### Lecture 11b: Annotation

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### Annotation

Why "gold"  $\neq$  perfect

**Quality Control** 

#### Factors in Annotation

Suppose you are tasked with building an annotated corpus. (E.g., with part-of-speech tags.) In order to estimate **cost** in time and money, you need to decide on:

- Source data (genre? size? licensing?)
- Annotation scheme (complexity? guidelines?)
- Annotators (expertise? training?)
- Annotation software (graphical interface?)
- Quality control procedures (multiple annotation, adjudication?)

#### Annotation Scheme

- Assuming a competent annotator, some kinds of annotation are straightforward for most inputs.
- ▶ Others are not.
  - Text may be ambiguous
  - There may be gray area between categories in the annotation scheme

### You play annotator

#### Noun or adverb?

- Yesterday was my birthday .
- Yesterday I ate a cake .
- ▶ He was fired **yesterday** for leaking the information .
- ▶ I read it in **yesterday** 's news .
- I had not heard of it until yesterday .

### You play annotator

#### Verb, noun, or adjective?

- ▶ We had been walking quite briskly
- Walking was the remedy, they decided
- In due time Sandburg was a walking thesaurus of American folk music.
- we all lived within walking distance of the studio
- a woman came along carrying a folded umbrella as a walking stick
- ► The **Walking** Dead premiered in the U.S. on October 31, 2010, on the cable television channel AMC

### Annotation: Not as easy as you might think

Pretty much any annotation scheme for language will have some difficult cases where there is gray area, and multiple decisions are plausible.

- ▶ Because human language needs to be **flexible**, it cuts corners and is reshaped over time.
- Not just syntax: wait till we get to semantics!

### **Annotation Guidelines**

However, we want a dataset's annotations to be as clean as possible so we can use them reliably in systems.

Documenting conventions in an annotation manual/standard/guidelines document is important to help annotators produce **consistent** data, and to help end users interpret the annotations correctly.

### **Annotation Guidelines**

- ▶ Penn Treebank: 36 POS tags (excluding punctuation).
- ► Tagging guidelines (3rd Revision): 34 pages
  - "The temporal expressions yesterday, today and tomorrow should be tagged as nouns (NN) rather than as adverbs (RB). Note that you can (marginally) pluralize them and that they allow a possessive form, both of which true adverbs do not." (p. 19)
  - An entire page on nouns vs. verbs.
  - 3 pages on adjectives vs. verbs.
- ▶ Penn Treebank bracketing (tree) guidelines: >300 pages!

### **Annotation Quality**

But even with extensive guidelines, human annotations won't be perfect:

- Simple error (hitting the wrong button)
- ▶ Not reading the full context
- Not noticing an erroneous pre-annotation
- Forgetting a detail from the guidelines
- Cases not anticipated by or not fully specified in guidelines (room for interpretation)

"Gold" data will have some tarnish. How can we measure its quality?

# Inter-annotator agreement (IAA)

- ► An important way to estimate the reliability of annotations is to have multiple people independently annotate a common sample, and measure **inter-annotator**/coder/rater **agreement**.
- ▶ Raw agreement rate: proportion of labels in agreement
- ▶ If the annotation task is perfectly well-defined and the annotators are well-trained and do not make mistakes, then (in theory) they would agree 100%.
- ▶ If agreement is well below what is desired (will differ depending on the kind of annotation), examine the sources of disagreement and consider additional training or refining guidelines.
- The agreement rate can be thought of as an upper bound (human ceiling) on accuracy of a system evaluated on that dataset.

### IAA: Beyond raw agreement rate

- Raw agreement rate counts all annotation decisions equally.
- ▶ Some measures take knowledge about the annotation scheme into account (e.g., counting singular vs. plural noun as a minor disagreement compared to noun vs. preposition).
- ▶ What if some decisions (e.g., POS tags) are far more frequent than others?
  - ▶ If 2 annotators both tagged *hell* as a noun, what is the chance that they agreed **by accident**? What if they agree that it is an interjection (rare tag)—is that equally likely to be an accident?
  - ▶ Chance-corrected measures such as Cohen's kappa  $(\kappa)$  adjust the agreement score based on label probabilities. (Cohen's assumes 2 raters, categorical labels)
  - ...but they make modeling assumptions about how "accidental" agreement would arise; important that these match the reality of the annotation process!

### Cohen's $\kappa$

- ▶ 2 raters (annotators A and B), categorical labels  $(y_1, y_2, ...)$
- From interannotator confusion matrix, compute:
  - Observed probability of agreement (i.e., raw agreement rate):  $p_0 = \hat{P}(A = B = y_1) + \hat{P}(A = B = y_2) + \cdots$
  - Expected agreement by chance if annotators' decisions were independent:

$$p_e = \hat{P}(A = y_1)\hat{P}(B = y_1) + \hat{P}(A = y_2)\hat{P}(B = y_2) + \cdots$$

Agreement above chance:

$$\kappa = \frac{p_o - p_e}{1 - p_e}$$

- ▶ Interpretation of  $\kappa$  is subjective.
  - ▶ Landis and Koch (1977): 0–0.20 is "slight" agreement, 0.21–0.40 is "fair", 0.41–0.60 is "moderate", 0.61–0.80 is "substantial", and 0.81–1 is "almost perfect"
- Assumes that chance is random guessing according to one's overall preferences—not always realistic!
- ▶ Tends to underestimate agreement for rare labels.

## Crowdsourcing

- Quality control is even more important when eliciting annotations from "the crowd".
- ► E.g., Amazon Mechanical Turk facilitates paying anonymous web users small amounts of money for small amounts of work ("Human Intelligence Tasks").
- ▶ Need to take measures to ensure annotators are qualified and taking the task seriously.
  - ▶ Redundancy to combat noise: Elicit 5+ annotations per data point.
  - Embed data points with known answers, reject annotators who get them wrong.