### Week 4 Video 7

#### Memory Algorithms

### Is future correctness enough?

 Up until this point we've been talking about predicting future correctness

### But what if you forget it tomorrow?

Another way to look at knowledge is – how long will you remember it?

#### Relevant for all knowledge

- Mostly studied in the context of memory for facts, rather than skills
- How do you say banana in Spanish?
- What is the capital of New York?
- Where are the Islands of Langerhans?

### Spacing Effect

- It has long been known that spaced practice (i.e. pausing between studying the same fact) is better than massed practice (i.e. cramming)
- Early adaptive systems implemented this behavior in simple ways (i.e. Leitner, 1972)

 Memory duration can be understood in terms of memory strength (referred to as activation)

Formula for probability of remembering

$$\square P(m) = \frac{1}{1 + e^{\frac{\tau - m}{S}}}$$

- Where m = activation strength of current fact
- $_{\Box}$   $\tau$  = threshold parameter for how hard it is to remember
- s is noise parameter for how sensitive memory is to changes in activation
- Note logistic function (like PFA)

Formula for activation

$$m_n(t_{1...n}) = \ln(\sum_{i=1}^n t_i^{-d})$$

- We have a sequence of n cases where the learner encountered the fact
- lacktriangle Each  $t_i$  represents how long ago the learner encountered the fact for the i-th time
- The decay parameter d represents the speed of forgetting under exponential decay

- Implications
- More practice = better memory
- More time between practices = better memory
  - Most efficient learning comes from dense practice followed by expanding amounts of time in between practices (Pavlik & Anderson, 2008)

### MCM (Mozer et al., 2009)

- Postulates that decay speed drops, the more times a fact is encountered
- Functionally complex model where
  - Knowledge strength (and therefore probability of remembering) is a function of the sum of the traces' actual contributions, divided by the product of their potential contributions
  - Power function is estimated as a combination of exponential functions

### DASH (Mozer & Lindsay, 2016)

- DASH Extends previous approaches to also include item difficulty and latent student ability
- Can use either MCM or ACT-R as its internal representation of how memory decays over time

### Duolingo (Settles & Mercer, 2016)

- Fits regression model to predict both recall and estimated half-life of memory (based on lag time)
- Based on estimate of exponential decay of memory

### Duolingo (Settles & Mercer, 2016)

- Uses feature set including
  - Time since word last seen
  - Total number of times student has seen the word
  - Total number of times student has correctly recalled the word
  - Total number of times student has failed to recalled the word
  - Word difficulty

#### Another area of active development

- Watch this space, approaches rapidly changing
- Recent emerging approaches have not yet gone "head to head" against each other

#### Next Week

Relationship Mining