

## 1. Extremely large models and GPT3

- > better than other models at language modeling and related tasks such as story completion
- > commands to bash one-liners
- > blend concepts i.e. find a word between 2 given words
- > copycat analogy problems

## GPT-3: Limitations and Open questions

- Seems to do poorly on more structured problems that involve decomposing into atomic / primitive skills:
  - RTE / arithmetic / word problems / analogy making
- Performing permanent knowledge updates interactively is not well studied.
- **Doesn't seem to exhibit human like generalization (systematicity).**
- **Language is situated and GPT-3 is merely learning from text without being exposed to other modalities.**

## What's new about GPT-3: Flexible “in-context” learning

- GPT-3 demonstrates some level of fast adaptation to completely new tasks. This happens via “in-context” learning
- The language model training (outer loop) is learning how to learn from the context [inner loop]



Source: <https://arxiv.org/pdf/2005.14165.pdf>

## 2. Compositional Representations and Systematic Generalization

### -1. Are neural representations compositional?

## Compositional Representations and Systematic Generalization

**Compositionality:** closely related to the idea of systematicity is the principle of compositionality.

### Rough Definition:

"The meaning of an expression is a function of the meaning of its parts"

### More concrete definition (Montague):

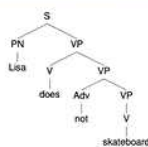
A homomorphism from syntax (structure) to semantics (meaning). That is, meaning of the whole is a function of immediate constituents (as determined by syntax)

### -2. Do neural NLP models generalize systematically?

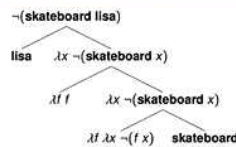
Systematicity and Compositional Generalization:  
Are neural representations compositional?

- According to Montague, Compositionality is about the existence of a homomorphism from syntax to semantics:

Lisa does not skateboard =  
 $\langle \text{Lisa}, \langle \text{does}, \langle \text{not}, \text{skateboard} \rangle \rangle \rangle$



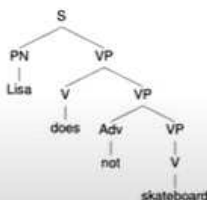
$m(\text{Lisa does not skateboard}) =$   
 $\langle m(\text{Lisa}), \langle m(\text{does}), \langle m(\text{not}), m(\text{skateboard}) \rangle \rangle \rangle$



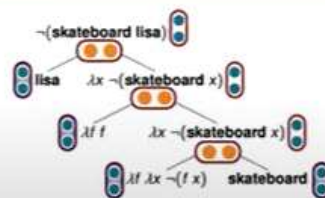
- Tree Reconstruction Error (TRE) [Andreas 2019]: Compositionality of **representations** is about how well the representation approximates an explicitly homomorphic function in a *learnt representation space*

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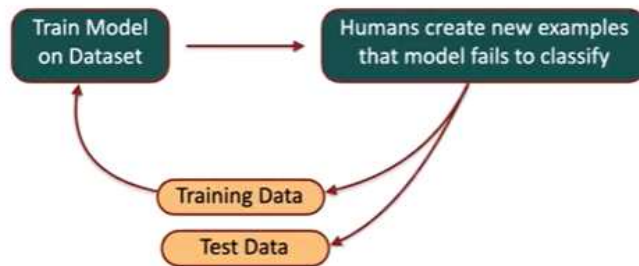
Lisa does not skateboard =  
 $\langle \text{Lisa}, \langle \text{does}, \langle \text{not}, \text{skateboard} \rangle \rangle \rangle$



$NN(\text{Lisa does not skateboard}) \approx$   
 $f(v(\text{Lisa}), f(v(\text{does}), f(v(\text{not}), v(\text{skateboard}))))$



### 3. Improving how we evaluate models in NLP

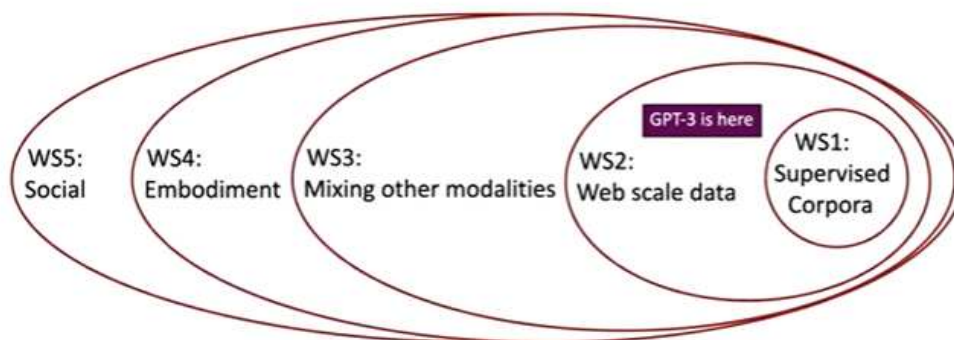


Overview of dynamic benchmarks

### 4. Grounding language to other modalities

#### Grounding Language to other modalities

- Many have articulated the need for using modalities other than text
- Bender and Koller [2020]: Impossible to acquire “meaning” (communicative intent of the speaker) from form (text / speech signal) alone
- Bisk et al [2020]: Training on only web-scale data limits the world scope of models.



### 5. Getting involved with NLP and Deep Learning Research