
  - We follow the topological ordering.
  - As we go in order, we calculate the least cost for that node:
    - \* We add the cost of an arc to the cost of reaching the node this arc originates from.
    - \* We take the minimum of the costs calculated for all arcs pointing to a node and store it for that node.
  - The key point is that we are storing partial results along the way, instead of recalculating everything, every time we compute a new path.

## 6 Grammar correction

### Context-dependent word correction

**Context-dependent word correction** = correcting words based on the surrounding context.

- This will handle errors which are real words, just not the right one or not in the right form.
- Essentially a fancier name for a **grammar checker** = a mechanism which tells a user if their grammar is wrong.

### Grammar correction—what does it correct?

- Syntactic errors = errors in how words are put together in a sentence: the order or form of words is incorrect, i.e., ungrammatical.
  - e.g., *The study was conducted mainly **be** John Black.*
  - A verb is where a preposition should be.
- **Local** syntactic errors: 1-2 words away
  - e.g., *The **kids** who are most upset by the little totem **is** going home early.*
  - Agreement error between subject *kids* and verb *is*
- **Long-distance** syntactic errors: (roughly) 3 or more words away
  - e.g., *They are leaving in about fifteen **minuets** to go to her house.*
  - ⇒ *minuets* and *minutes* are both plural nouns, but only one makes sense here

### More on grammar correction

- Semantic errors = errors where the sentence structure sounds okay, but it doesn't really mean anything.
  - e.g., *They are leaving in about fifteen **minuets** to go to her house.*
  - ⇒ *minuets* and *minutes* are both plural nouns, but only one makes sense here

There are many different ways in which grammar correctors work, two of which we'll focus on:

- Bigram model (bigrams of words)
- Rule-based model



### Bigram grammar correctors

We could also look at **bigrams**: now we are talking about bigrams of words, i.e., two words appearing next to each other.

- **Question:** Given the previous word, what is the probability of the current word?
  - e.g., given *these*, we have a 5% chance of seeing *reports* and a 0.001% chance of seeing *report* (*these report cards*).
  - Thus, we will change *report* to *reports*
- But there's a major problem: we may hardly ever see *these reports*, so we won't know the probability of that bigram.
- **(Partial) Solution:** use bigrams of **parts of speech**.
  - e.g., What is the probability of a noun given that the previous word was an adjective?

### Rule-based grammar correctors

We can write regular expressions to target specific error patterns. For example:

- *To a certain extend, we have achieved our goal.*
  - Match the pattern *some* or *certain* followed by *extend*, which can be done using the regular expression `some|certain extend`
  - Change the occurrence of *extend* in the pattern to *extent*.
- Naber (2003) uses 56 such rules to build a grammar corrector which works nearly as well as that in commercial products.

## 6.1 Syntax

### Beyond regular expressions

- But what about correcting the following:
  - *A baseball teams were successful.*
- We should change *A* to *The*, but a simple regular expression doesn't work because we don't know where the word *teams* might show up.
  - *A wildly overpaid, horrendous baseball teams were successful.* (Five words later; change needed.)
  - *A player on both my teams was successful.* (Five words later; no change needed.)
- We need to look at how the sentence is constructed in order to build a better rule.

### Syntax

- **Syntax** = the study of the way that sentences are constructed from smaller units.
- There cannot be a "dictionary" for sentences since there is an infinite number of possible sentences:
  - (6) The house is large.
  - (7) John believes that the house is large.
  - (8) Mary says that John believes that the house is large.

There are two basic principles of sentence organization:

- Linear order
- Hierarchical structure (Constituency)

### Linear order

- **Linear order** = the order of words in a sentence.
- A sentence has different meanings based on its linear order.
  - (9) John loves Mary.
  - (10) Mary loves John.
- Languages vary as to what extent this is true, but linear order in general is used as a guiding principle for organizing words into meaningful sentences.
- Simple linear order as such is not sufficient to determine sentence organization though. For example, we can't simply say "The verb is the second word in the sentence."
  - (11) I **eat** at really fancy restaurants.
  - (12) Many executives **eat** at really fancy restaurants.

### Constituency

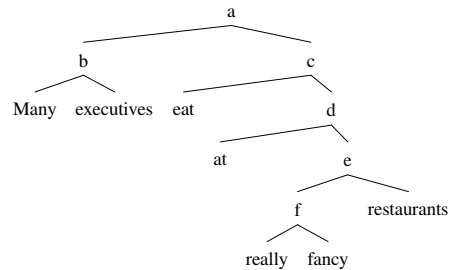
- What are the "meaningful units" of a sentence like *Many executives eat at really fancy restaurants*?
  - Many executives
  - really fancy
  - really fancy restaurants
  - at really fancy restaurants
  - eat at really fancy restaurants
- We refer to these meaningful groupings as **constituents** of a sentence.
- There are many "tests" to determine what a constituent is, but we will not concern ourselves with them here.

## Hierarchical structure

- Constituents can appear within other constituents. We can represent this in a bracket form or in a **syntactic tree**
- Constituents shown through brackets:

[[Many executives] [eat [at [[really fancy] restaurants]]]]

- Constituents displayed as a tree:



## Categories

- We would also like some way to say that *Many executives* and *really fancy restaurants* are the same type of grouping, or constituent, whereas *at really fancy restaurants* seems to be something else.
- For this, we will talk about different **categories**
  - Lexical
  - Phrasal

## Lexical categories

**Lexical categories** are simply word classes, or what you may have heard as **parts of speech**. The main ones are:

- verbs: *eat, drink, sleep, ...*
- nouns: *gas, food, lodging, ...*
- adjectives: *quick, happy, brown, ...*
- adverbs: *quickly, happily, well, westward*
- prepositions: *on, in, at, to, into, of, ...*
- determiners/articles: *a, an, the, this, these, some, much, ...*

## Determining lexical categories

How do we determine which category a word belongs to?

- **Distribution:** Where can these kinds of words appear in a sentence?
  - e.g., Nouns like *mouse* can appear after articles (“determiners”) like *the*, while a verb like *eat* cannot.
- **Morphology:** What kinds of word prefixes/suffixes can a word take?
  - e.g., Verbs like *walk* can take a *ed* ending to mark them as past tense. A noun like *mouse* cannot.

## Closed & Open classes

We can add words to some classes, but not to others. This also seems to correlate with whether a word is “meaningful” or just a **function word** = only meaning comes from its usage in a sentence.

**Open classes:** new words can be easily added:

- verbs
- nouns
- adjectives
- adverbs

**Closed classes:** new words cannot be easily added:

- prepositions
- determiners

## Phrasal categories

What about phrases? Can we assign them categories?

We can also look at their distribution and see which ones behave in the same way.

- The joggers ran through the park.

What other phrases can we put in place of *The joggers*?

## Phrasal categories (cont.)

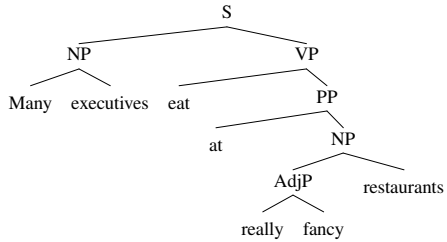
- Susan
- students
- you
- most dogs
- some children

- a huge, lovable bear
- my friends from Brazil
- the people that we interviewed

Since all of these contain nouns, we consider these to be *noun phrases*, abbreviated with NP.

### Building a tree

Other phrases work similarly (S = sentence, VP = verb phrase, PP = prepositional phrase, AdjP = adjective phrase):



### Phrase Structure Rules

- We can give rules for building these phrases. That is, we want a way to say that a determiner and a noun make up a noun phrase, but a verb and an adverb do not.
- **Phrase structure rules** are a way to build larger constituents from smaller ones.
  - e.g.,  $S \rightarrow NP VP$   
This says:
    - \* A sentence (S) constituent is composed of a noun phrase (NP) constituent and a verb phrase (VP) constituent. (hierarchy)
    - \* The NP must precede the VP. (linear order)

### Some other English rules

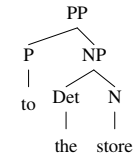
- $NP \rightarrow Det N$  (*the cat, a house, this computer*)
- $NP \rightarrow Det AdjP N$  (*the happy cat, a really happy house*)
  - For phrase structure rules, as shorthand parentheses are used to express that a category is optional.
  - We thus can compactly express the two rules above as one rule:
  - $NP \rightarrow Det (AdjP) N$

– Note that this is different and has nothing to do with the use of parentheses in regular expressions.

- $AdjP \rightarrow (Adv) Adj$  (*really happy*)
- $VP \rightarrow V$  (*laugh, run, eat*)
- $VP \rightarrow V NP$  (*love John, hit the wall, eat cake*)
- $VP \rightarrow V NP NP$  (*give John the ball*)
- $PP \rightarrow P NP$  (*to the store, at John, in a New York minute*)
- $NP \rightarrow NP PP$  (*the cat on the stairs*)

### Phrase Structure Rules and Trees

With every phrase structure rule, you can draw a tree for it.



### Phrase Structure Rules in Practice

Try analyzing these sentences and drawing trees for them, based on the phrase structure rules given above.

- The man in the kitchen drives a truck.
- That dang cat squeezed some fresh orange juice.
- The mouse in the corner by the stairs ate the cheese.

### Properties of Phrase Structure Rules

- **generative** = a schematic strategy that describes a set of sentences completely.
- potentially **(structurally) ambiguous** = have more than one analysis
  - (13) We need more intelligent leaders.
  - (14) Paraphrases:
    - We need leaders who are more intelligent.
    - Intelligent leaders? We need more of them!
- **hierarchical** = categories have internal structure; they aren't just linearly ordered.
- **recursive** = property allowing for a rule to be reapplied (within its hierarchical structure).  
e.g.,  $NP \rightarrow NP PP PP \rightarrow P NP$   
The property of recursion means that the set of potential sentences in a language is **infinite**.

### Context-free grammars

A **context-free grammar** (CFG) is essentially a collection of phrase structure rules.

- It specifies that each rule must have:
  - a left-hand side (LHS): a single **non-terminal** element = (phrasal and lexical) categories
  - a right-hand side (RHS): a mixture of non-terminal and terminal elements  
**terminal** elements = actual words
- A CFG tries to capture a natural language completely.

Why “context-free”? Because these rules make no reference to any context surrounding them. i.e. you can’t say “PP → P NP” *when* there is a verb phrase (VP) to the left.

### 6.2 Computing with Syntax

#### Parsing

So, using these phrase structure (context-free) rules, **parse** a sentence = assign a structure to a sentence. Do you parse top-down or bottom-up (or a mixture)?

- **top-down**: build a tree by starting at the top (i.e.  $S \rightarrow NP VP$ ) and working down the tree.
- **bottom-up**: build a tree by starting with the words at the bottom and working up to the top.

There are many, many parsing techniques out there.

### 6.3 Grammar correction rules

#### Writing grammar correction rules

So, with context-free grammars, we can now write some correction rules, which we will just sketch here.

- *A baseball teams were successful.*

A followed by PLURAL NP: change  $A \rightarrow The$

- *John at the taco.*

The structure of this sentence is NP PP, but that doesn’t make up a whole sentence. We need a verb somewhere.

## 7 Caveat emptor

#### Is this really how spell checkers work?

As far as we know, yes, but:

- Many spell checkers are proprietary and the way they work is kept secret; we don’t know how they work exactly, which hampers research and thereby progress.
- Others, such as aspell and ispell, are **open source** spell checkers, meaning that anyone can

- contribute to their further development, and
- see how they work, which makes it possible to understand exactly what they will and what they won’t catch.

(cf. <http://aspell.sourceforge.net/> and <http://fmg-www.cs.ucla.edu/fmg-members/geoff/ispell.html>)

### Dangers of spelling and grammar correction

- The more we depend on spelling correctors, the less we try to correct things on our own. But spell checkers are not 100%
- A study at the University of Pittsburgh found that students made **more** errors when using a spell checker!

	high SAT scores	low SAT scores
use checker	16 errors	17 errors
no checker	5 errors	12.3 errors

(cf., <http://www.wired.com/news/business/0,1367,58058,00.html>)

### A Poem on the Dangers of Spell Checkers

Michael Livingston (<http://www.courses.rochester.edu/livingston/guide/phonix.html>)

Eye halve a spelling chequer  
 It came with my pea sea.  
 It plainly marques four my revue  
 Miss steaks eye kin knot sea.  
 Eye strike a key and type a word  
 And weight four it two say  
 Weather eye am wrong oar write  
 It shows me strait a weigh.  
 As soon as a mist ache is maid  
 It nose bee fore two long  
 And eye can put the error rite  
 Its rare lea ever wrong.  
 Eye have run this poem threw it  
 I am shore your pleased two no  
 Its letter perfect awl the weigh  
 My chequer tolled me sew.

### References

- The discussion is based on Markus Dickinson (to appear). *Writer’s Aids*. In Keith Brown (ed.): *Encyclopedia of Language and Linguistics. Second Edition.* Elsevier.
- A major inspiration for that article and our discussion is Karen Kukich (1992): *Techniques for Automatically Correcting Words in Text*. ACM Computing Surveys, pages 377–439.

- For a discussion of the confusion matrix, cf. Mark D. Kernighan, Kenneth W. Church and William A. Gale (1990). A spelling Correction Program Based on a Noisy Channel Model. In *Proceedings of COLING-90*. pp. 205–210.
- An open-source style/grammar checker is described in Daniel Naber (2003). *A Rule-Based Style and Grammar Checker*. Diploma Thesis, Universität Bielefeld. <http://www.danielnaber.de/language-tool/>