# Resource-bounded Crowd-sourcing of Commonsense Knowledge

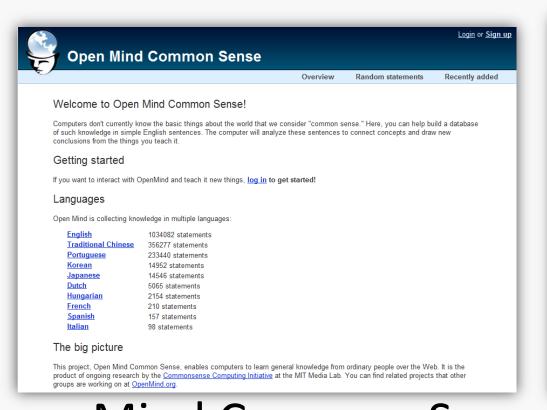
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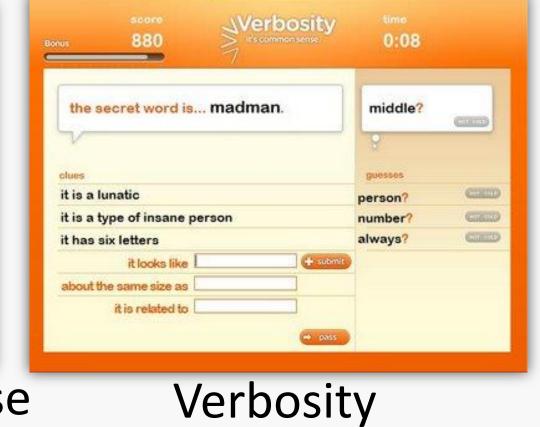
### The Framework

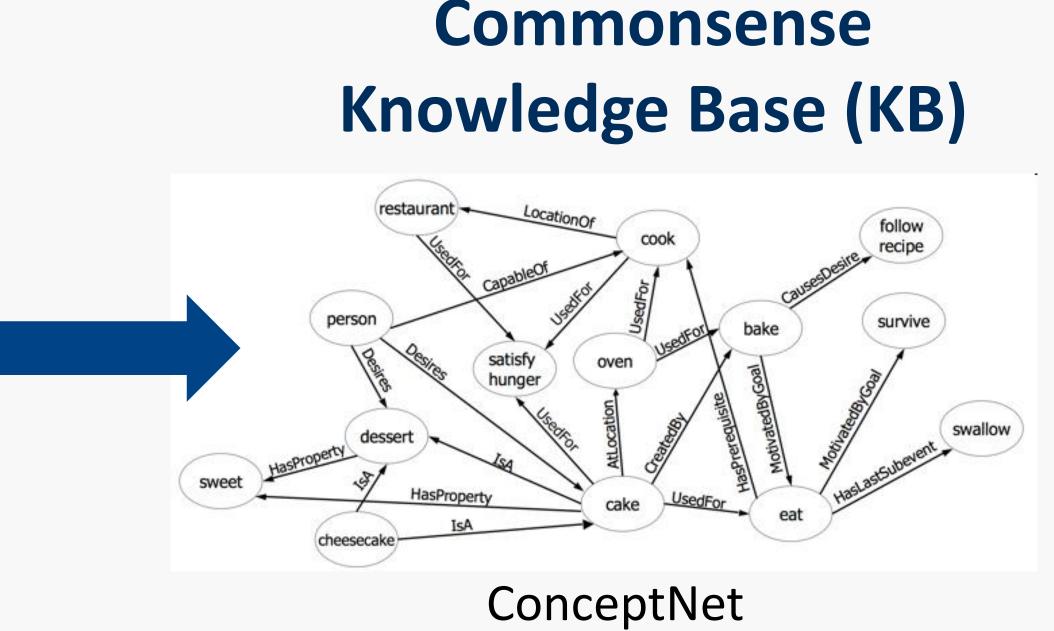
Ask commonsense question: Dog is a \_\_\_\_? m users Answer question: Dog is an animal. Dog is a creature. Resource bound: Answer O questions

within time T

### Website/games for crowd-sourcing







Open Mind Common Sense



Virtual Pets

Common Consensus

What are some things you would use to:

watch a movie?

# The Problem

- > Unguided collection suffers from high redundancy, e.g. Virtual Pets
- > How do we crowd-source commonsense knowledge effectively within the resource bound, i.e. users answer limited questions within a time T?

# **Overlapped data in Virtual Pets** Only 2/3 new assertions! 1500 1000 500

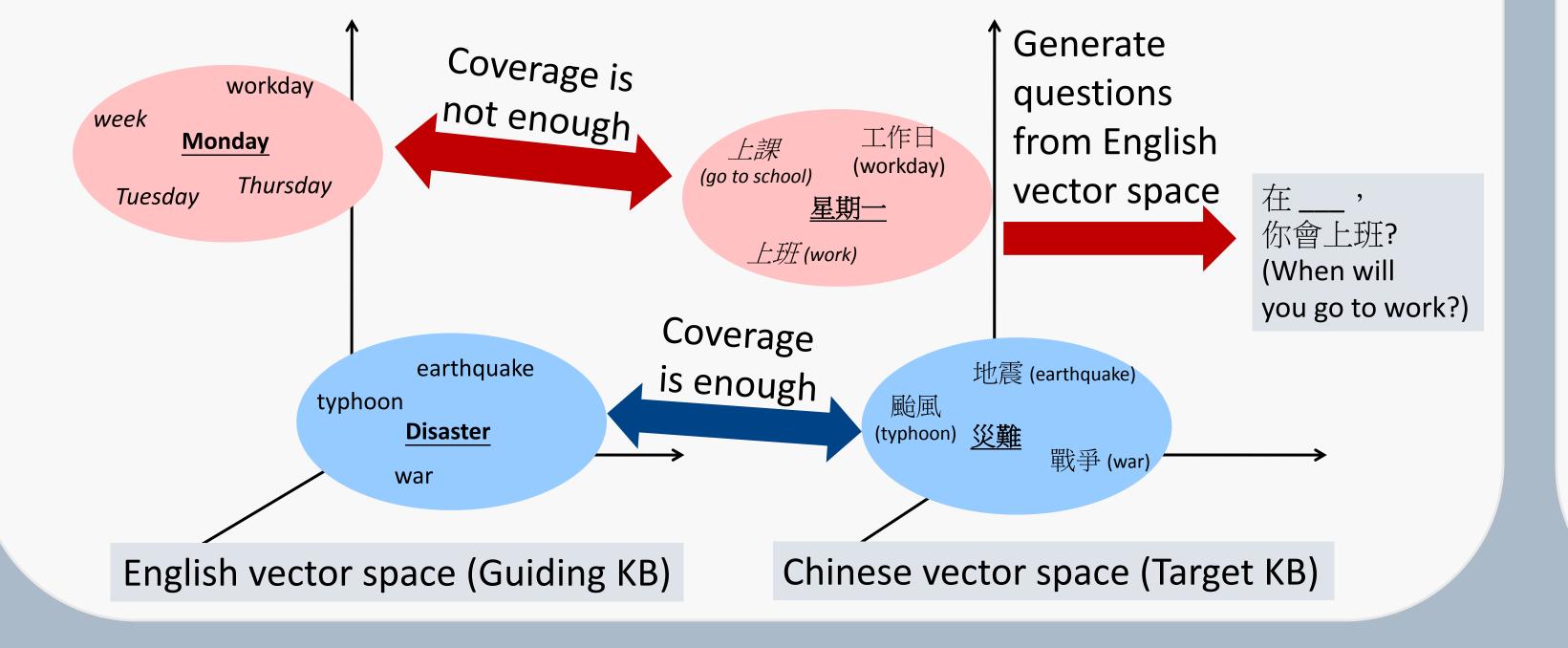
# KB Approximation

#### Main Idea:

- > We should identify the most productive questions for acquiring answers
- > A guiding KB helps us identify the questions by estimating the answers of a question and their inference results before asking users

### Algorithm:

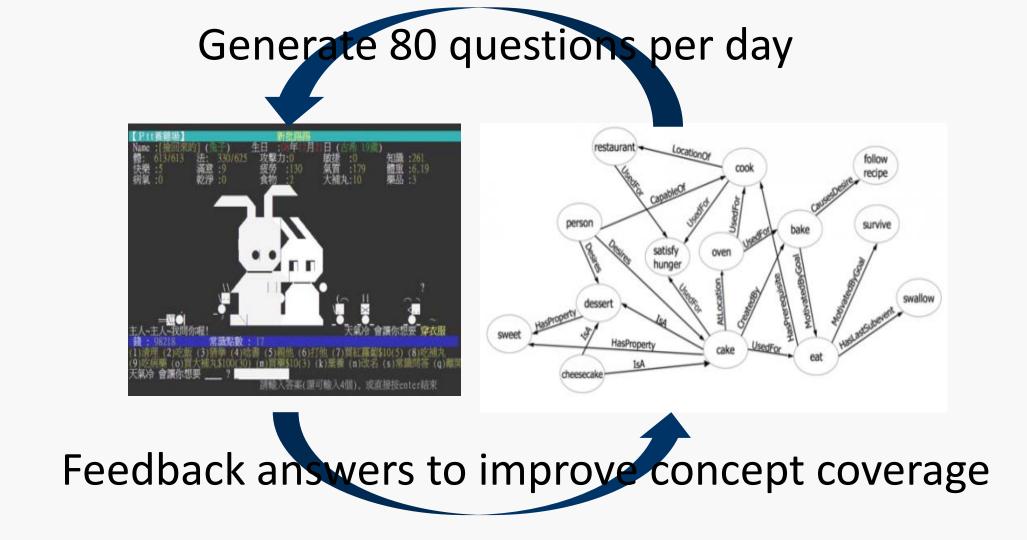
- 1. Represent every concept in a KB as a feature vector in vector space. For example, "dog" is a concept and "is an animal" is its feature.
- 2. For every mapped concept in the two KB, find their similar concepts.
- 3. If the overlap of the similar concepts in guiding KB and target KB is low, transfer features in guiding KB to create questions for crowd-sourcing.



### **Experimental Result**

### Setup:

- Enhance Chinese ConceptNet by generating questions from English ConceptNet
- > 173.8 questions are answered in Virtual Pets per day
- > Perform the collection process for six weeks



#### **Results:**

Quality of generated question

	By our algorithm	By players
# of questions	480	4,329
# of bad questions	28	264
% of good questions	94.17%	93.90%

- Precision of collected answers 85.37% (about 80% for Virtual Pets without guide)
- > Improvement of concept coverage

	week 0	week 5	week 6	week 7	week 8
$ c^- $	8,630	5,783	5,495	5,450	5,435
Δ	_	33.51%	36.33%	36.85%	37.02%

<sup>\*|</sup>c-|: # of concept whose coverage score < 0.5