UNIVERSITY OF CAMBRIDGE

# Applying Language Models to Language Learning

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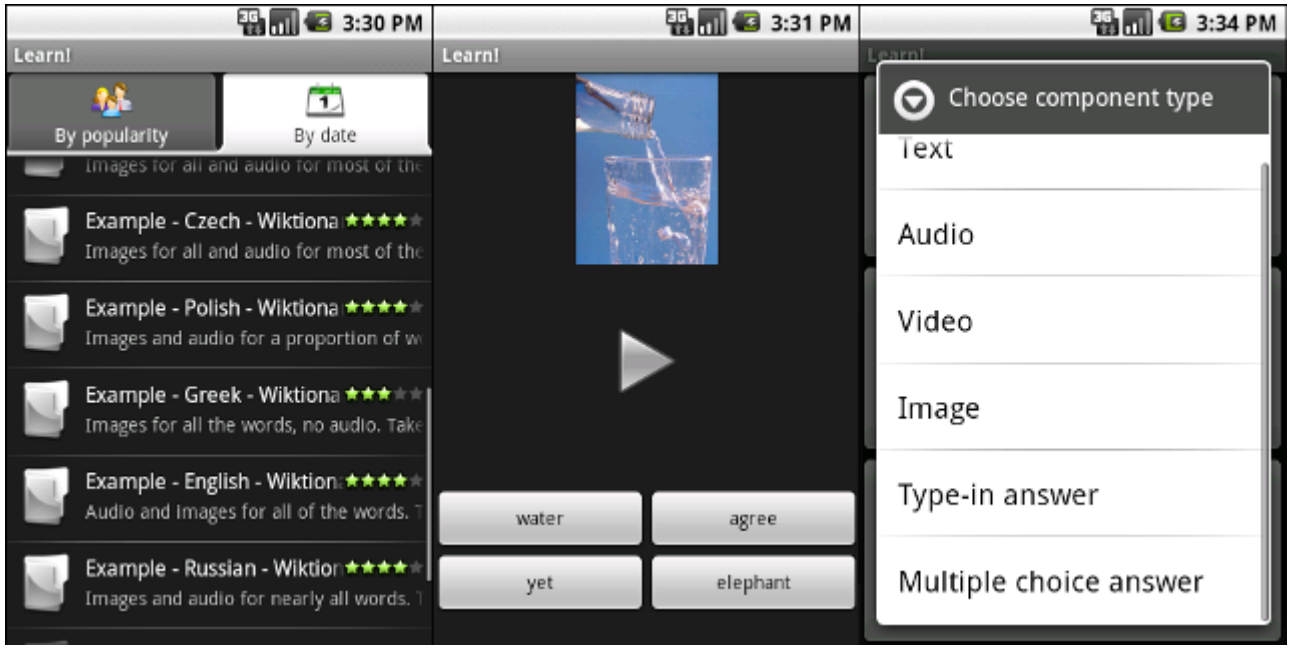
# Digital flash cards!

Popular approach to learning. DTG's own *Learn!* Android app has sets for:

- Basic chemistry
- US Navy core values
- Career development strategies
- ...Languages?

# Language learning through flash cards

Teaching vocab not too hard:

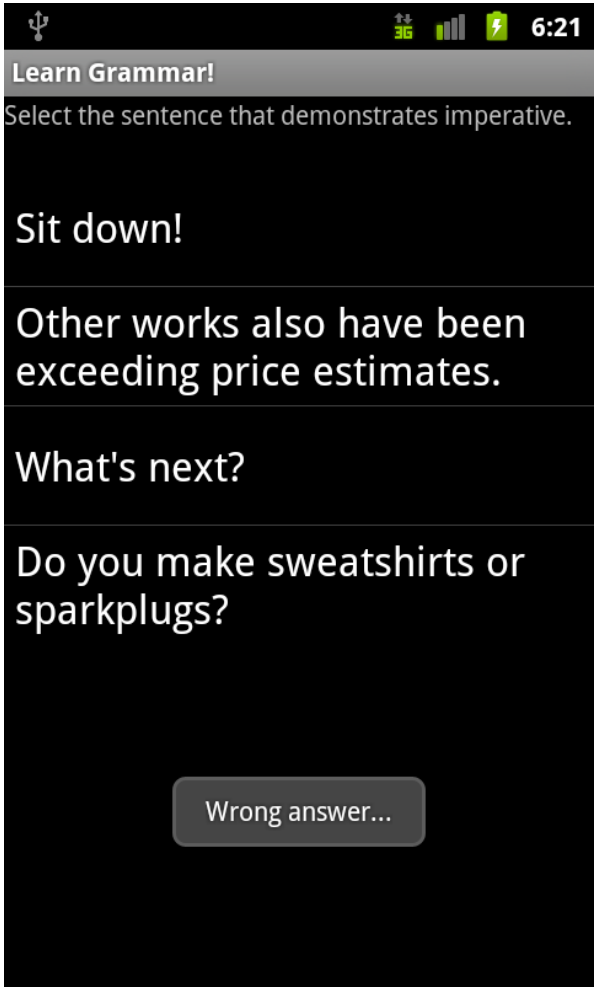


# Language learning through flash cards

But grammar?

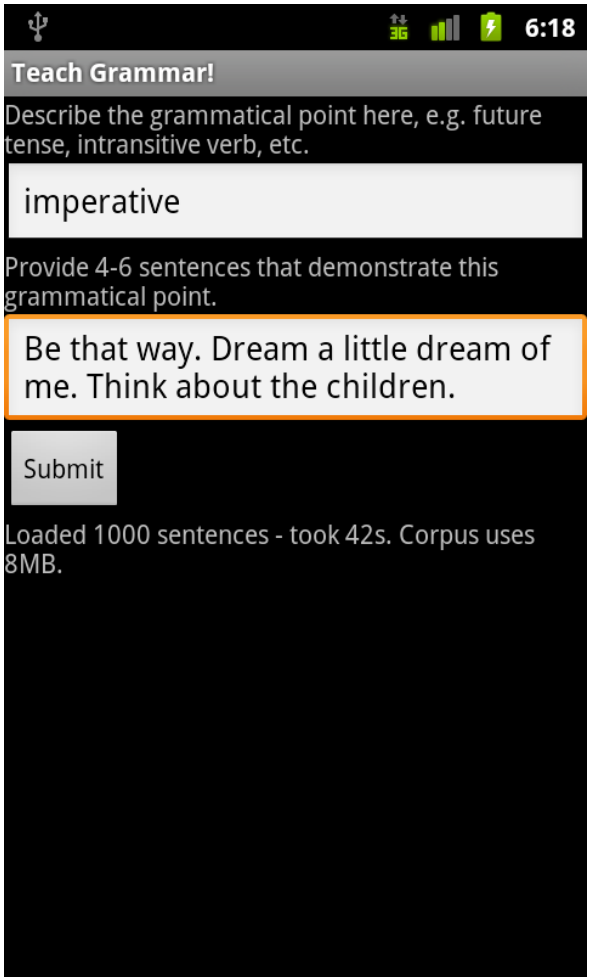
# Teaching grammar

- Need to make it quick and easy to create large sets of examples and counter-examples of arbitrary grammar points, e.g.
  - Future tense
  - Relative clauses
  - The imperative
  - Pretty much anything else – code should be task-independent



# Approach

1. Prompt 'teacher' for:
   - A title for our learning task (uninterpreted)
   - ~5 example sentences
   - 5 counter-example sentences
2. Work out what examples have in common (that's not in a counter-example)
3. Find many other sentences in a pre-parsed corpus exhibiting this *commonality*



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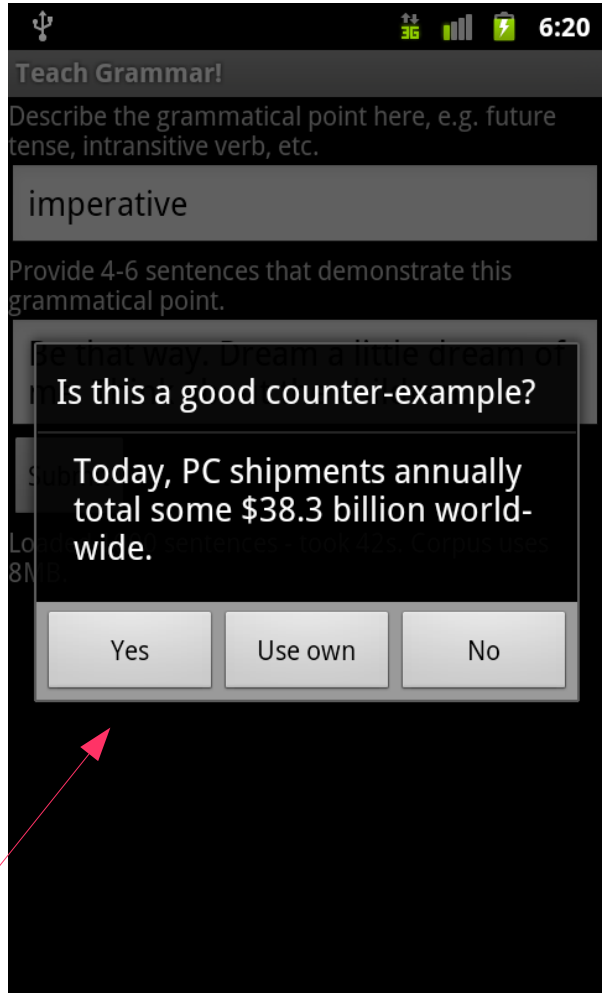



*Save user typing by suggesting random sentences as counter-examples*

# Analysing a sentence

- Each word has*:
  - A part of speech (POS) tag: plural noun, adjective, past tense verb...
  - A base form ('lemma'): *thought* → *think*

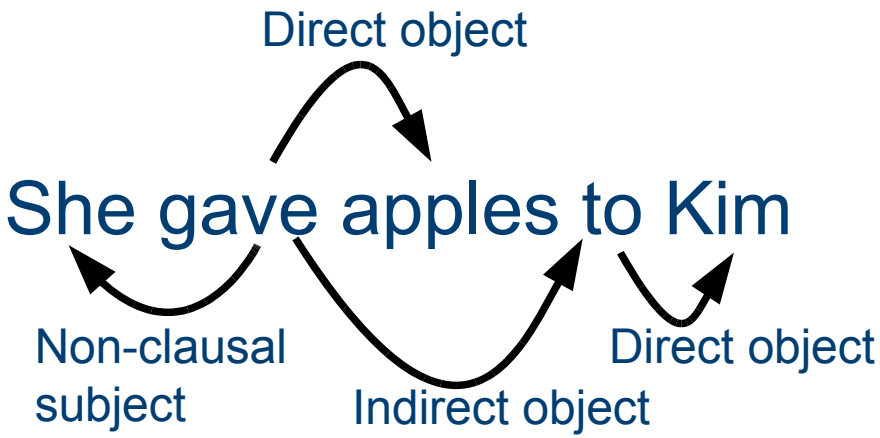
*Most features derived using C&C – Stephen Clark's syntactic parser

# Analysing a sentence

| Word | Lemma | POS |
|---|---|---|
| She | she | Personal pronoun |
| gave | *give* | Past tense |
| apples | *apple* | Plural common noun |
| to | to | Preposition |
| Kim | Kim | Proper noun |

# Analysing a sentence

We can also extract *grammatical relations* (GRs) between words:



# Commonality

What do the example sentences **have in common**?

- Used to find other, similar sentences

# Commonality – case study #1

Consider the “going to” future tense, e.g.

- He's going to be rich one day.
- I'm going to think about it.
- Becky is going to read her book.

(But *not “I'm going to the shop to buy some tea” - present continuous*)

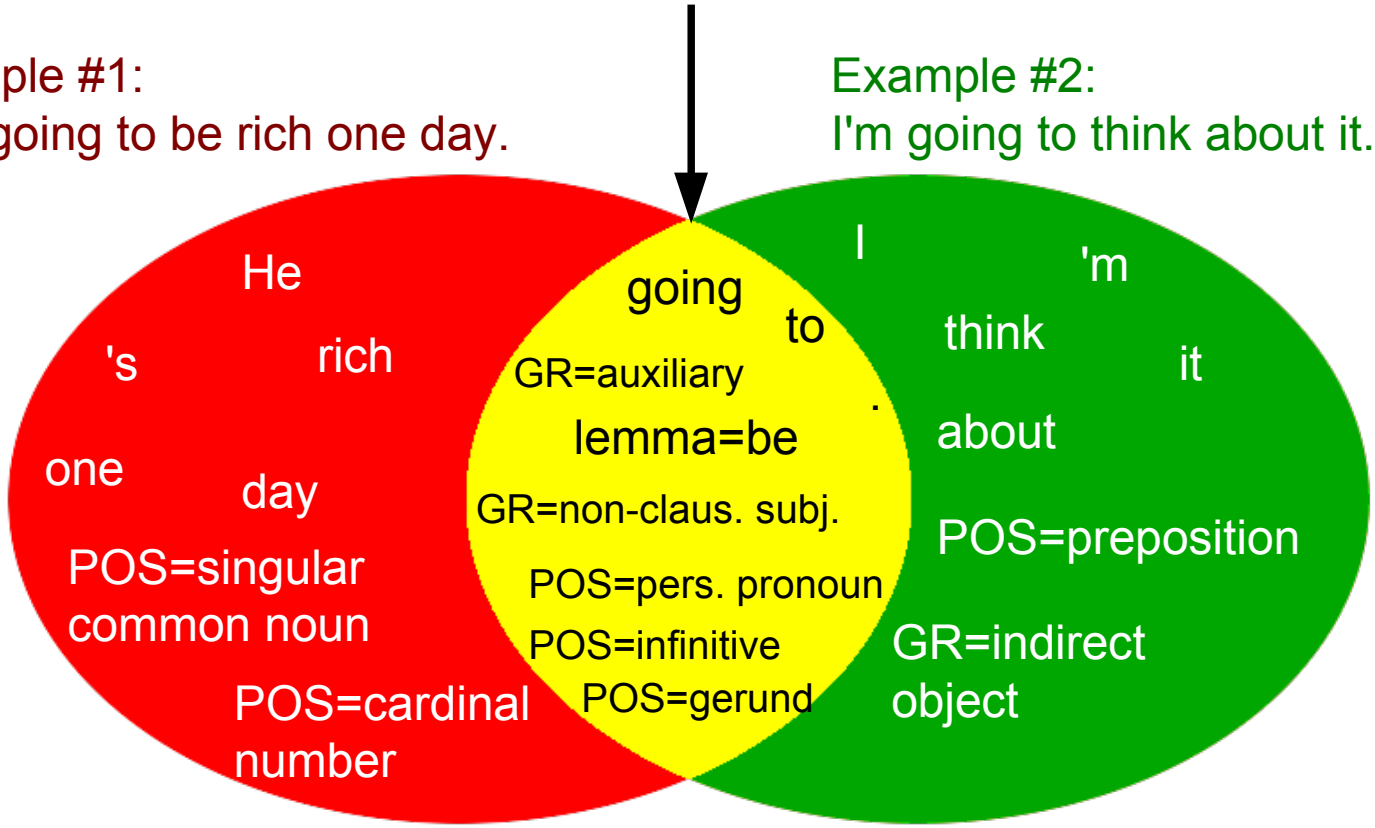
Can't spot imperative using approaches described so far...

# Weak commonality

*Weak* commonality: **intersection** of features observed in examples

Example #1:
He's going to be rich one day.

Example #2:
I'm going to think about it.



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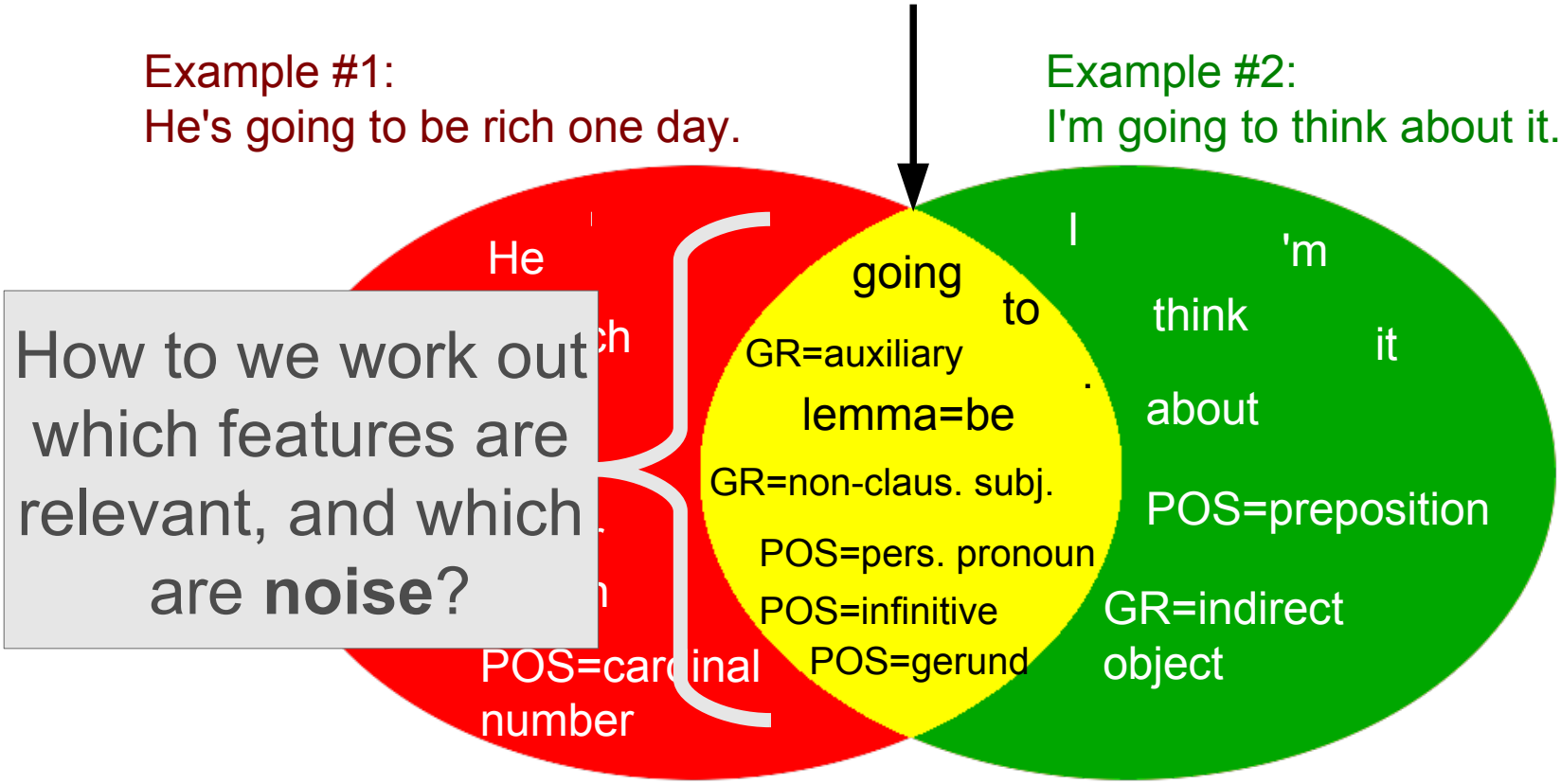
*Weak* commonality: **intersection** of features observed in examples

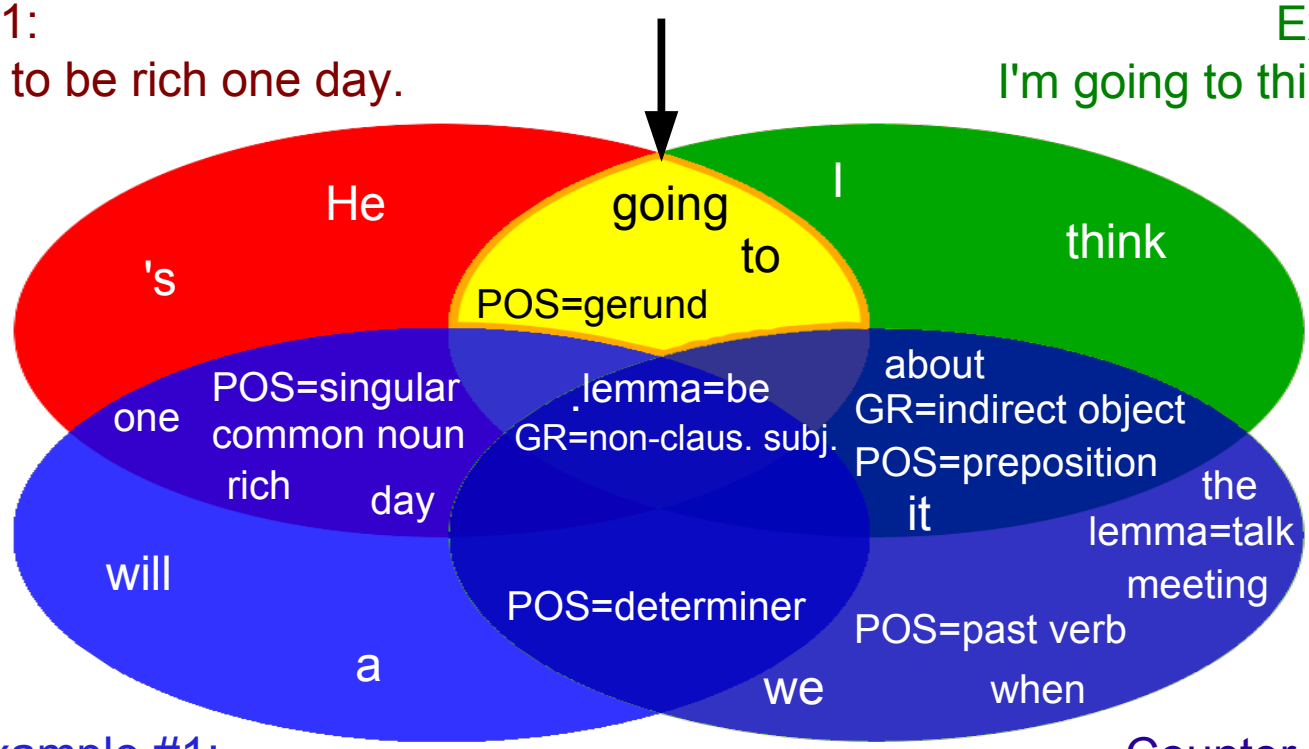


# Strong commonality

*Strong* commonality: **intersection** of features observed in examples, **minus union** of features in counter-examples

Example #1:
He's going to be rich one day.

Example #2:
I'm going to think about it.




Counter-example #1:
He will be a rich man one day.

Counter-example #2:
It was when we talked about the meeting.

# Commonality – case study #2

Another case study: the imperative mood, e.g.

- Be that way.
- Dream a little dream of me.
- Think about the children.

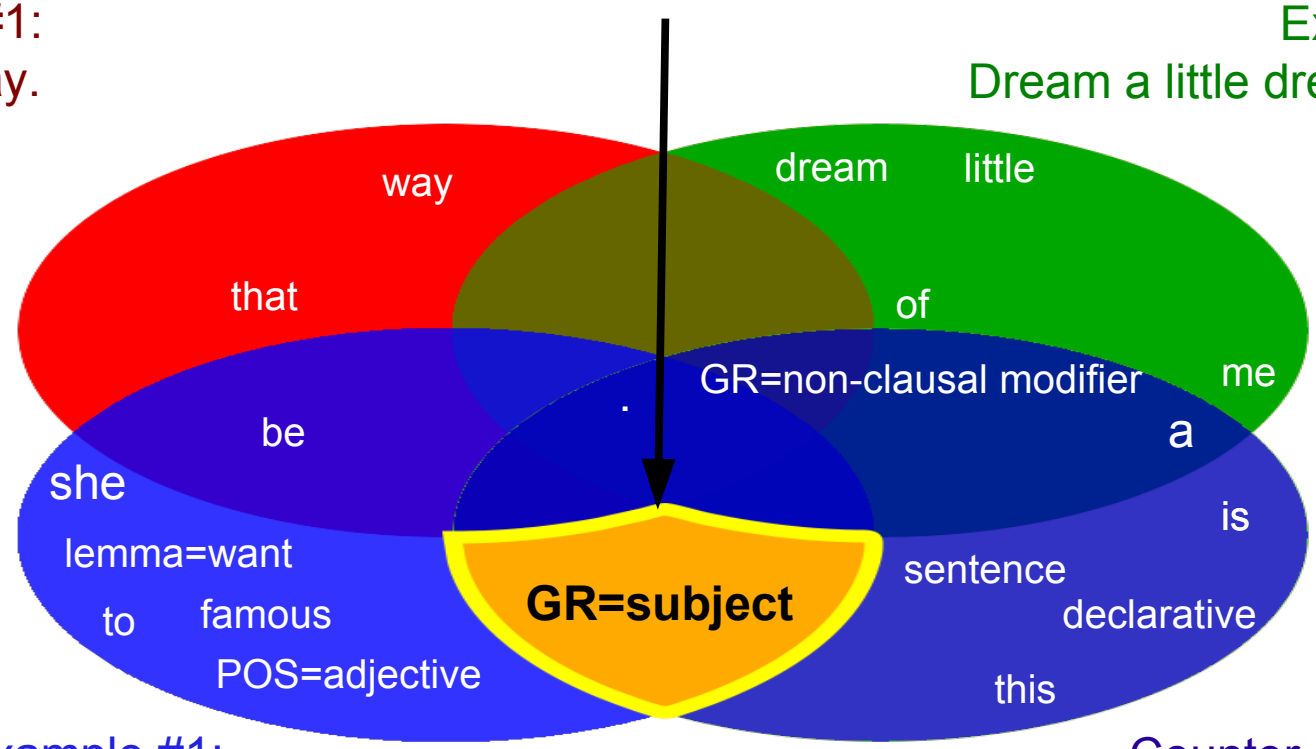
Can't spot imperative using approaches described so far...

# Absence-of commonality

*Absence-of* commonality: **intersection** of features observed in **counter-examples**, minus union of features in **examples**

Example #1:
Be that way.

Example #2:
Dream a little dream of me.




Counter-example #1:
She wanted to be famous.

Counter-example #2:
This is a declarative sentence.

# But it's not enough...

- This much works in a few simple cases, but:
  - False positive: “I'm going to the shops to buy some tea”
  - **Composite features** (e.g. a noun that is the subject of a verb) far more likely to capture commonality...

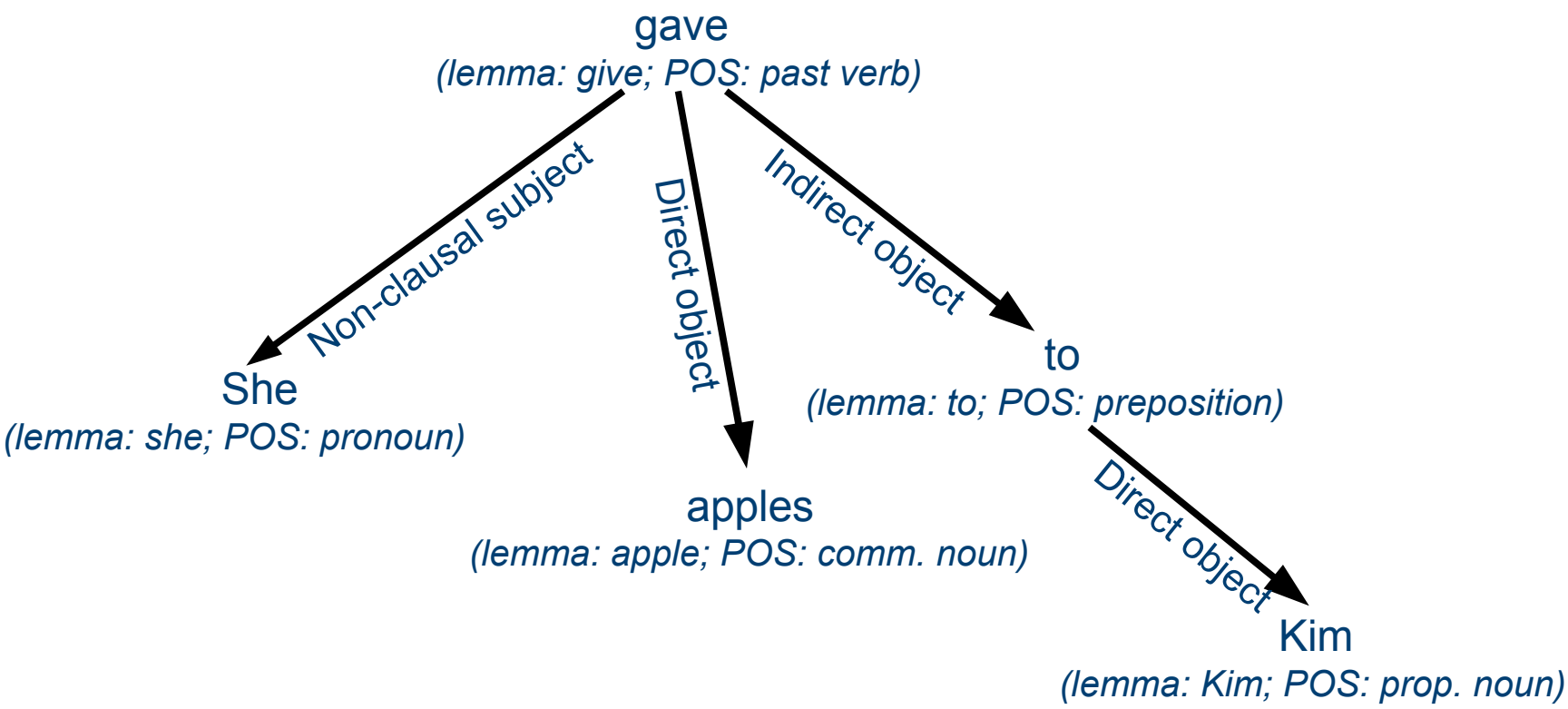
# Analysing a sentence again – dependency structures

GRs + words = *almost* a tree...

(directed graph, but sometimes cyclic)

*(Thank you, Google Image Search)*

# Analysing a sentence again – dependency structures



# Break [almost-]tree up into smaller features: 'twigs'

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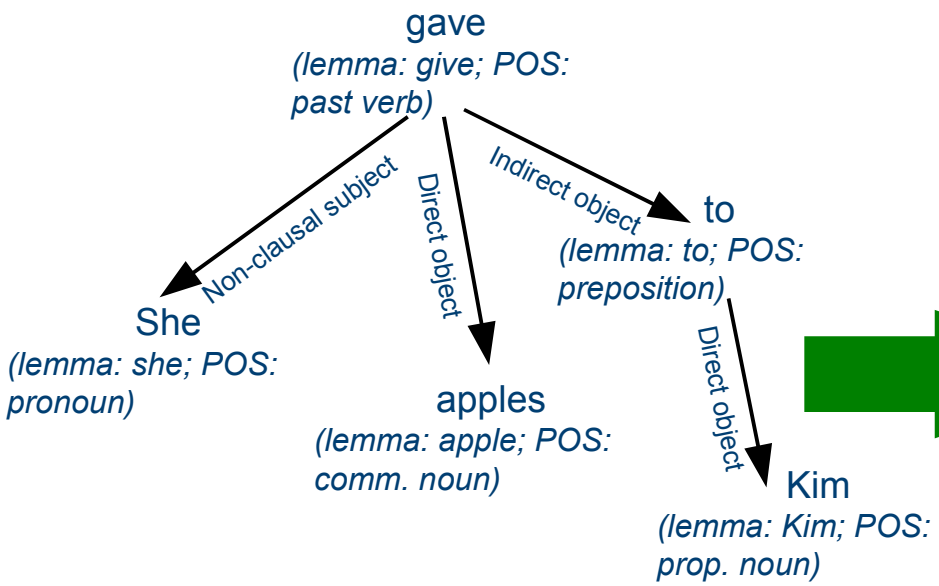
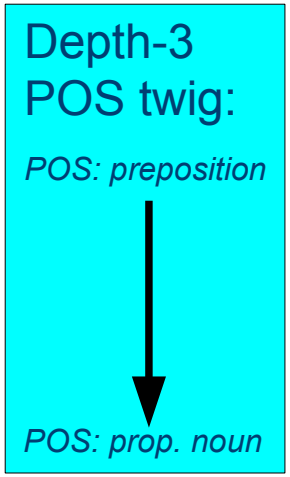
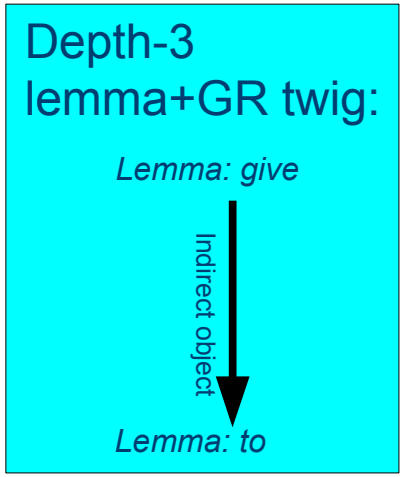
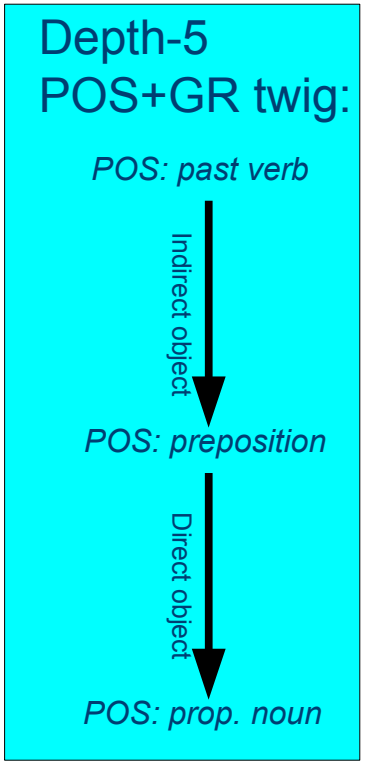



1. Break each example's graph into 'twigs'
2. Check which twigs are common to all examples



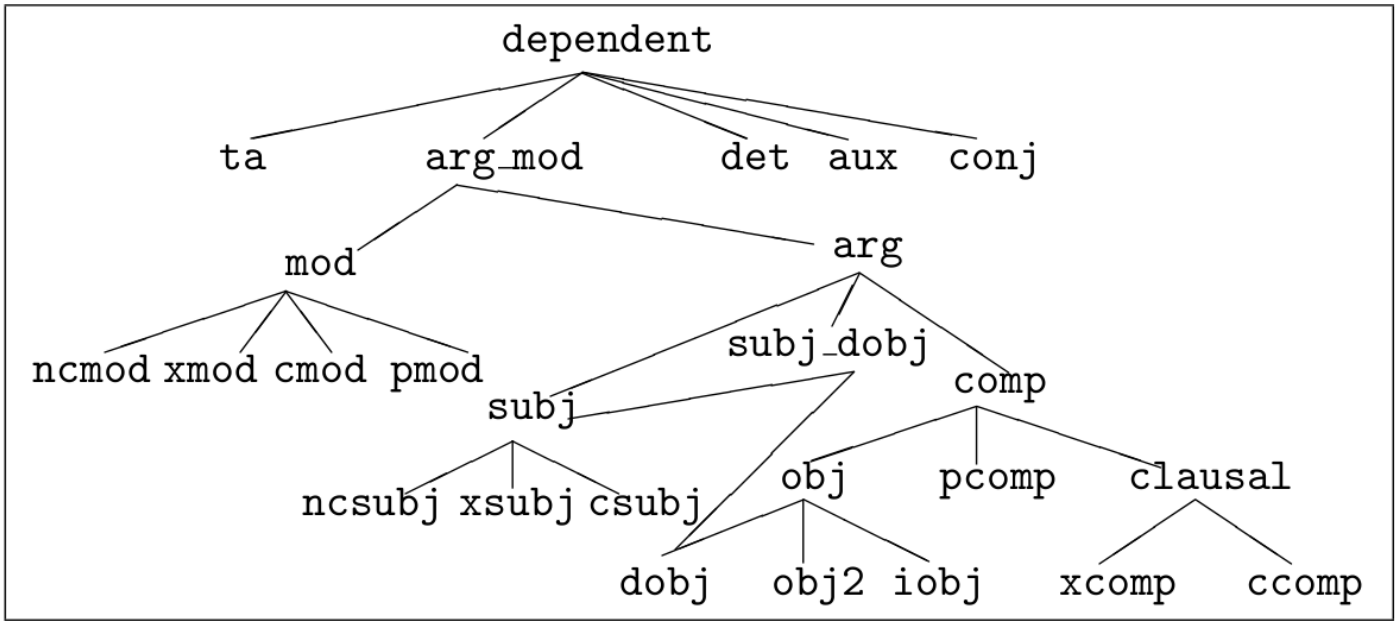



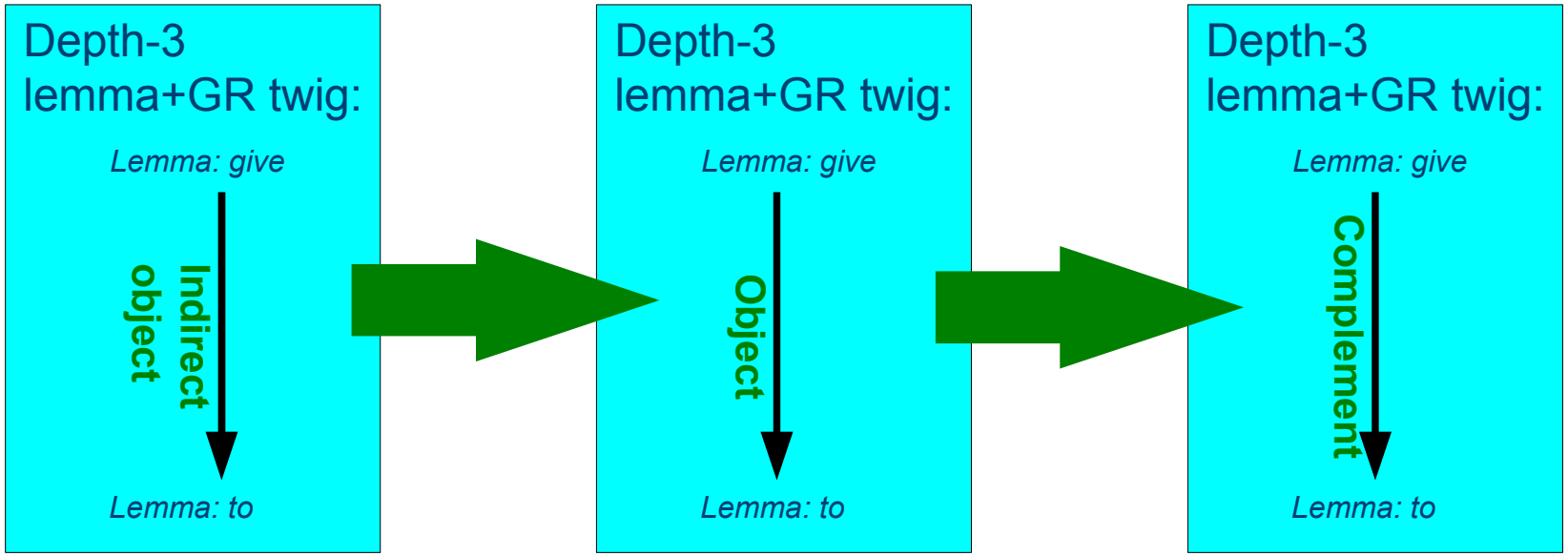

*...etc...*

# Sometimes this is still not enough...

- What if our learning task is transitive verbs, but we're not interested in the direct/indirect object distinction?
  - Use GR type hierarchy to generate features



# Exploiting the GR type hierarchy




- Go up GR type hierarchy, generate new feature at each level
- Might only find commonality between example sentences at a higher level
- Same approach for POS tags, but requires custom hierarchy: standard Penn-Treebank set is flat

# Running *LearnGrammar!* on Android (Nexus One)

- It works! (Mostly. Formal evaluation outstanding)
- Corpus load is slow
  - ~45s for 1000 pre-parsed sentences after mucking with file format
  - Run in background thread while user enters sentences
- Example sentence parsing via webservice call
- Searching for similar sentences is currently ~85s
  - Heavy pruning of redundant features required
  - Planning on implementing lazy search to avoid long up-front pause