Compositional Language Modeling for Icon-Based Augmentative and Alternative Communication

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Introduction

Augmentative and Alternative Communication (AAC) techniques are used by individuals who experience speech/language impairments to communicate with their surroundings. Icons are often employed as a symbol set since they:

- Require less selections than letters
- Do not require literacy
- Can be more intuitive than words

Limitation with Using Icons

Limited icon display layouts (often limited by the screen size) prevent conveying rich messages.

While a 'scroll-down' or a tree-based layout can allow for large vocabularies, they incur a high cost of searching for the desired icon.

The SymbolStix Icon Set

We are working with a human curated commercial icon set used by a variety of AAC devices.

name: agree
word type: verb
synonyms: agreement
agreed
agreeing
flexibility
on the same page see
eye to eye





name: afraid

word type: adjective

synonyms:

eerie frightened

fear halloween

feared nervous

fearful scared

fears scary

upset terrified

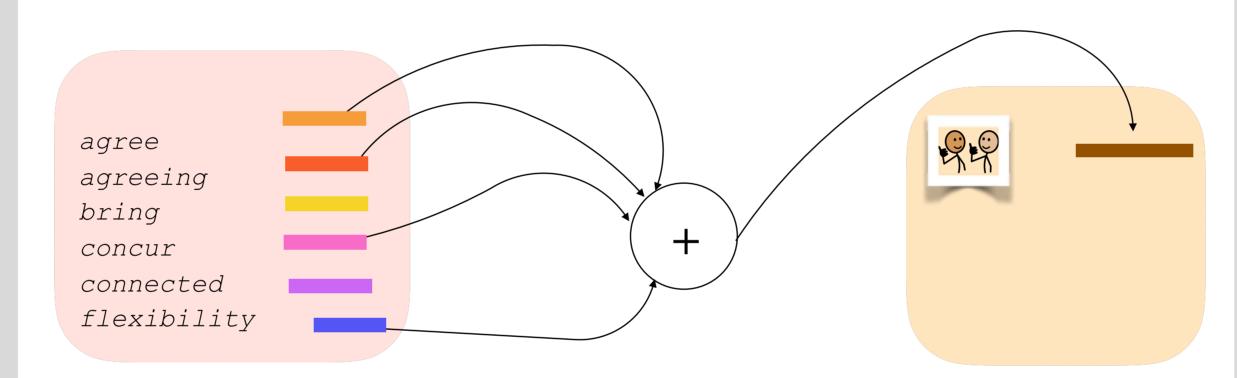
Each icon represents a brand set of meanings, including multiple terms as well as entire phrases.

Addressing the problems

Having an **Icon language model (LM)** to predict to the next icon can speed up the process to compose a message

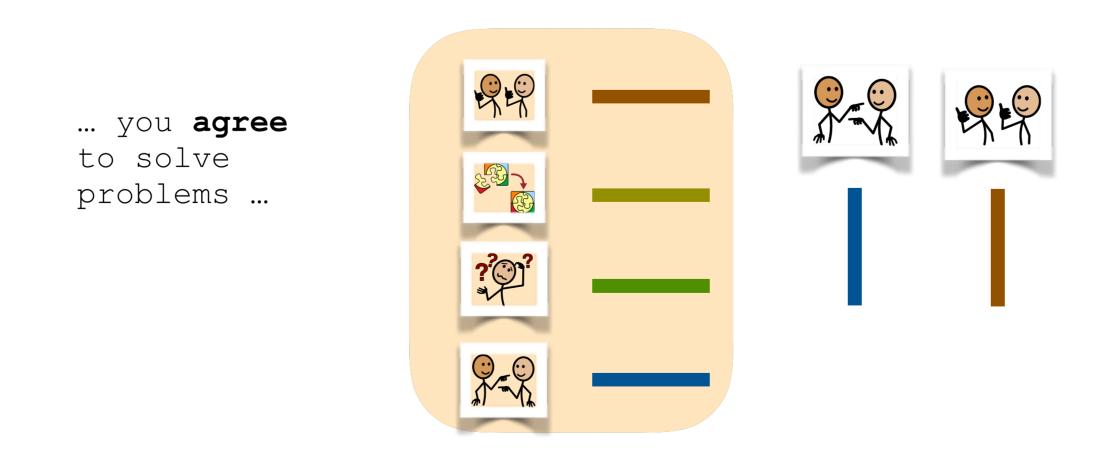
Problem: The Icon set is corpusless Goal: Simulate the Icon Language

1. Icon Representation



We create icon embeddings by composing metadata term vectors from a pre-trained word embedding set. This way, each icon's vector reflects its **broad** meaning.

2. The Simulated Corpus



To simulate an icon corpus, we incorporate a textual corpus and map each textual term to its icon equivalent.

Experiments

We trained our LMs on the SUBLEX-US corpus using two **different embedding types** and two **mapping strategies**, and evaluated their performance on a simulated predictive typing icon task.



 No meaningful difference was found Lower performance that might be due to sparse patterns

Simulated Corpus examples



There are tradeoffs when simulating an icon language following textual sequences.

Our simulation should be further compared to a naturally evolved icon language. Then, we'll conduct experiments with real users, develop multi-sense representations, and work towards multi-phrase representations of icons.