


Applied Text Mining in Python

Semantic Text Similarity

Which pair of words are most similar?

- deer , elk
- deer , giraffe
- deer , horse
- deer , mouse

Which pair of words are most similar?

- deer , elk 
 - deer , horse
 - deer , house
 - deer , roof
-
- How can we quantify such similarity?

Applications of Text Similarity

- **Grouping similar words into semantic concepts**
- **As a building block in natural language understanding tasks**
 - **Textual entailment**
 - **Paraphrasing**

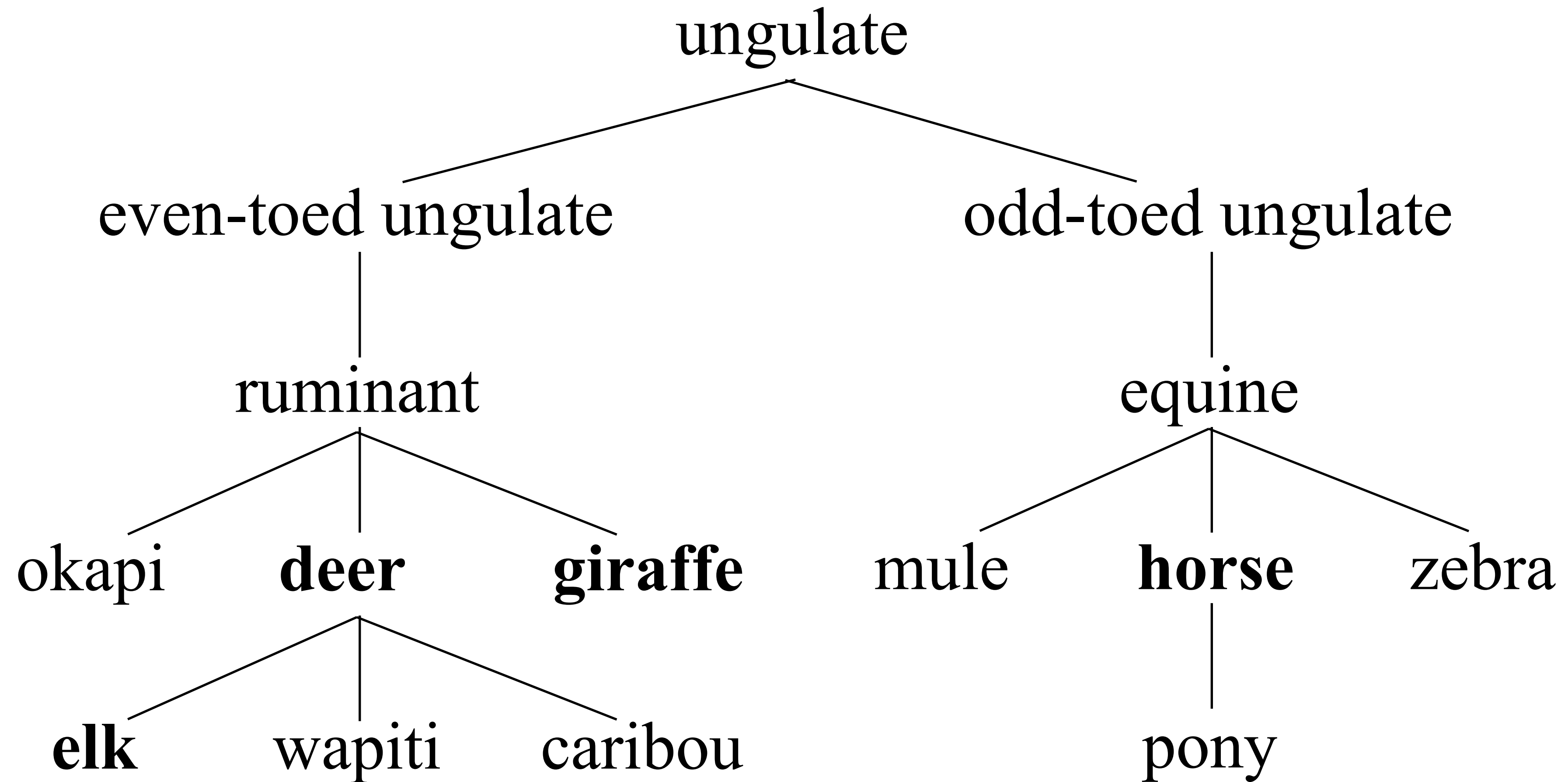
WordNet

- Semantic dictionary of (mostly) English words, interlinked by semantic relations
- Includes rich linguistic information
 - part of speech, word senses, synonyms, hypernyms/hyponyms, meronyms, distributional related forms, ...
- Machine-readable, freely available

Semantic Similarity Using WordNet

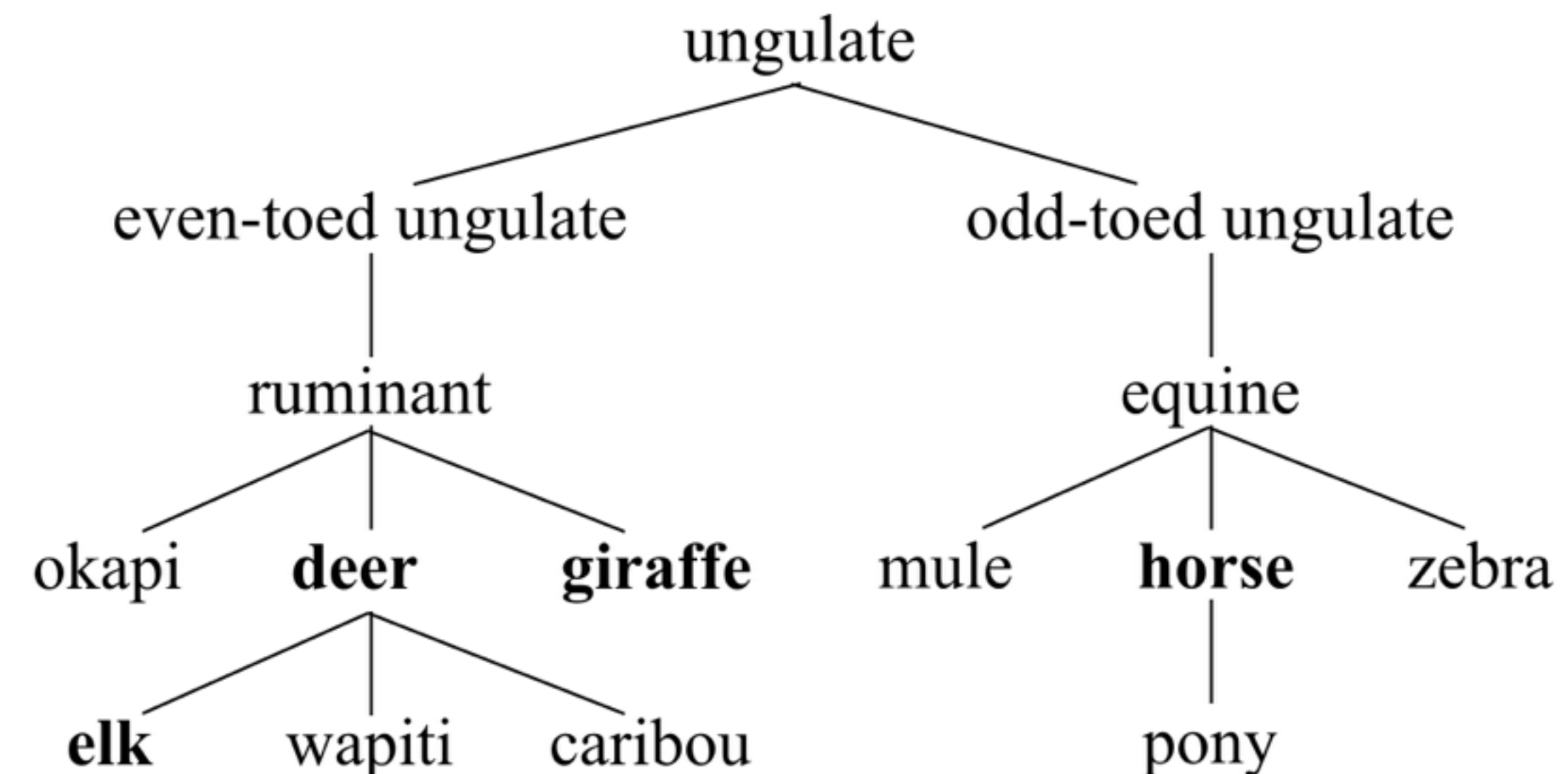
- **WordNet organizes information in a hierarchy**
- Many similarity measures use the hierarchy in some way
- Verbs, nouns, adjectives all have separate hierarchies

Coming back to our deer example



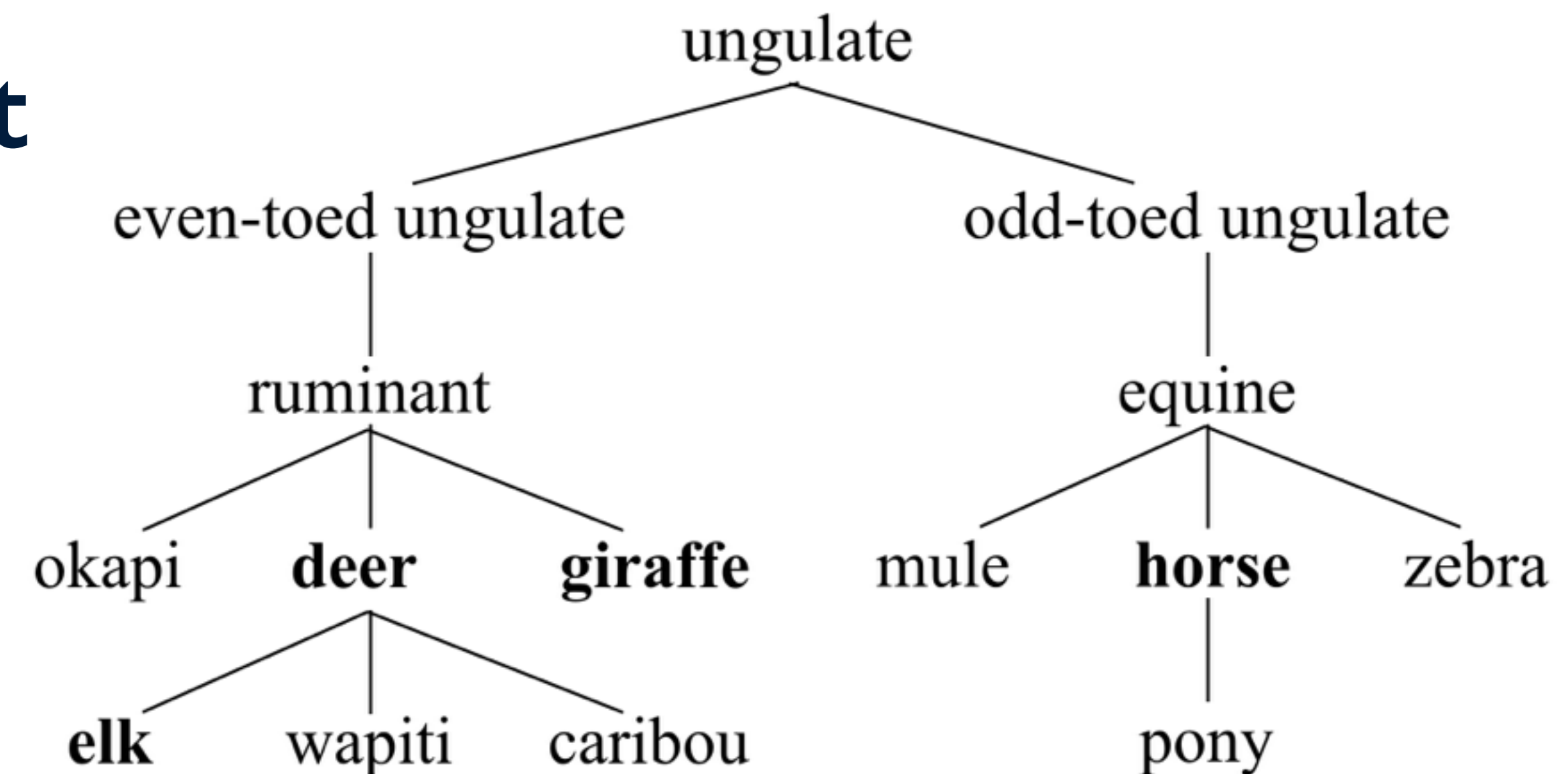
Path Similarity

- Find the shortest path between the two concepts
- Similarity measure inversely related to path distance
- $\text{PathSim}(\text{deer}, \text{elk}) = 0.5$
- $\text{PathSim}(\text{deer}, \text{giraffe}) = 0.33$
- $\text{PathSim}(\text{deer}, \text{horse}) = 0.14$



Lowest Common Subsumer (LCS)

- Find the closest ancestor to both concepts
- $\text{LCS}(\text{deer}, \text{elk}) = \text{deer}$
- $\text{LCS}(\text{deer}, \text{giraffe}) = \text{ruminant}$
- $\text{LCS}(\text{deer}, \text{horse}) = \text{ungulate}$



Lin Similarity

- Similarity measure based on the information contained in the LCS of the two concepts
- $\text{LinSim}(u, v) = 2 \times \log P(\text{LCS}(u, v)) / (\log P(u) + \log P(v))$
- $P(u)$ is given by the information content learnt over a large corpus.

How to do it in Python?

- **WordNet easily imported into Python through NLTK**

```
import nltk  
from nltk.corpus import wordnet as wn
```

- **Find appropriate sense of the words**

```
deer = wn.synset('deer.n.01')  
elk = wn.synset('elk.n.01')
```

...

How to do it in Python? (2)

- **Find path similarity**

```
deer.path_similarity(elk)      0.5  
deer.path_similarity(horse)  0.14285714285714285
```

- **Use an information criteria to find Lin similarity**

```
from nltk.corpus import wordnet_ic  
brown_ic = wordnet_ic.ic('ic-brown.dat')  
  
deer.lin_similarity(elk, brown_ic)      0.7726998936065773  
deer.lin_similarity(horse, brown_ic)  0.8623778273893673
```

Collocations and Distributional Similarity

- “You know a word by the company it keeps” [Firth, 1957]
- Two words that frequently appears in similar contexts are more likely to be semantically related
 - The friends met at a **café**.
 - Shyam met Ray at a **pizzeria**.
 - Let's meet up near the **coffee shop**.
 - The secret meeting at the **restaurant** soon became public.

Distributional Similarity: Context

- Words before, after, within a small window
- Parts of speech of words before, after, in a small window
- Specific syntactic relation to the target word
- Words in the same sentence, same document, ...

Strength of association between words

- How frequent are these?
 - Not similar if two words don't occur together often
- Also important to see how frequent are individual words
 - 'the' is very frequent, so high chances it co-occurs often with every other word
- Pointwise Mutual Information $PMI(w,c) = \log [P(w,c) / P(w)P(c)]$

How to do it in Python?

- **Use NLTK Collocations and Association measures**

```
import nltk
from nltk.collocations import *

bigram_measures = nltk.collocations.BigramAssocMeasures()

finder = BigramCollocationFinder.from_words(text)
finder.nbest(bigram_measures.pmi, 10)
```

- **finder also has other useful functions, such as frequency filter**

```
finder.apply_freq_filter(10)
```


Take Home Concepts

- Finding similarity between words and text is non-trivial
- WordNet is a useful resource for semantic relationships between words
- Many similarity functions exist
- NLTK is a useful package for many such tasks