

# Data Recombination for Neural Semantic Parsing



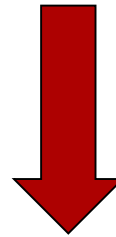
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Stanford University



# Semantic Parsing

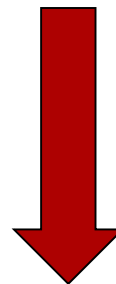
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*What states border Texas ?*



Semantic Parser

*And(State, NextTo(StateId(Texas)))*



Executor

[New Mexico, Oklahoma, Arkansas, Louisiana]



# Outline

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- Neural Semantic Parser
- Data Recombination
- Discussion



# Outline

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- Neural Semantic Parser
- Data Recombination
- Discussion



# Semantic Parsing

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- Traditional semantic parsers are engineering-heavy
- Can we get good performance with a domain-general model?



# Semantic Parsing

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- Geoquery
  - Input: *what is the population of iowa ?*
  - Output: `_answer ( A , ( _population ( B , A ) , _const ( B , _stateid ( iowa ) ) ) )`
- ATIS
  - Input: *list all flights from chicago to milwaukee*
  - Output: `( _lambda $0 e ( _and ( _flight $0 ) ( _from $0 chicago:_ci ) ( _to $0 milwaukee:_ci ) ) )`
- Overnight
  - Input: *what restaurants have takeout*
  - Output: `( call list ( call filter ( call getProperty ( call singleton en.restaurant ) ( string ! type ) ) ( string takeout ) ) )`



# Sequence-to-sequence Models

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- Machine Translation
  - Input: the blue house
  - Output: la maison bleue
- Syntactic Parsing
  - Input: the dog barked
  - Output: (S (NP DT NN )<sub>NP</sub> (VP VBD)<sub>VP</sub> )<sub>S</sub>
- And many more...



# Semantic Parsing

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- Treat semantic parsing as a sequence-to-sequence task
- Use domain-general attention-based neural model

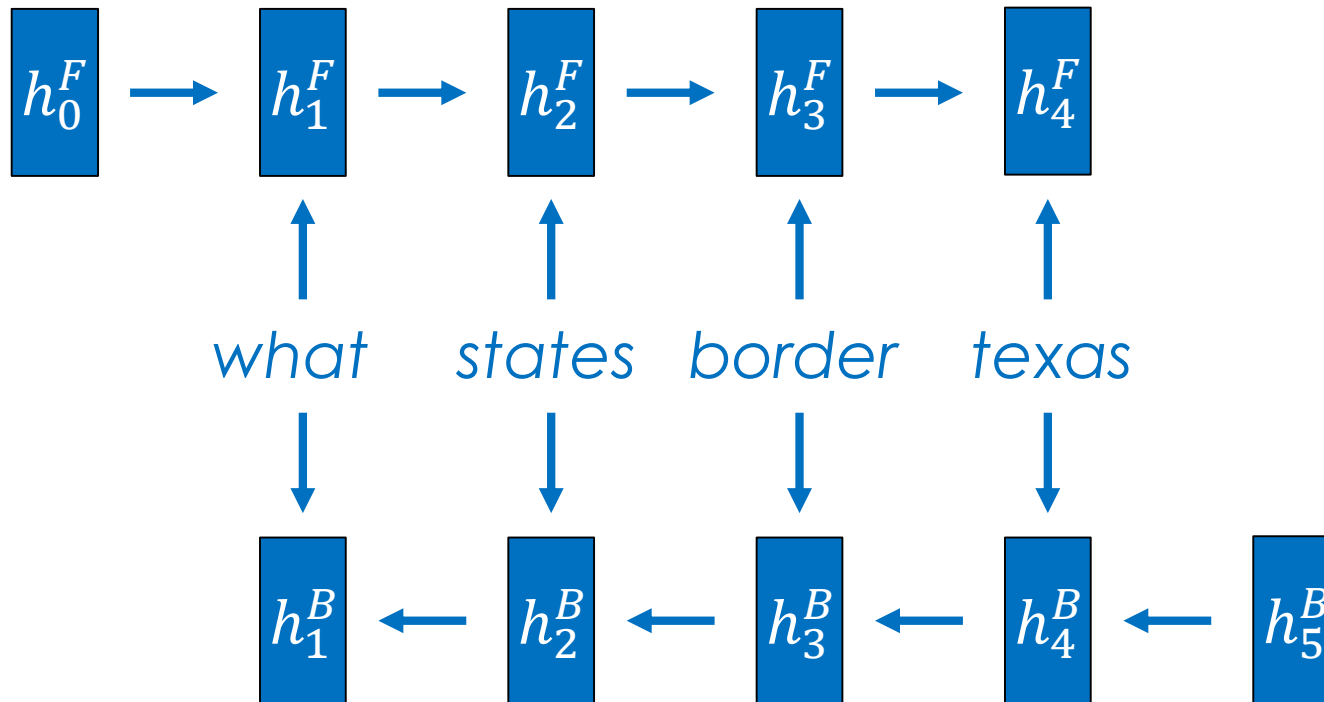
Sutskever et al. (2014), Bahdanau et al. (2014), Luong et al. (2015).





# Neural Semantic Parser

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# Neural Semantic Parser

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$h_1^F$

$h_2^F$

$h_3^F$

$h_4^F$

*what states border texas*

$h_1^B$

$h_2^B$

$h_3^B$

$h_4^B$



# Neural Semantic Parser

---

*what states border texas*

$b_1$

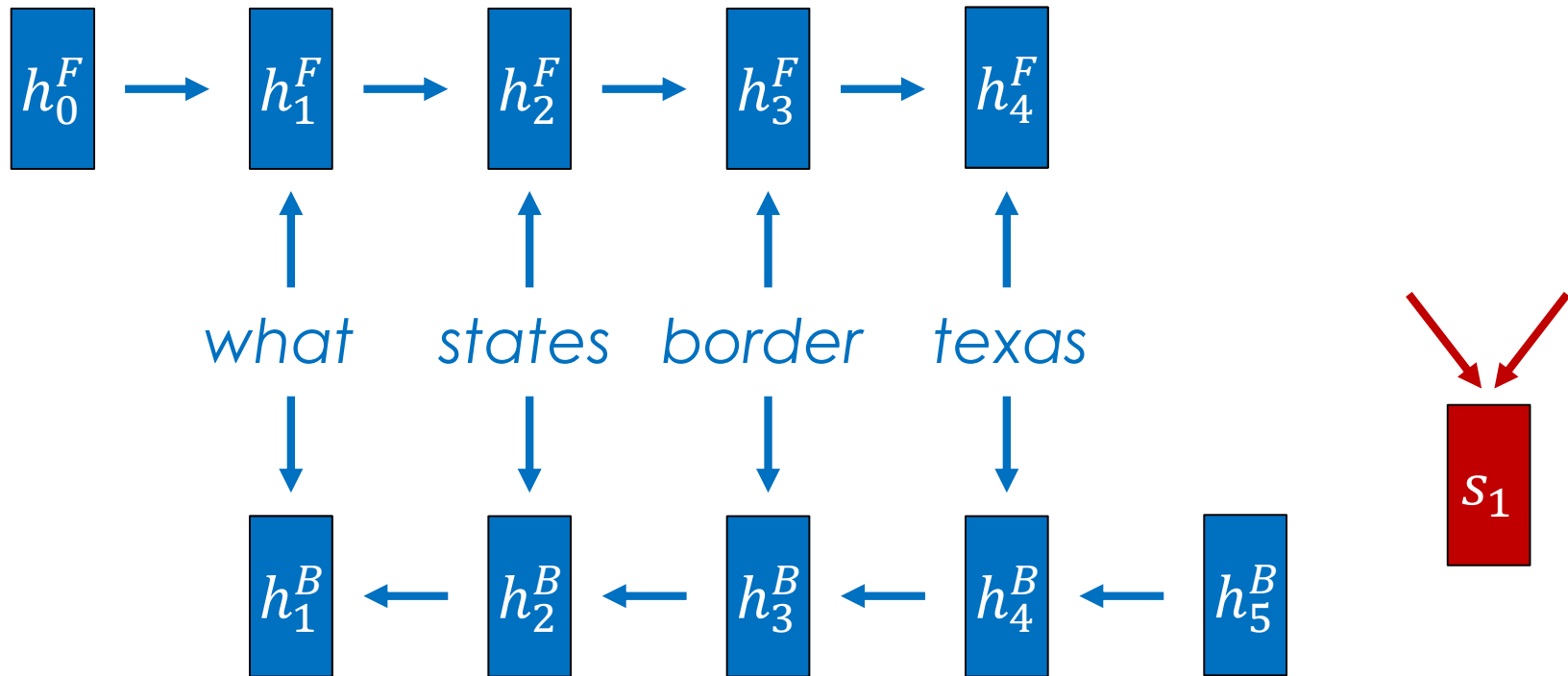
$b_2$

$b_3$

$b_4$



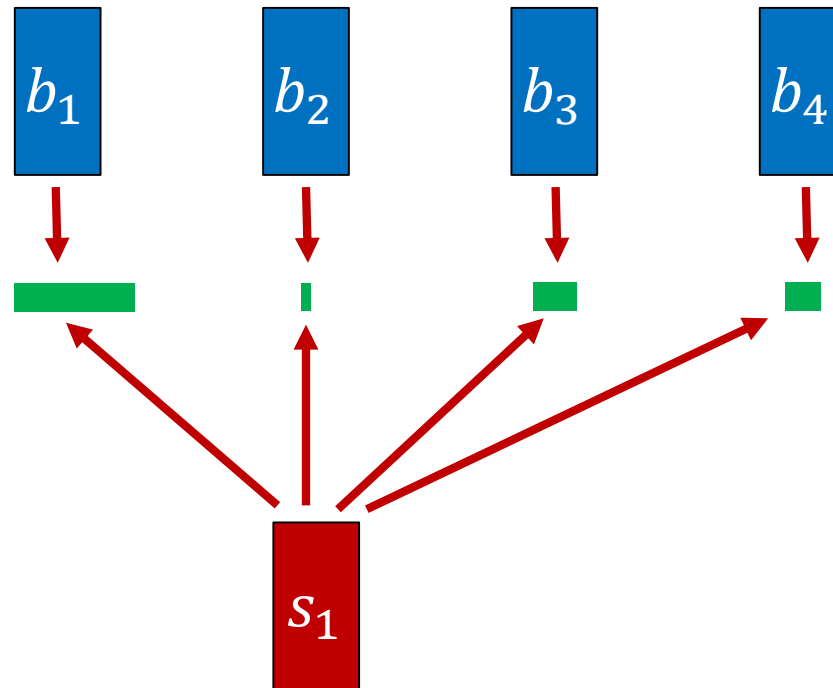
# Neural Semantic Parser





# Neural Semantic Parser

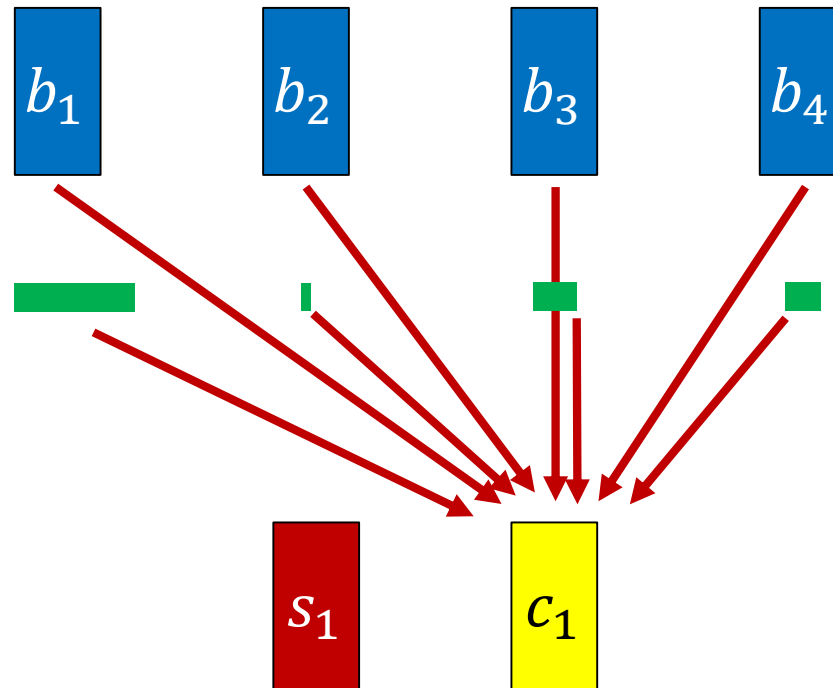
*what states border texas*





# Neural Semantic Parser

*what states border texas*





# Neural Semantic Parser

*what states border texas*

$b_1$

$b_2$

$b_3$

$b_4$

■

■

■

■

$s_1$

$c_1$

$y_1 =$

(

)

And

City

State

iowa

utah

...

■

■

■

■

■

■

■



# Neural Semantic Parser

*what states border texas*

$b_1$

$b_2$

$b_3$

$b_4$

■

■

■

■

$s_1$

$c_1$

$y_1 =$

(

)

And

City

State

iowa

utah

...

■

■

■

■

■

■

■

■





# Neural Semantic Parser

---

*what states border texas*

$b_1$

$b_2$

$b_3$

$b_4$

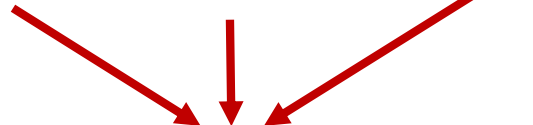


$s_1$

$c_1$

$y_1 = \text{And}$

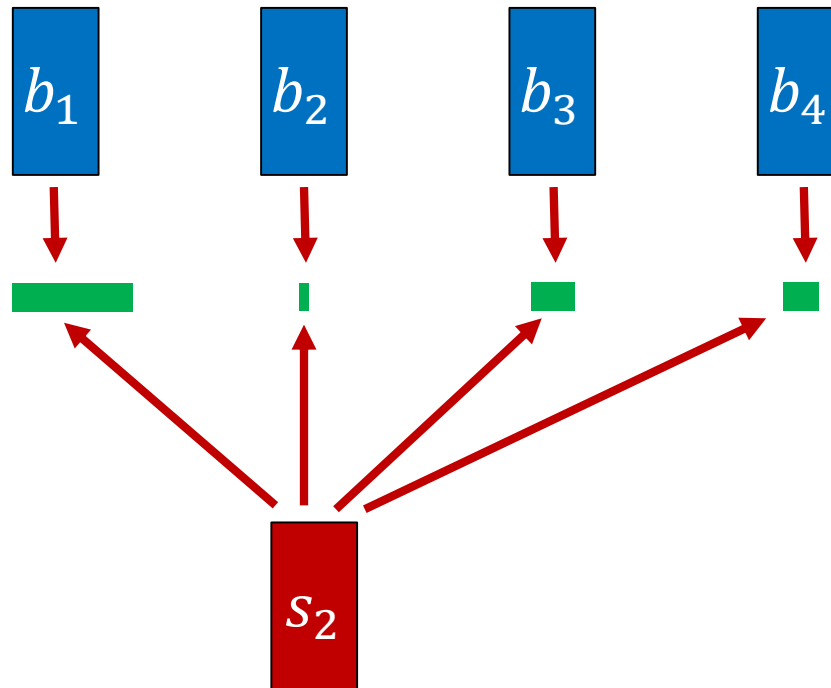
$s_2$





# Neural Semantic Parser

*what states border texas*



$y =$  And



# Rare Entities

---

city('alabama','al','birmingham',284413). city('california','ca','san diego',875538).  
city('alabama','al','mobile',200452). city('california','ca','san  
city('alabama','al','montgomery',177857). francisco',678974).  
city('alabama','al','huntsville',142513). city('california','ca','san jose',629442).  
city('alabama','al','tuscaloosa',75143). city('california','ca','long beach',361334).  
city('alaska','ak','anchorage',174431). city('california','ca','oakland',339337).  
city('arizona','az','phoenix',789704). city('california','ca','sacramento',275741).  
city('arizona','az','tucson',330537). city('california','ca','anaheim',219311).  
city('arizona','az','mesa',152453). city('california','ca','fresno',218202).  
city('arizona','az','tempe',106919). city('california','ca','santa ana',203713).  
city('arizona','az','glendale',96988). city('california','ca','riverside',170876).  
city('arizona','az','scottsdale',88622). city('california','ca','huntington  
city('arkansas','ar','little rock',158915). beach',170505).  
city('arkansas','ar','fort smith',71384). city('california','ca','stockton',149779).  
city('arkansas','ar','north little rock',64388). city('california','ca','glendale',139060).  
city('california','ca','los angeles',2966850). city('california','ca','fremont',131945).  
...



# Rare Entities

---

- Rare entities pose a problem!
  - If we see an entity name we didn't see during training, how can we even generate the right logical form?
- Solution: **Attention-based copying**



# Neural Semantic Parser

*what states border texas*

$b_1$

$b_2$

$b_3$

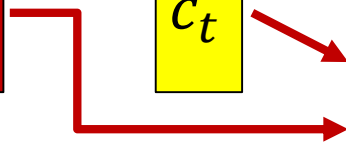
$b_4$



$y =$  And ( State , NextTo ( StateId (

$S_t$

$C_t$

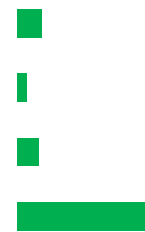


$y_t =$

(  
)  
And  
City  
State  
iowa  
utah  
...



what  
states  
border  
texas





# Neural Semantic Parser

*what states border texas*

$b_1$

$b_2$

$b_3$

$b_4$

$y =$  And (  
State ,  
NextTo (  
StateId (

$s_1$

$c_1$

$y_1 =$

(  
)  
And  
City  
State  
iowa  
utah  
...

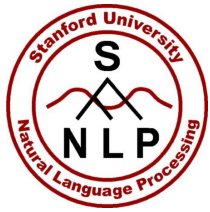
what  
states  
border  
texas



# Implementation Details

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- Training: maximize loglikelihood of correct logical form with SGD
- Test time
  - Decode with beam search
  - Add missing parentheses
  - Prune logical forms that result in execution error



# Results (Take 1)

System	Geoquery	ATIS	Overnight
Zettlemoyer and Collins (2007)		<b>84.6</b>	
Kwiatkowski et al. (2010)	88.9		
Liang et al. (2011)*	<b>91.1</b>		
Kwiatkowski et al. (2013)	<b>89.0</b>		
Zhao and Huang (2015)	88.9	84.2	
Wang et al. (2015)			58.8
<b>Our RNN Model</b>	85.0	76.3	<b>75.8</b>

How can we do better?

\* Not directly comparable; used a seed lexicon for predicates.





# Outline

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- Neural Semantic Parser
- Data Recombination
- Discussion



# Structural Regularities

---

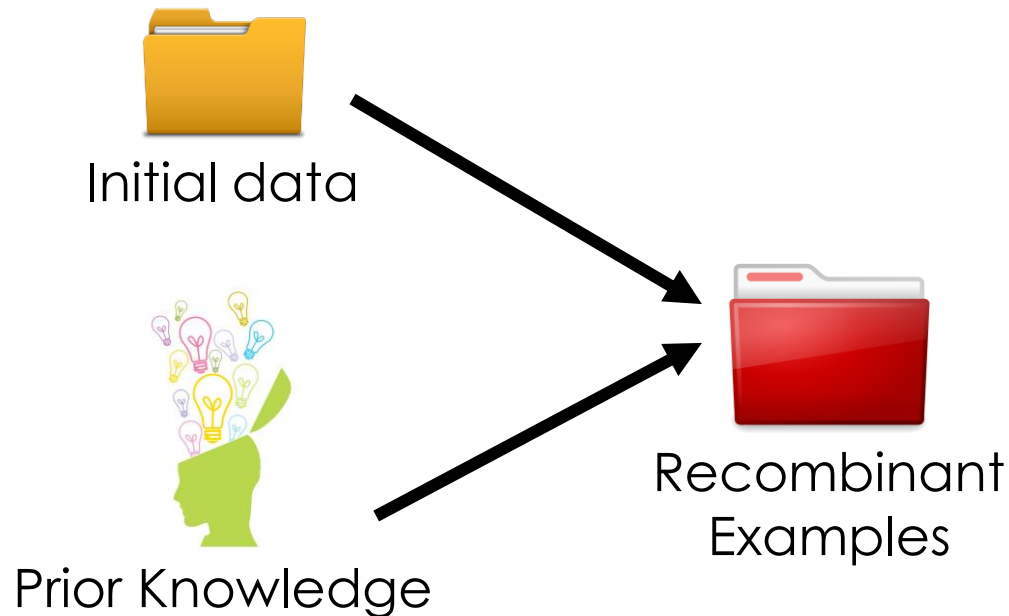
- Given
  - *what are the major cities in iowa ?*  
`And(City, Major, LocatedIn(StateId(iowa)))`
- We know how to parse
  - *what are the major cities in **texas** ?*  
`And(City, Major, LocatedIn(StateId(texas)))`



# Structural Regularities

---

- How do we build a neural model that respects compositional structural regularities?





# Recombinant Examples

---

Dataset:

*what are the major cities in iowa ?*

`And(City, Major, LocatedIn(StateId( iowa )))`

*what are states that border texas ?*

`And(State, NextTo(StateId( texas )))`



# Recombinant Examples

---

*what are the major cities in iowa ?*

`And(City, Major, LocatedIn(StateId(iowa)))`

Step 1: Apply high-precision alignment rules



# Recombinant Examples

---

*what are the major cities in iowa ?*

And(City, Major, LocatedIn(**StateId(iowa)**))

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments



# Recombinant Examples

---

ROOT → (*what are the major cities in* STATE ?,  
And(City, Major, LocatedIn(STATE)))

STATE → (*iowa*, StateId(*iowa*))

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments

Step 3: Generate grammar rules by abstracting aligned fragments



# Recombinant Examples

---

ROOT  $\rightarrow$  (*what are states that border* STATE *?*,  
And(State, NextTo(STATE)))

STATE  $\rightarrow$  (*texas*, StateId(*texas*))

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments

Step 3: Generate grammar rules by abstracting aligned fragments

(Repeat for other examples)





# Recombinant Examples

---

*what are the major cities in* STATE ?

**And(City, Major, LocatedIn(STATE))**

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments

Step 3: Generate grammar rules by abstracting aligned fragments

Step 4: Combine rules from different examples to form new “recombinant” examples



# Recombinant Examples

---

*what are the major cities in texas ?*

`And(City, Major, LocatedIn(StateId(texas)))`

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments

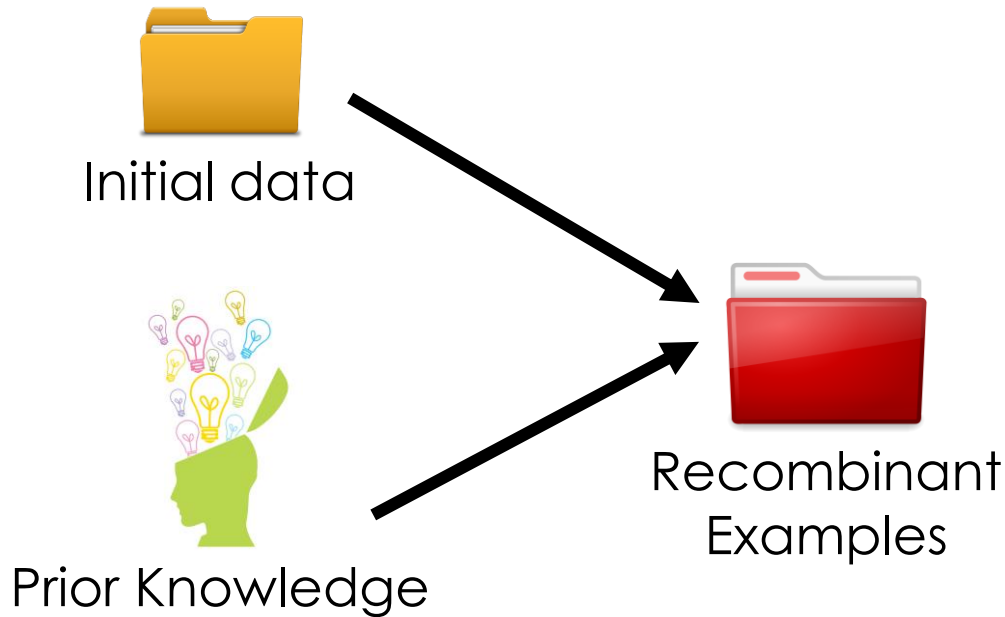
Step 3: Generate grammar rules by abstracting aligned fragments

Step 4: Combine rules from different examples to form new “recombinant” examples



# Data Recombination

---

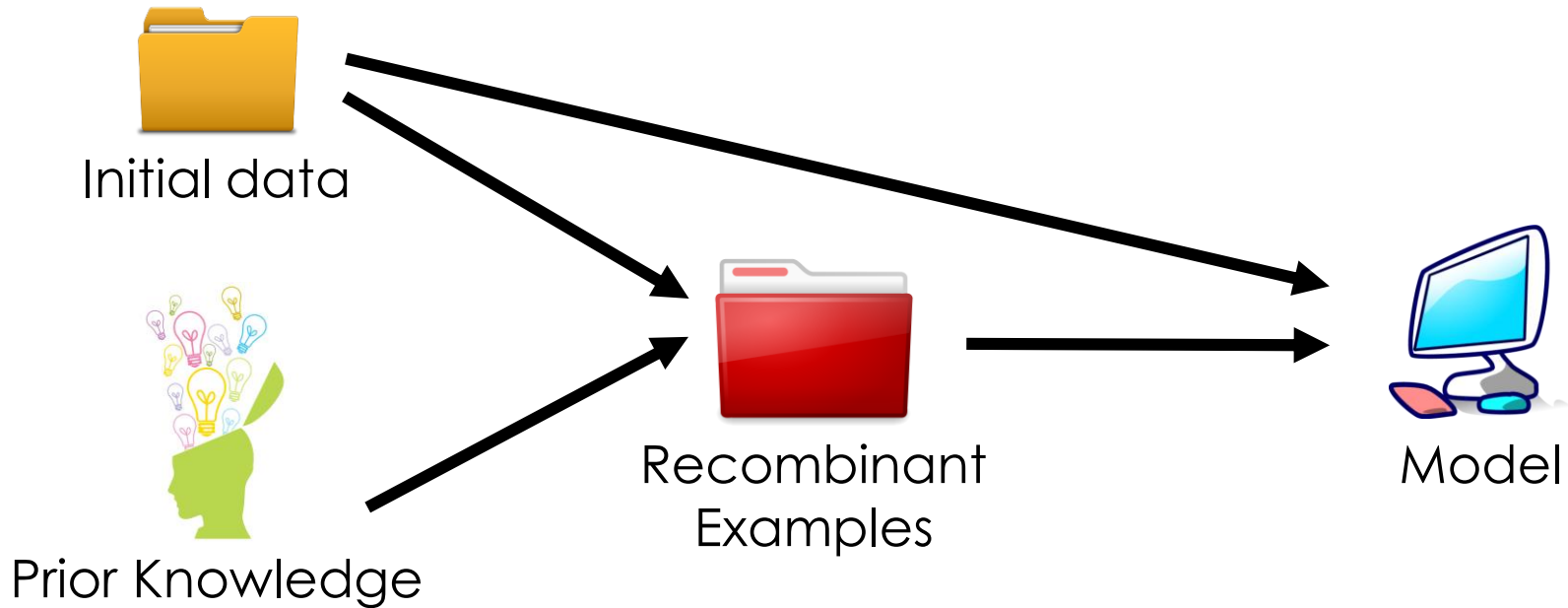


Step 1: **Generate recombinant examples** using training data and prior knowledge about domain



# Data Recombination

---



Step 2: **Train on these examples** to increase model's awareness of task structure



# Training the Model

---

- At each epoch, sample new recombinant examples from grammar
  - Choose production rules uniformly at random
- Train on recombinant examples plus original training examples in 1:1 ratio



# Results (Take 2)

System	Geoquery	ATIS	Overnight
Zettlemoyer and Collins (2007)		<b>84.6</b>	
Kwiatkowski et al. (2010)	88.9		
Liang et al. (2011)*	<b>91.1</b>		
Kwiatkowski et al. (2013)	<b>89.0</b>		
Zhao and Huang (2015)	88.9	84.2	
Wang et al. (2015)			58.8
Our RNN Model	85.0	76.3	<b>75.8</b>
+ Abstracting Entities	85.4	79.9	75.3

\* Not directly comparable; used a seed lexicon for predicates.



# Structural Regularities, Part 2

---

- Given
  - *what are the major cities in iowa ?*  
`And(City, Major, LocatedIn(StateId(iowa)))`
  - *what are states that border texas ?*  
`And(State, NextTo(StateId(texas)))`
- We know how to parse
  - *what are the major cities in  
states that border texas ?*  
`And(City, Major, LocatedIn(  
And(State, NextTo(StateId(texas))))))`



# Abstracting Whole Phrases

---

What are *states that border texas* ?

```
And(State, NextTo(StateId(texas)))
```

Step 1: Apply high-precision alignment rules





# Abstracting Whole Phrases

---

What are *states that border texas* ?

```
And(State, NextTo(StateId(texas)))
```

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments



# Abstracting Whole Phrases

---

What are *states that border texas* ?

**And(State, NextTo(StateId(texas)))**

Step 1: Apply high-precision alignment rules

Step 2: Infer types of aligned fragments

New Grammar Rule

- STATE → (*states that border texas*,  
**And(State, NextTo(StateId(texas)))**)



# Results (Take 3)

System	Geoquery	ATIS	Overnight
Zettlemoyer and Collins (2007)		<b>84.6</b>	
Kwiatkowski et al. (2010)	88.9		
Liang et al. (2011)*	<b>91.1</b>		
Kwiatkowski et al. (2013)	<b>89.0</b>		
Zhao and Huang (2015)	88.9	84.2	
Wang et al. (2015)			58.8
Our RNN Model	85.0	76.3	75.8
+ AbsEntities	85.4	79.9	75.3
+ AbsWholePhrases	87.5	---	<b>75.9</b>

\* Not directly comparable; used a seed lexicon for predicates.



# Composition of Strategies

---

- Grammar induction strategies are functions on grammars
  - Process each rule in grammar independently to generate new grammar rules
- Initial grammar
  - ROOT → (*what are the major cities in iowa ?*,  
And(City, Major, LocatedIn(StateId(iowa))))
  - ROOT → (*what are states that border texas ?*,  
And(State, NextTo(StateId(texas))))



# Composition of Strategies

---

Initial Gramar



Abstract Whole Phrases



AbsWholePhrases Grammar



# Composition of Strategies

---

STATE →

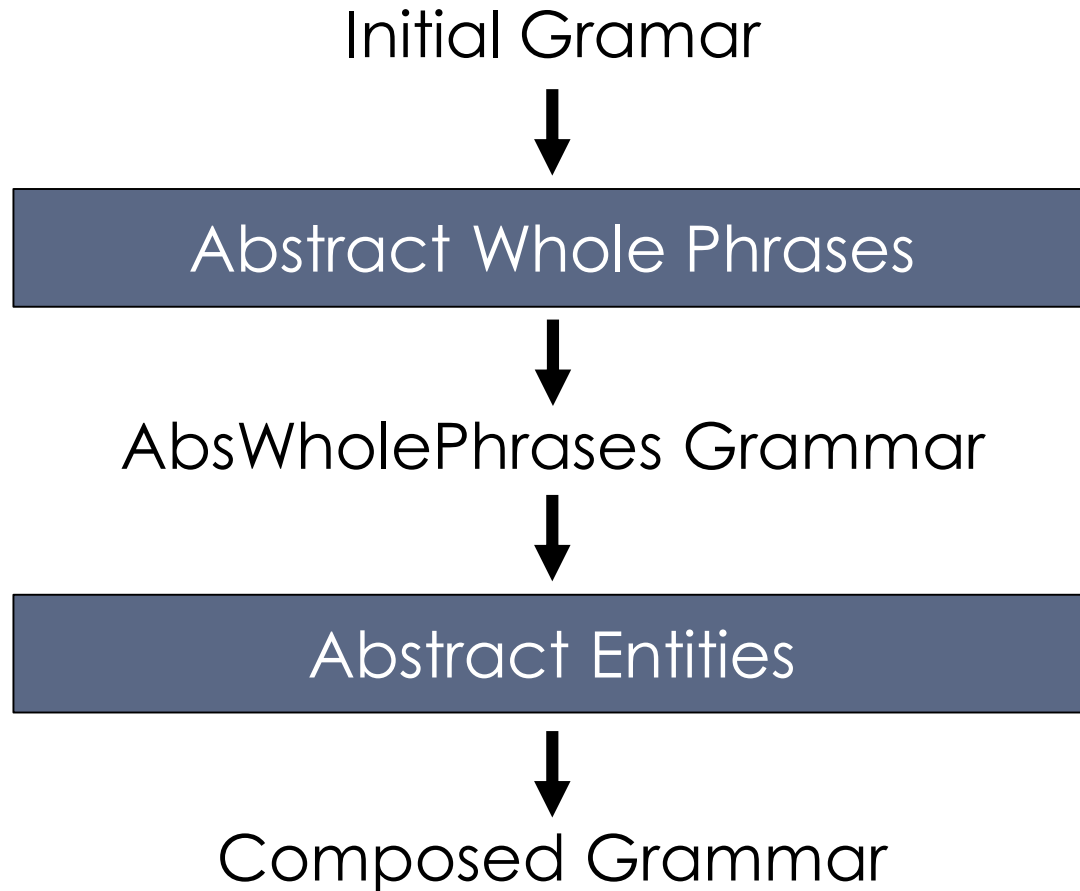
*(states that border texas ,*

*And(State, NextTo(StateId(texas))))*



# Composition of Strategies

---





# Results (Take 4)

System	Geoquery	ATIS	Overnight
Zettlemoyer and Collins (2007)		<b>84.6</b>	
Kwiatkowski et al. (2010)	88.9		
Liang et al. (2011)*	<b>91.1</b>		
Kwiatkowski et al. (2013)	<b>89.0</b>		
Zhao and Huang (2015)	88.9	84.2	
Wang et al. (2015)			58.8
Our RNN Model	85.0	76.3	75.8
+ AbsEntities	85.4	79.9	75.3
+ AbsWholePhrases	87.5	---	<b>75.9</b>
+ AbsWholePhrases, AbsEntities	88.9	---	75.3

\* Not directly comparable; used a seed lexicon for predicates.





# What else can we do?

---

- What happens if we artificially make longer examples?



# Concatenation

---

- Given
  - *what are the major cities in iowa ?*  
`And(City, Major, LocatedIn(StateId(iowa)))`
  - *what are states that border texas ?*  
`And(State, NextTo(StateId(texas)))`
- Create new example
  - *what are the major cities in iowa ? </s> what are states that border texas ?*  
`And(City, Major, LocatedIn(StateId(iowa)))`  
`</s> And(State, NextTo(StateId(texas)))`



# Concatenation

---

- Can cast as grammar induction
- Compose with previous grammar induction strategies



# Final Results

System	Geoquery	ATIS	Overnight
Zettlemoyer and Collins (2007)		<b>84.6</b>	
Kwiatkowski et al. (2010)	88.9		
Liang et al. (2011)*	<b>91.1</b>		
Kwiatkowski et al. (2013)	89.0		
Zhao and Huang (2015)	88.9	84.2	
Wang et al. (2015)			58.8
Our RNN Model	85.0	76.3	75.8
+ AbsEntities	85.4	79.9	75.3
+ AbsWholePhrases	87.5	---	75.9
+ AbsWholePhrases, AbsEntities	88.9	---	75.3
+ AbsWholePhrases, AbsEntities, Concat	<b>89.3</b>	<b>83.3</b>	<b>77.5</b>

\* Not directly comparable; used a seed lexicon for predicates.



# Outline

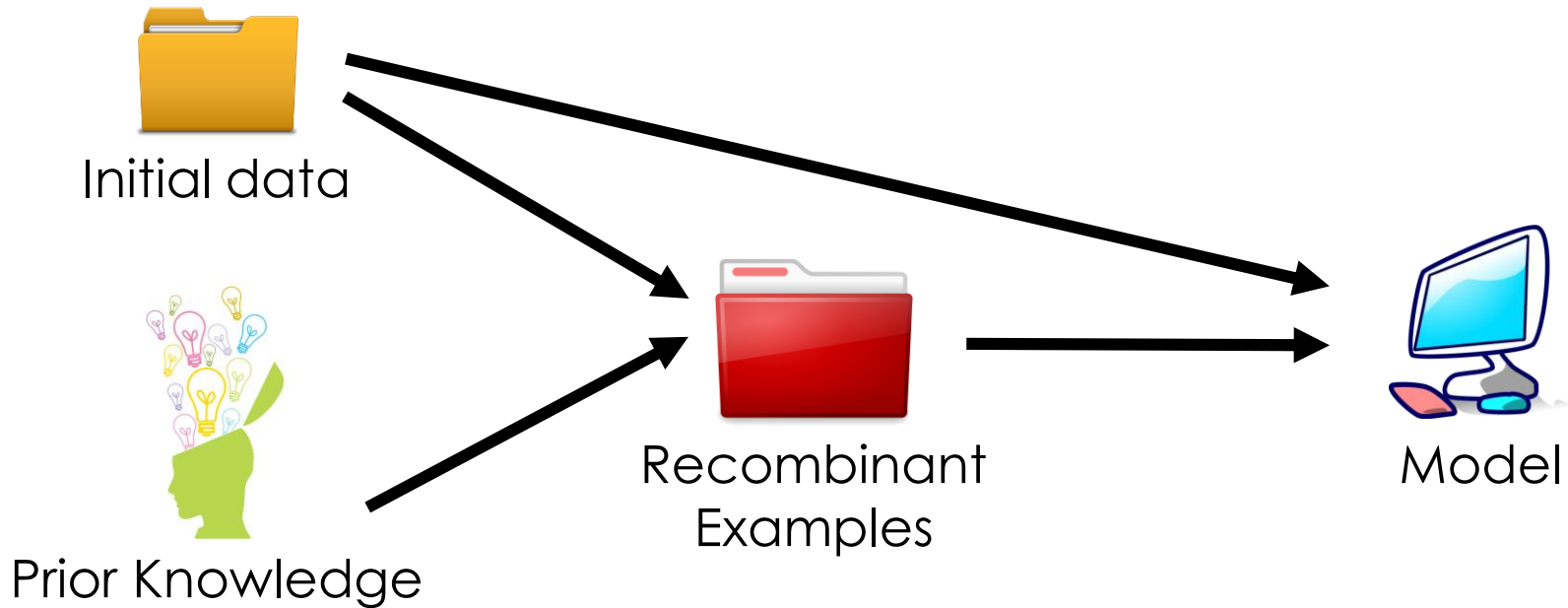
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- Neural Semantic Parser
- Data Recombination
- Discussion



# Data Recombination

---



Step 1: **Generate recombinant examples** using training data and prior knowledge about domain



# Data Recombination

---

- Builds a **generative model** from the data
  - Mixture of the empirical distribution and induced probabilistic grammar
  - Train on examples  $(x, y)$  sampled from this model



# Data Augmentation

---

- Apply local transformations to a single example to generate more examples

$x =$



$y =$  "puppy"





# Data Augmentation

---

- Apply local transformations to a single example to generate more examples

$x =$



$y =$  "puppy"

Horizontal Reflection



# Data Augmentation

---

- Apply local transformations to a single example to generate more examples

$x =$



$y =$  "puppy"

Cropping



# Data Recombination

---

- Get a richer generative model by combining information from **multiple examples**

*what are the major cities in iowa ?*

`And(City, Major, LocatedIn(StateId( iowa )))`

*what are states that border texas ?*

`And(State, NextTo(StateId( texas )))`



# Empirical Distribution

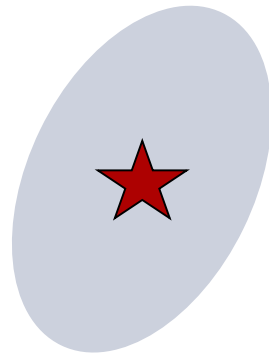
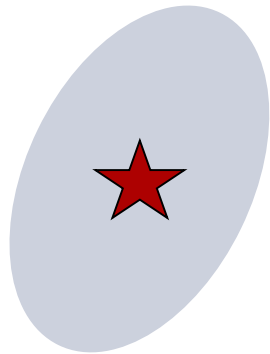
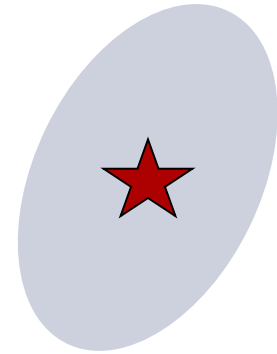
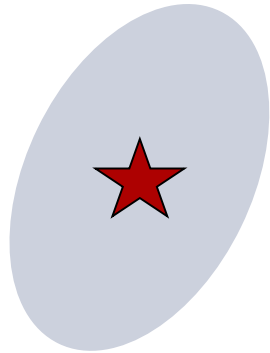
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# Data Augmentation

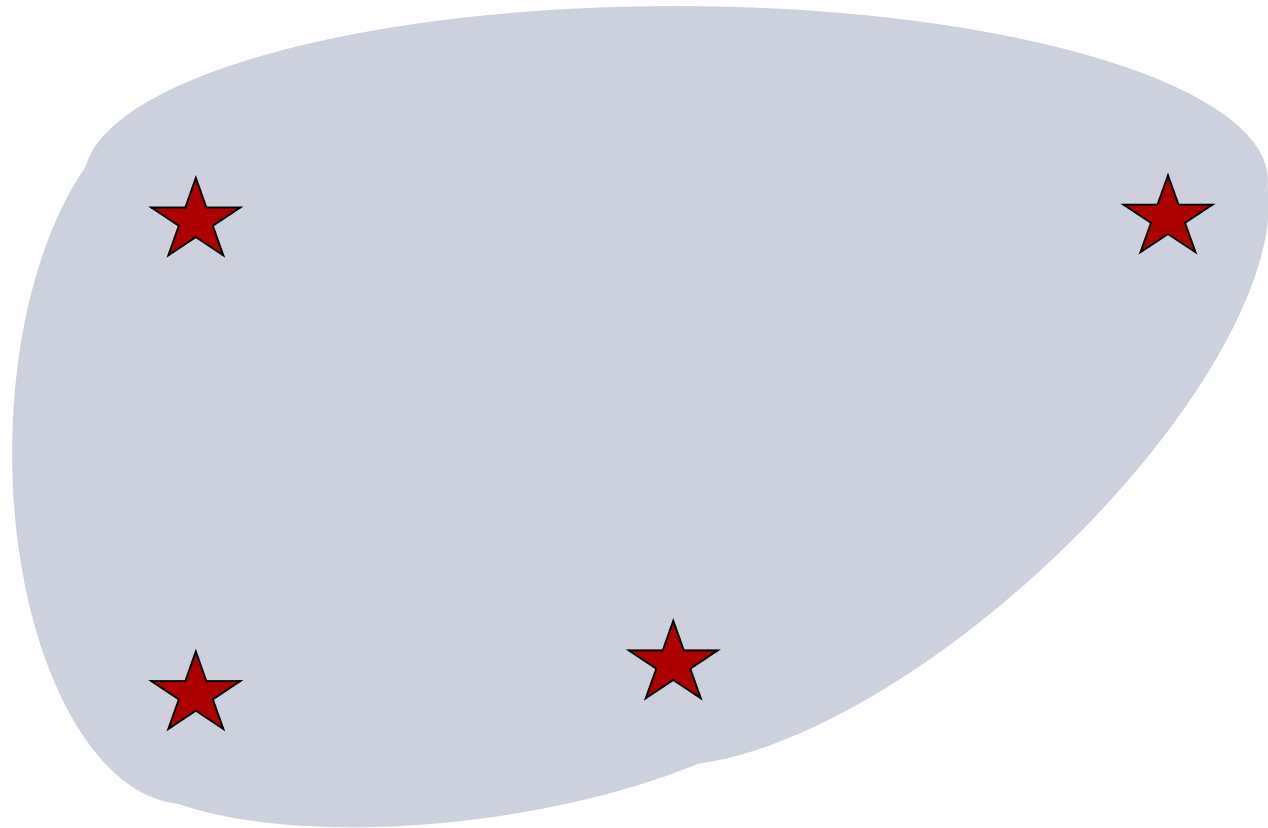
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# Data Recombination

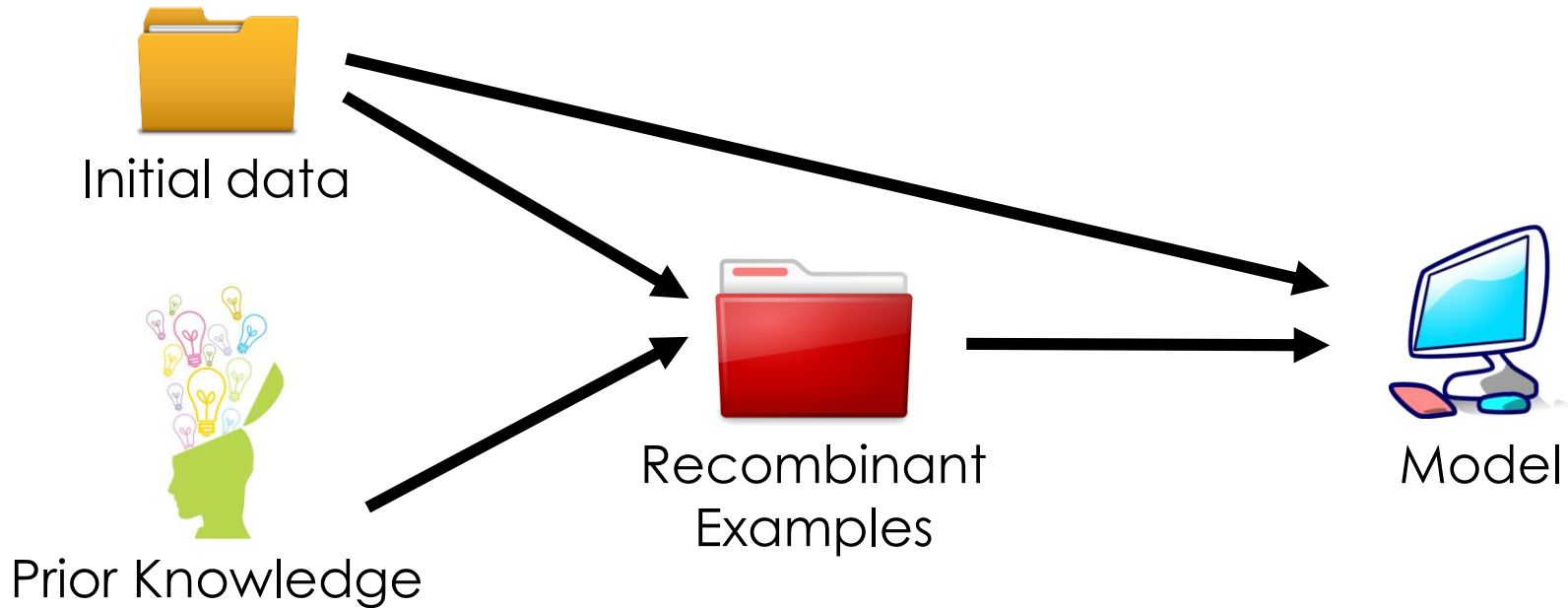
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# Thank you!

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Code, data, and experiments available on CodaLab

<https://worksheets.codalab.org/worksheets/0x50757a37779b485f89012e4ba03b6f4f/>



# Composition of Strategies

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STATE →

*(states that border texas ,*

*And(State, NextTo(StateId(texas)))*)

New Grammar Rule

- STATE → *(states that border* STATEID,  
*And(State, NextTo(STATEID)))*





# An Artificial Experiment

---

- Make up some simple artificial data
  - *friends of relatives of alice*  
`Friend(Relative(alice))`
  - *colleagues of brothers of bob*  
`Colleague(Brother(bob))`
- Generate recombinant examples to train on
  - **Same length:** only swap entities
    - *friends of relatives of bob*  
`Friend(Relative(bob))`
  - **Longer:** nest whole phrases
    - *friends of relatives of colleagues of brothers of alice*  
`Friend(Relative(Colleague(Brother(alice))))`



# An Artificial Experiment

---

- Test only on “short” examples like those in the original training data
  - *siblings of roommates of eve*  
**Sibling(Roommate(eve))**



# An Artificial Experiment

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