

# A Game-Theoretic Approach to Generating Spatial Descriptions

Dave Golland

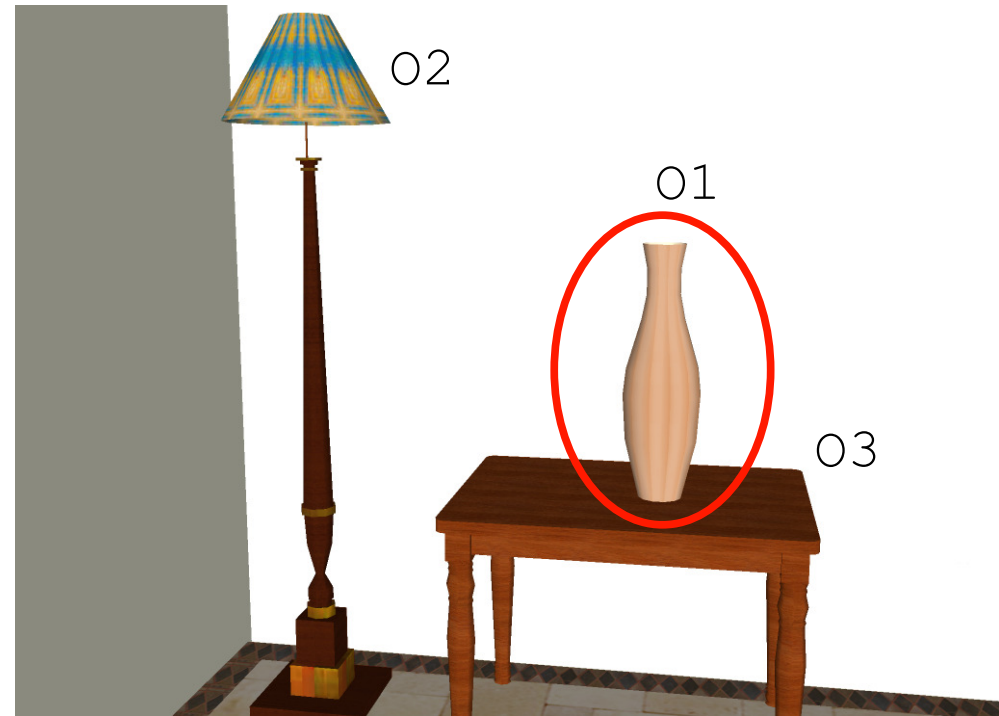
Percy Liang

Dan Klein



# Language is about Communication

Goal: refer to 01

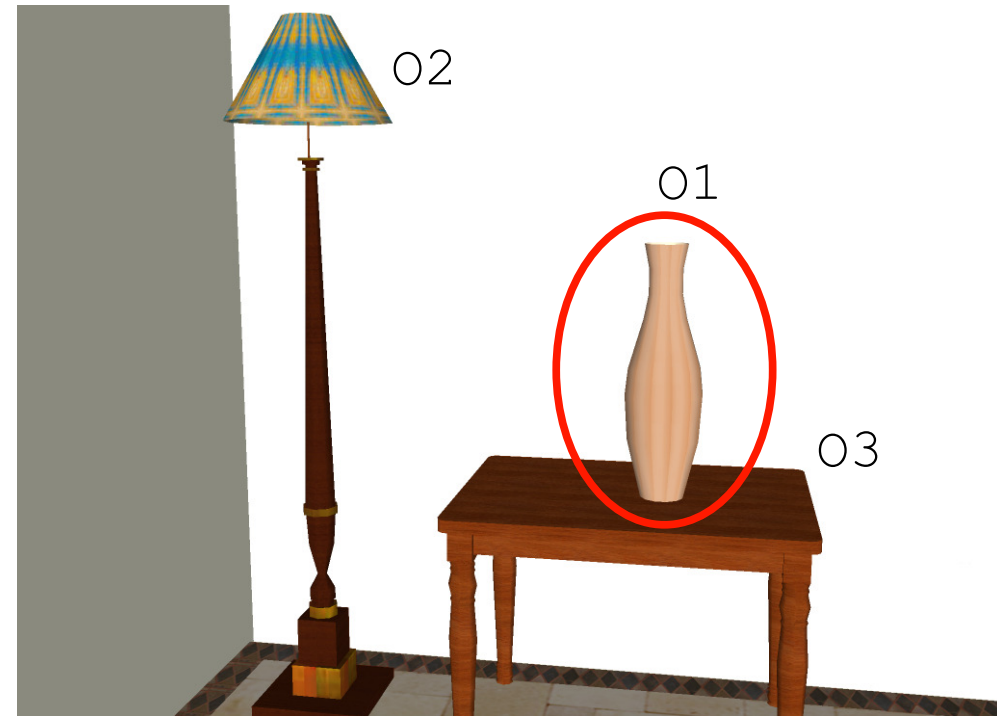


[Grice, 1975]

# Language is about Communication

Goal: refer to 01

Strategy 1: speak the truth  
(Maxim of Quality)



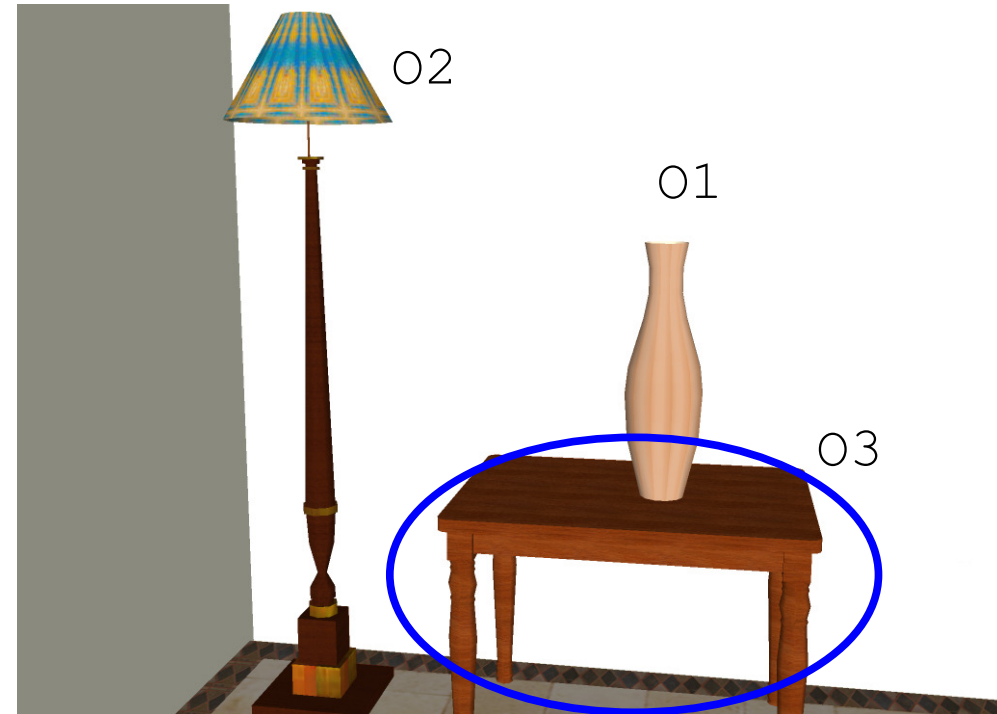
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# Language is about Communication

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*on* 03



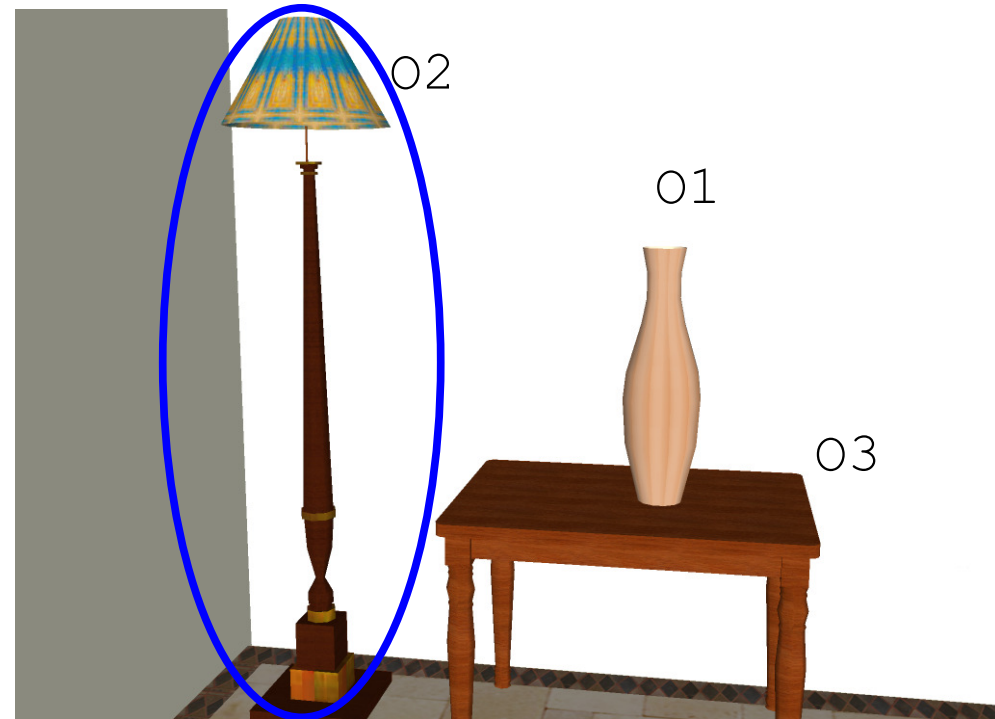
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# Language is about Communication

Goal: refer to 01

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*on 03*      *right of 02*



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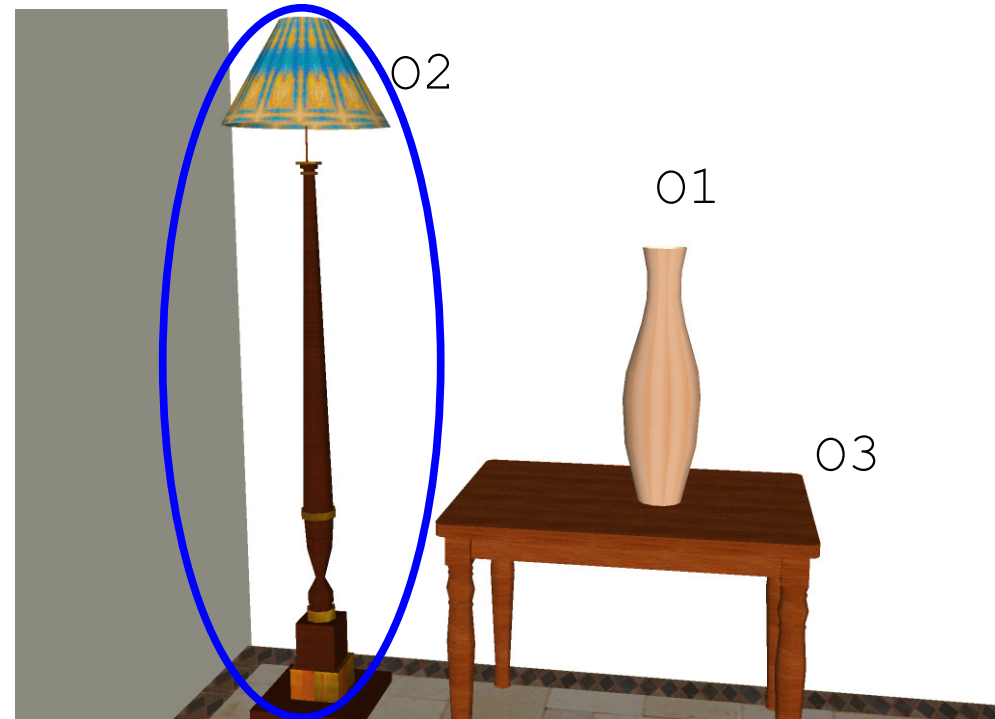
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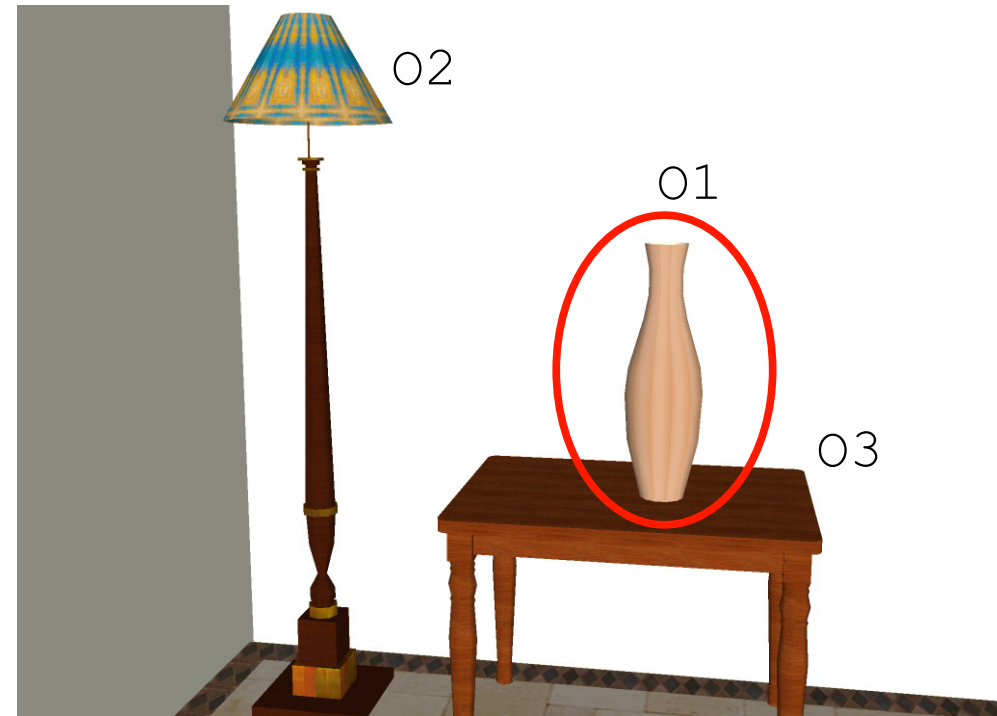
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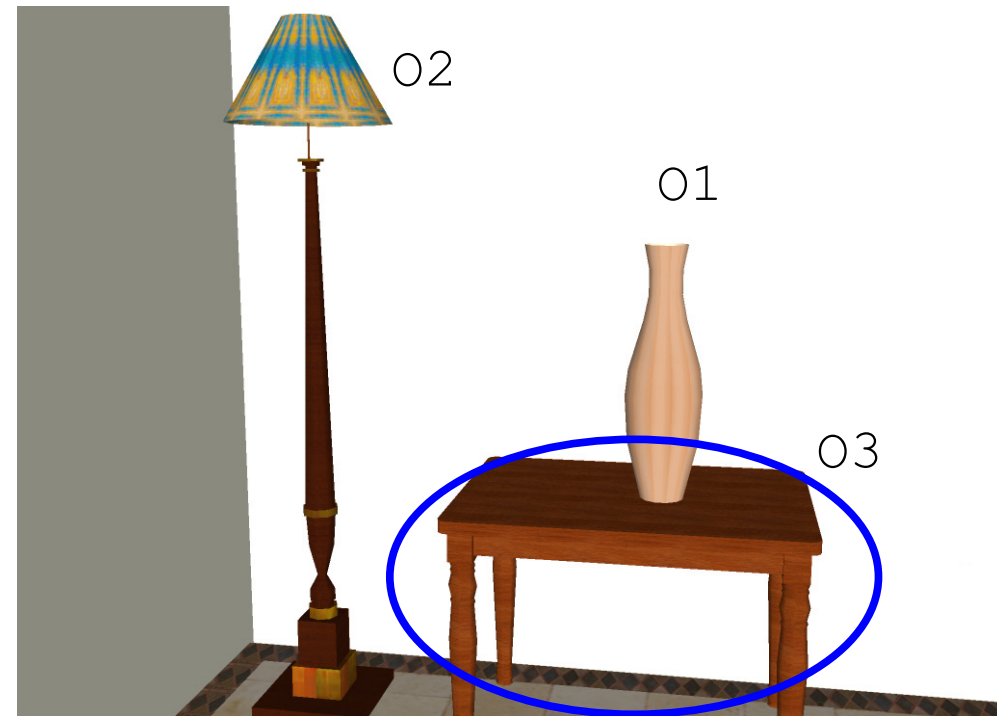
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[Grice, 1975]



# Actual Example



# Language Game

speaker

# Language Game

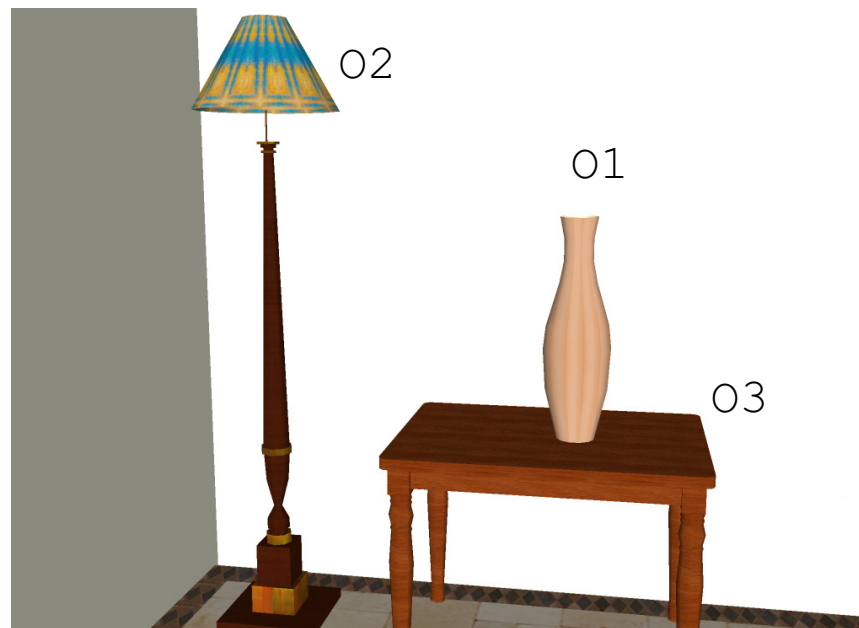
speaker

listener

# Language Game

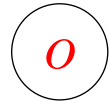
speaker

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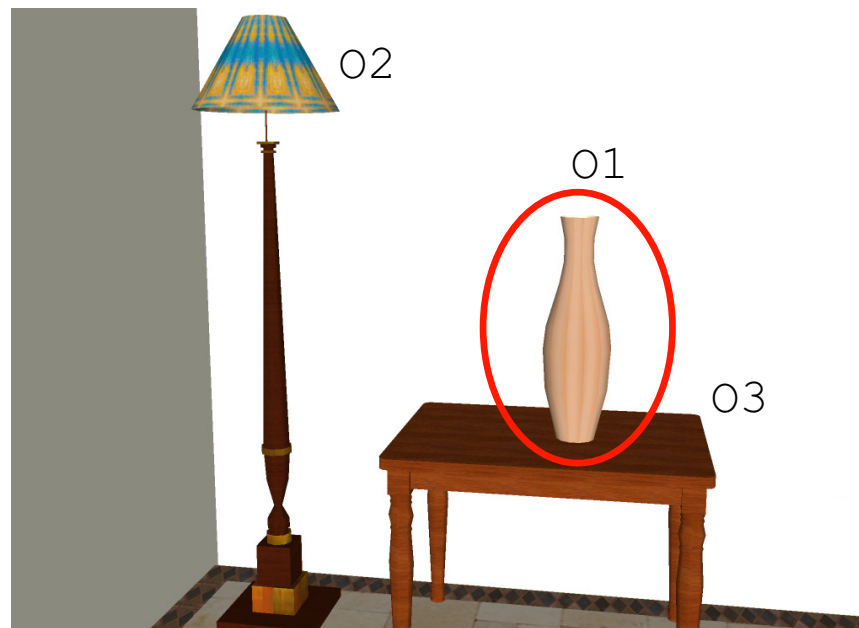
# Language Game

target

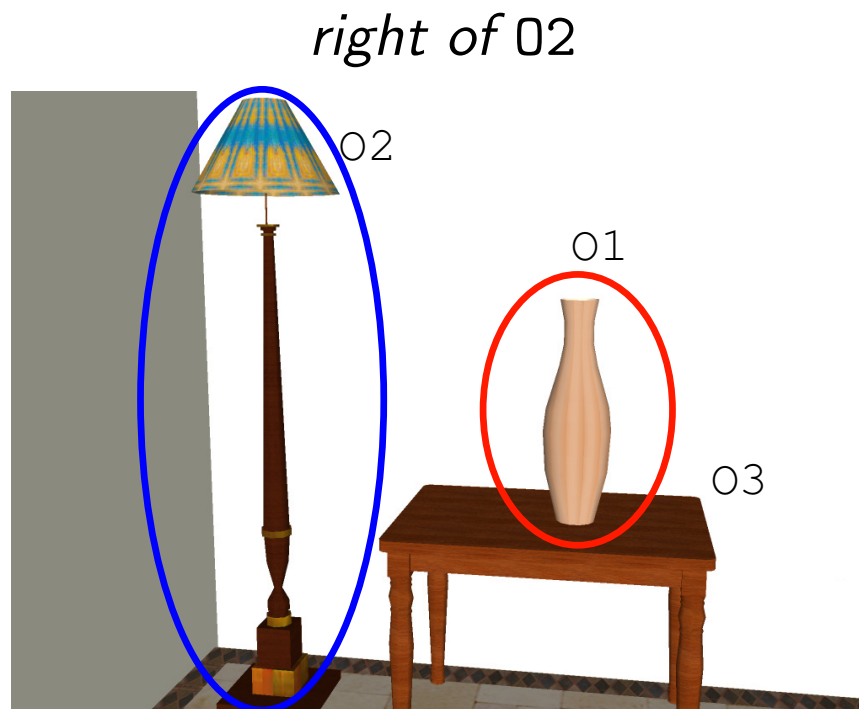
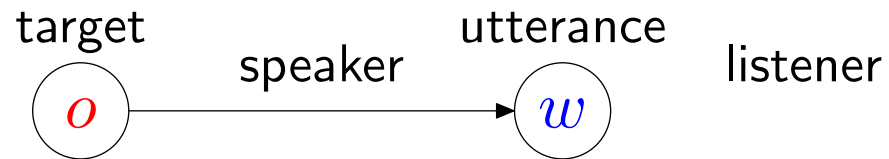


speaker

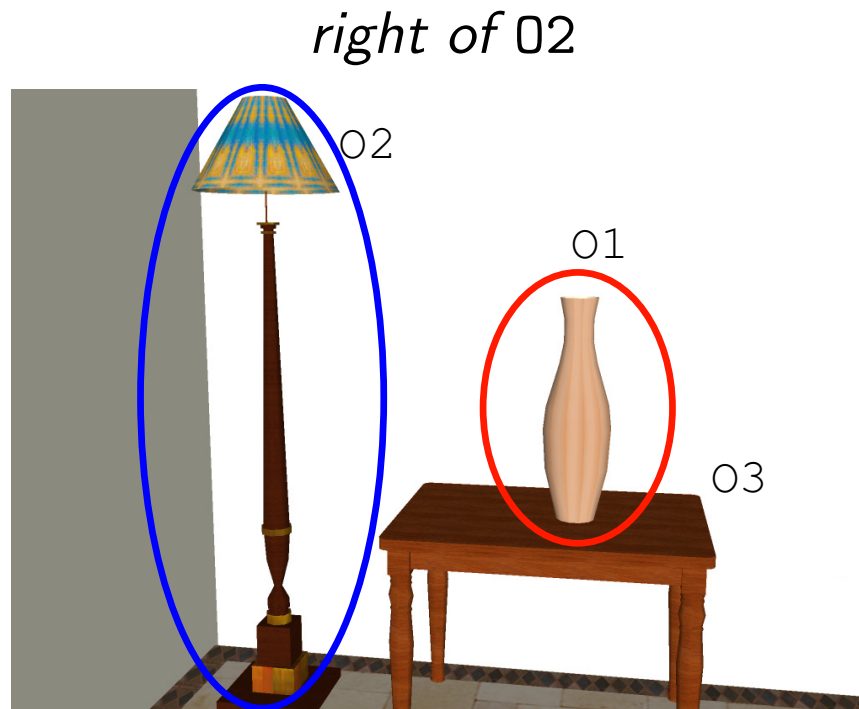
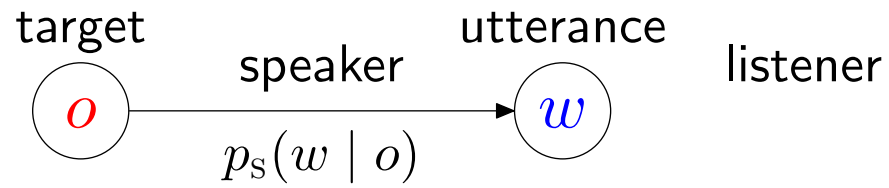
listener



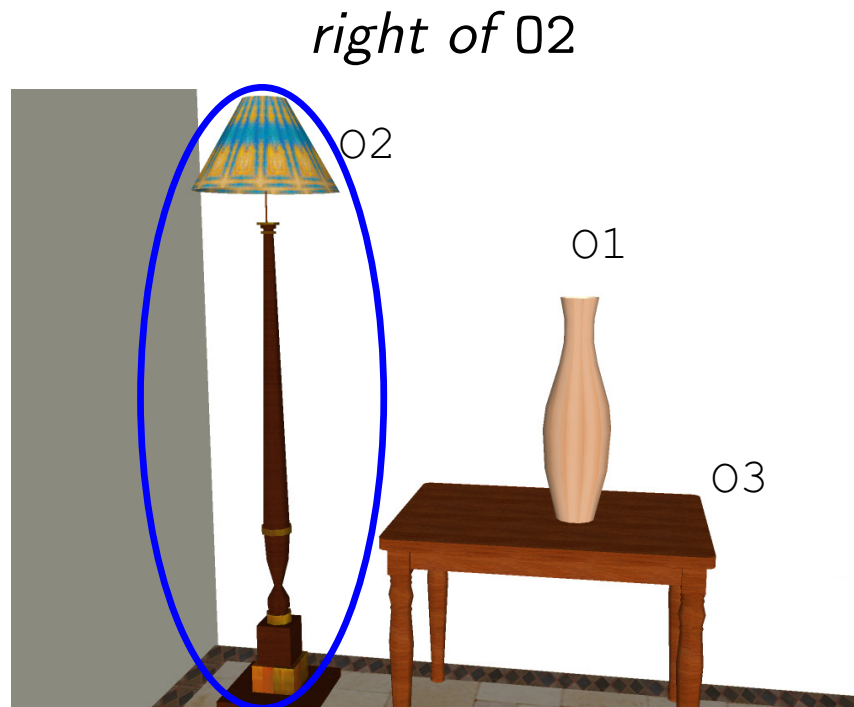
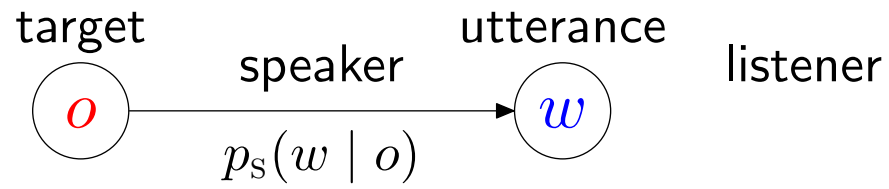
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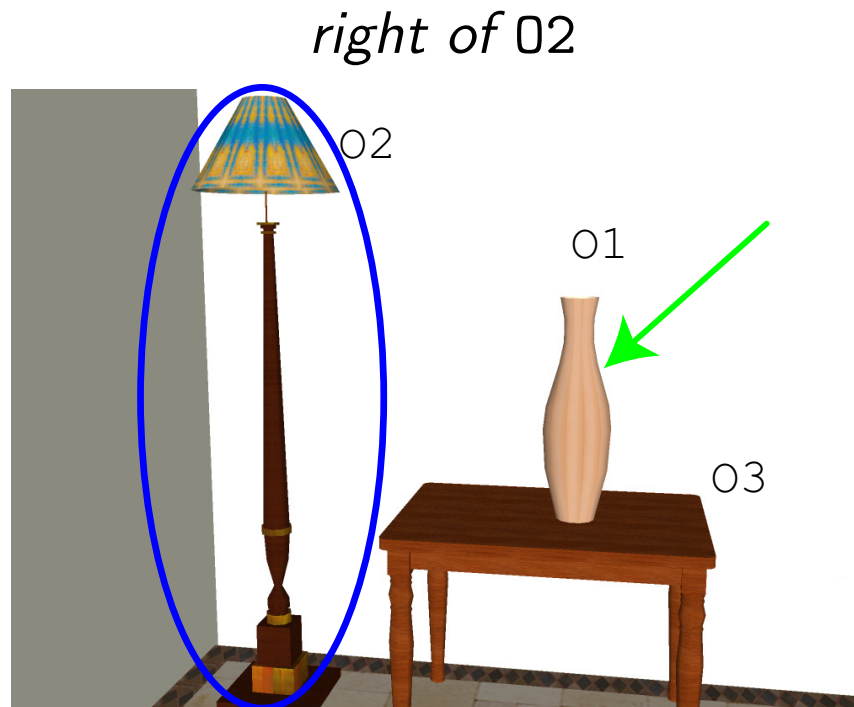
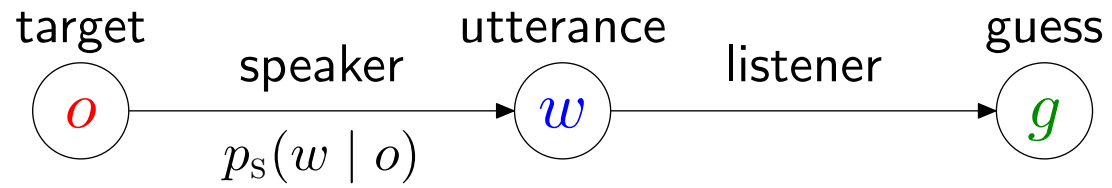


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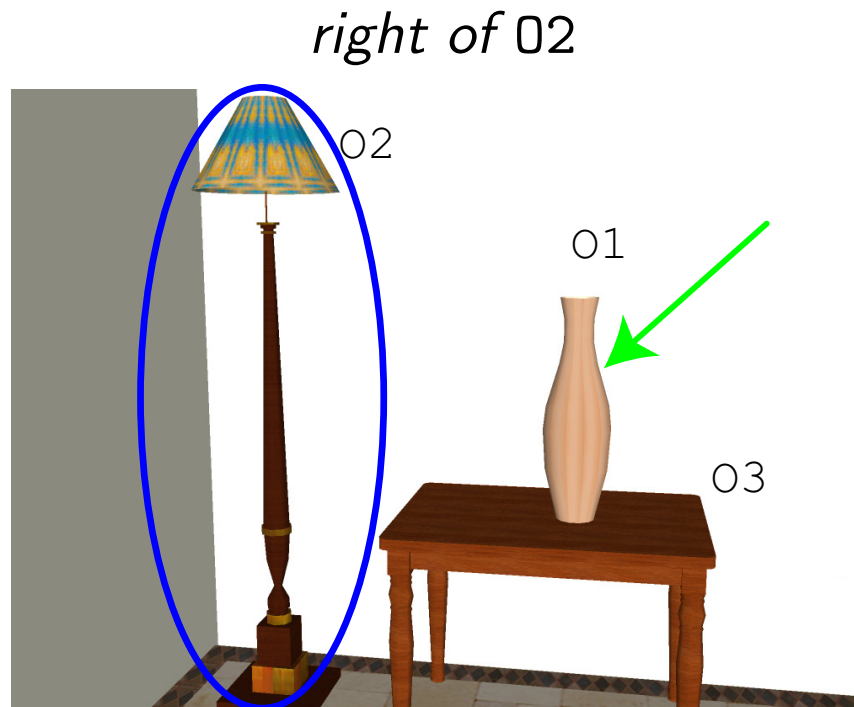
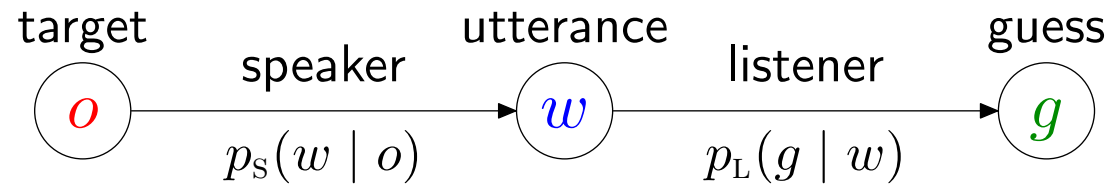




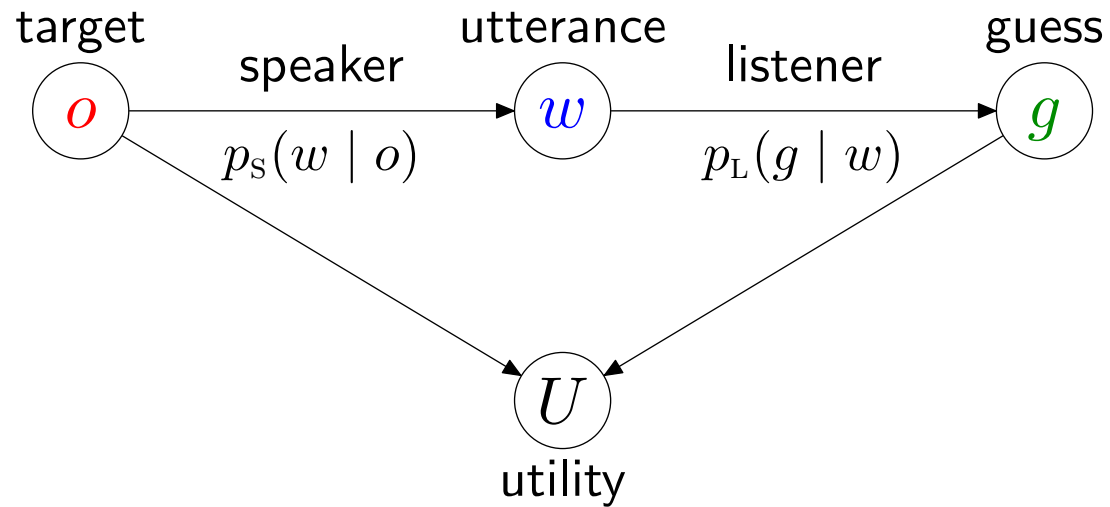
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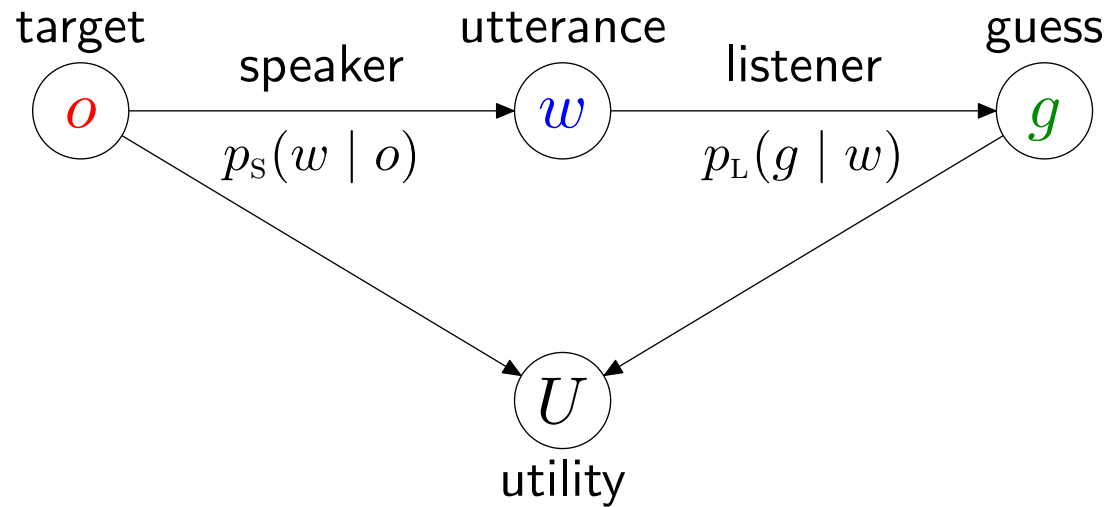


# Language Game



$$U(o, g) = \mathbb{I}[o = g] = \begin{cases} 1 & \text{if } o = g \\ 0 & \text{otherwise} \end{cases}$$

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$$EU(S, L) = \mathbb{E}_{S, L}[U(o, g)]$$

# Speaker Strategies

Assign scores to utterances via:

$$p_s(w|o)$$

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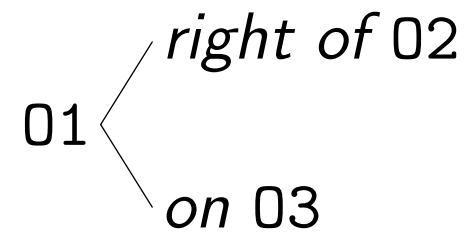
$$p_s(w|o)$$

Two speaker strategies:

1. semantics only (Maxim of Quality)
2. semantics + pragmatics (Maxims of Quality + Manner)

# Semantics Only

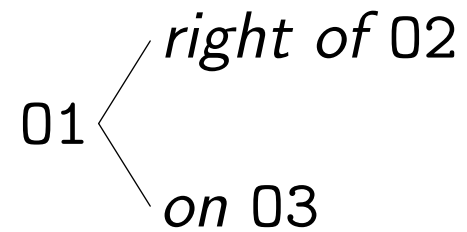
Game tree:





# Semantics Only

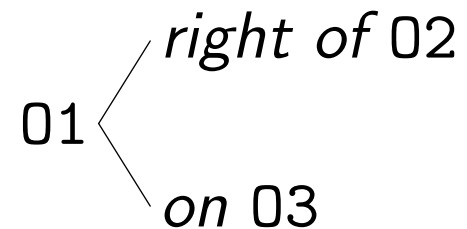
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Reflex speaker because it does not consider consequence of actions.

# Semantics + Pragmatics

Maximize wrt.  $p_S(w|o)$ :

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01  $\left\{ \begin{array}{l} \textit{right} \\ \textit{of 02} \\ \textit{on 03} \end{array} \right.$

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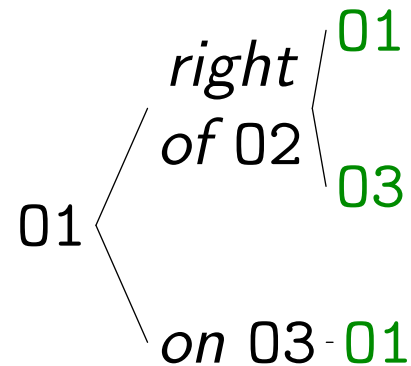
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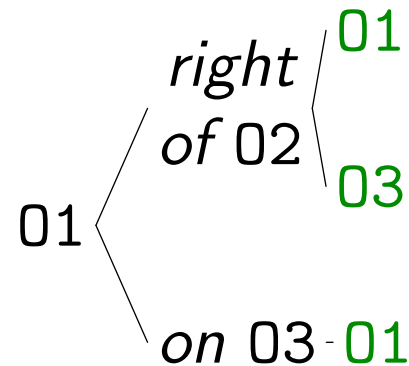
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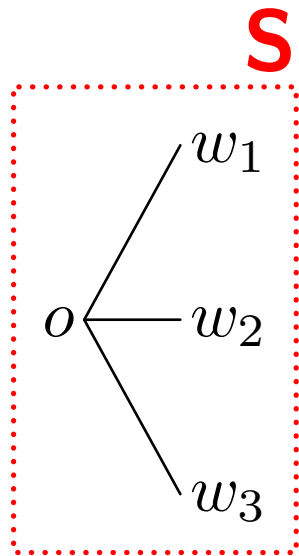
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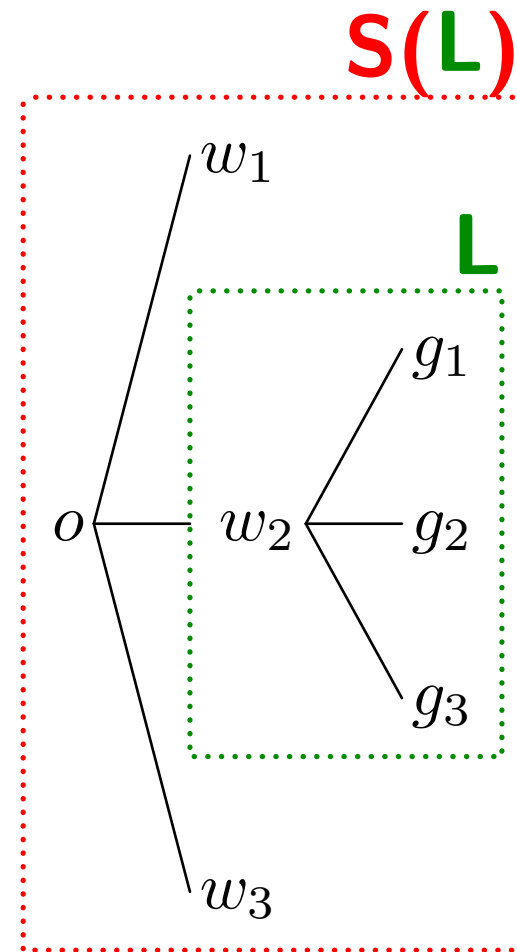
Rational speaker because it is optimal with respect to given listener.



# Reflex vs. Rational



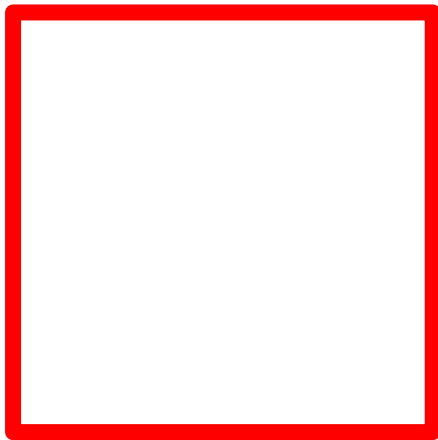
Reflex  
(semantics only)



Rational  
(semantics + pragmatics)

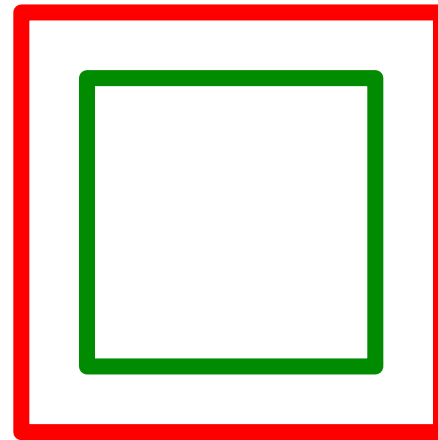
# Reflex vs. Rational

S



Reflex  
(semantics only)

S(L)



Rational  
(semantics + pragmatics)

# Experimental Setup

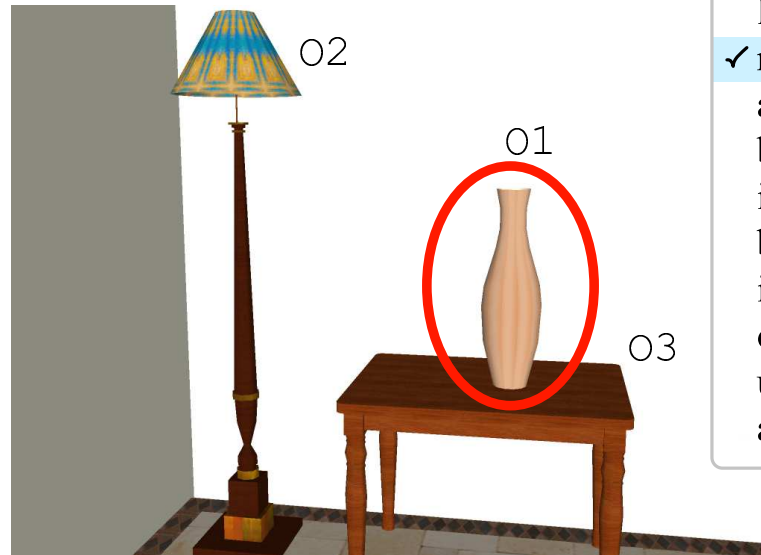
Google sketchup: 43 rooms, average of 22 objects per room



# Data Collection with Mechanical Turk

**Question:** Where is the object outlined in red?

**Answer:** The object outlined in red is



left of  
✓ right of  
above  
below  
in front of  
behind  
inside of  
on  
under  
across from

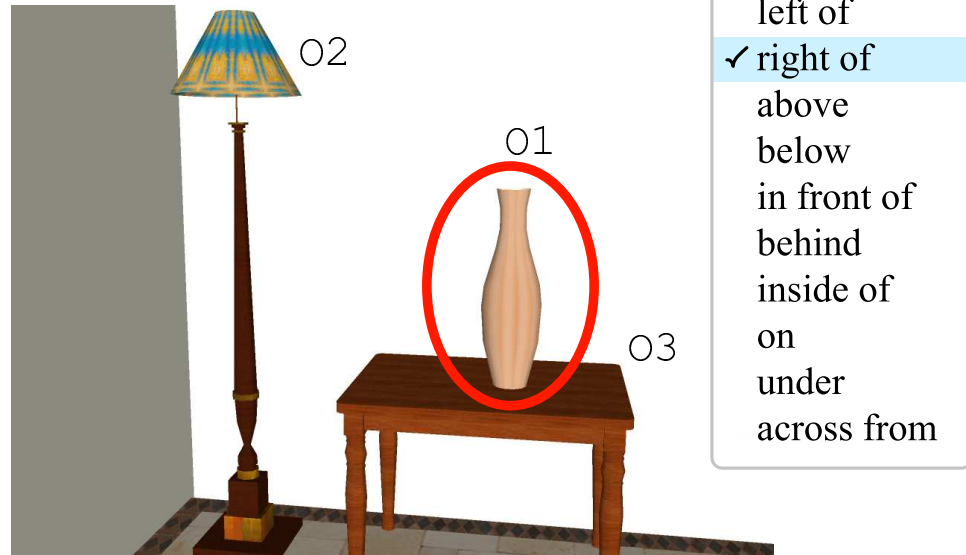
02

Speaker: *o* → *w*

# Data Collection with Mechanical Turk

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**Answer:** The object outlined in red is



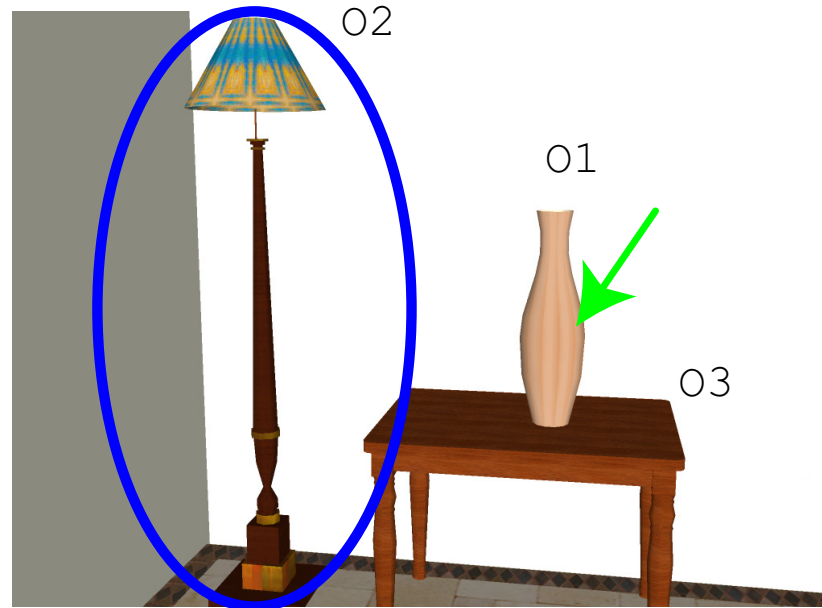
Speaker: *o* → *w*

Yields annotated data:

$$\{(o_1, w_1), \dots, (o_n, w_n)\}$$

# Evaluation with Mechanical Turk

**Question:** What object is **right of** O2 ?



Listener:  $w \rightarrow g$

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compute success metric:

$$\text{SUCCESS}(S) = \frac{1}{n} \sum_i \mathbb{I}[o_i = g_i]$$

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Note: only collecting data and evaluating are done by humans.

# Results

LITERAL agents put mass uniformly on true outputs

	<u>Speaker</u>	<u>Success</u>
Reflex	<sup>S</sup> <input type="checkbox"/> LITERAL	4.6%
Rational	<sup>S(L)</sup> <input type="checkbox"/> LITERAL	33.7%

Rational speaker outperforms reflex speaker.

# Key Points

Rational speaker  $\boxed{S(L)}$  outperforms reflex speaker  $\boxed{S}$

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Up next: extensions for improving the listener model.

# Listener Extensions

- Training a Listener
- Generating Complex Utterances
- Modeling Listener Confusion

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We also train  $p_S(w|o; \theta_S)$  using the same data and features

to get the reflex <sup>S</sup>  LEARNED speaker



# Features

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Features inspired by [Regier, 2001; Tellex, 2009; Landau, 1993]

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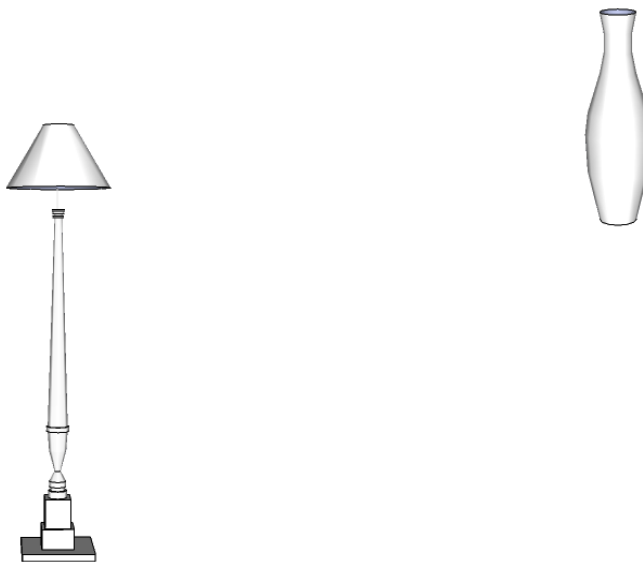


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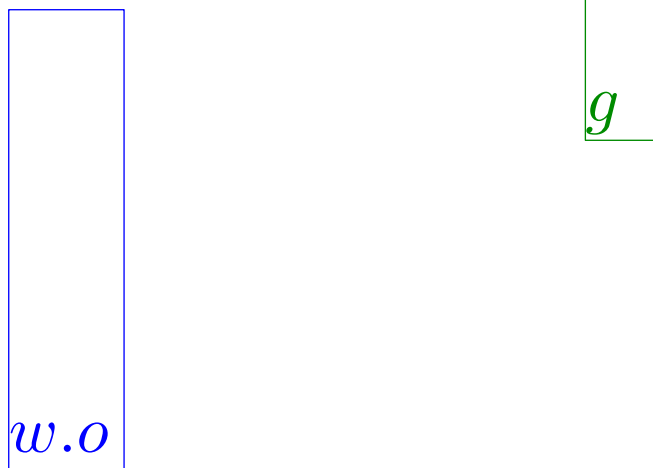


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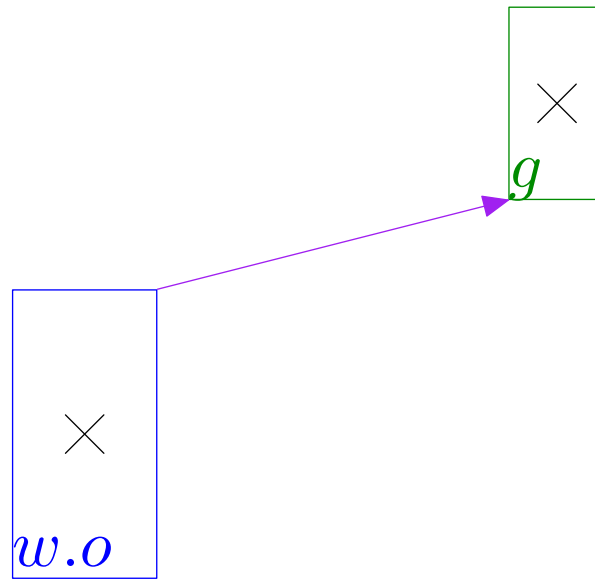
The features  $\phi(g, w)$  are defined between:

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- $g$  and  $w.o$  are bounding boxes

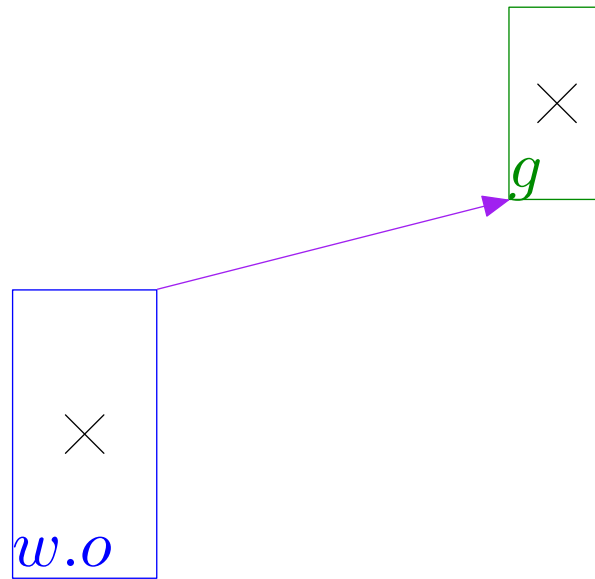


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# Distance Features



# Distance Features



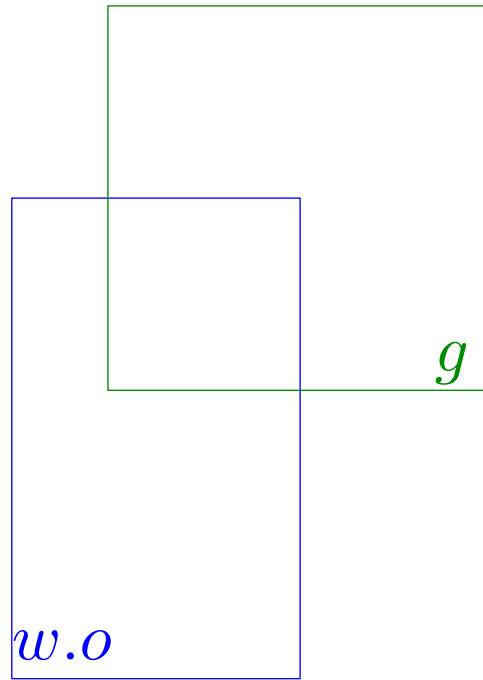
$\phi_{dist}$  = value of shortest distance between  $g$  and  $w.o$

$\phi_{top1}$  =  $\mathbb{I}[g \text{ is closest to } w.o]$

$\phi_{top5}$  =  $\mathbb{I}[g \text{ is among top 5 closest to } w.o]$

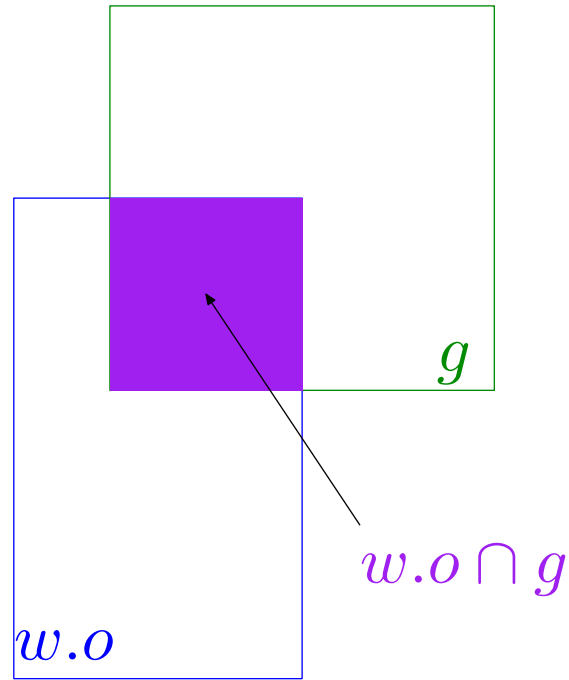
$\phi_{top10}$  =  $\mathbb{I}[g \text{ is among top 10 closest to } w.o]$

# Containment Features





# Containment Features



$$\phi_{cont2} = \text{vol}(w.o \cap g) / \text{vol}(g)$$

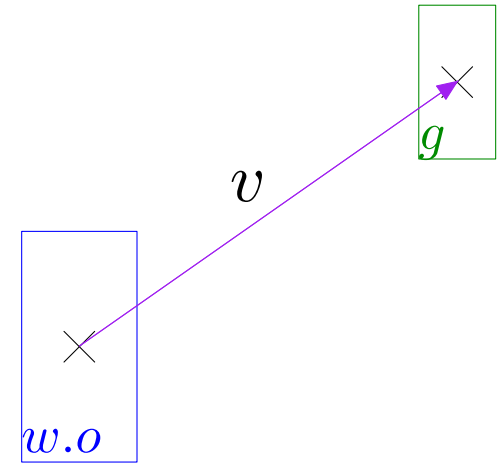
$$\phi_{cont1} = \text{vol}(w.o \cap g) / \text{vol}(w.o)$$

# Projection Features

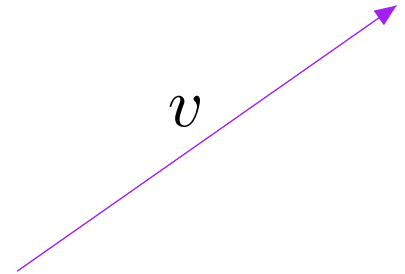
$$\begin{array}{|c|} \hline \times \\ \hline g \end{array}$$

$$\begin{array}{|c|} \hline \times \\ \hline w.o \end{array}$$

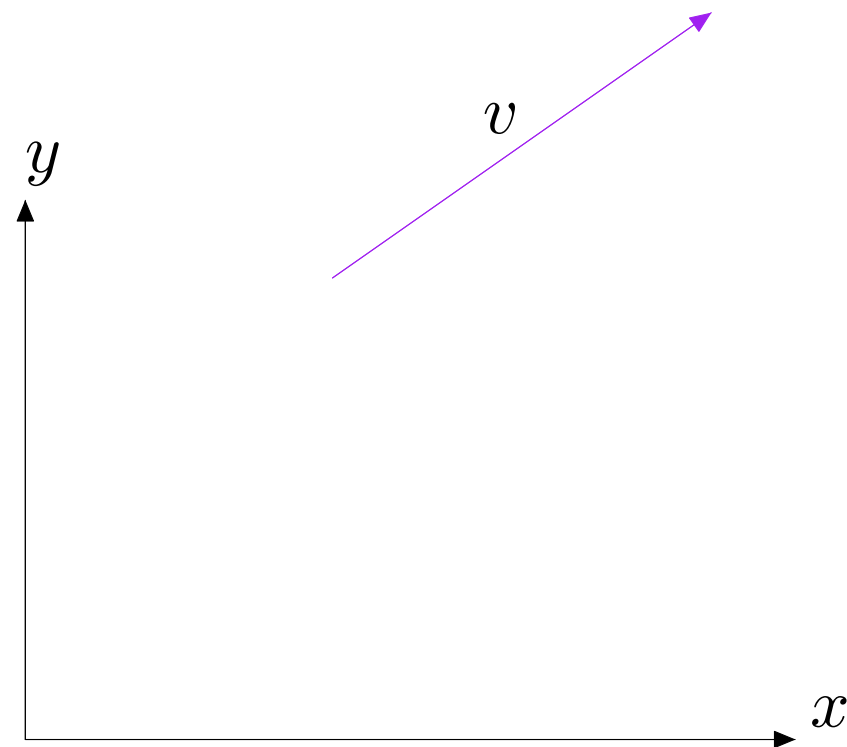
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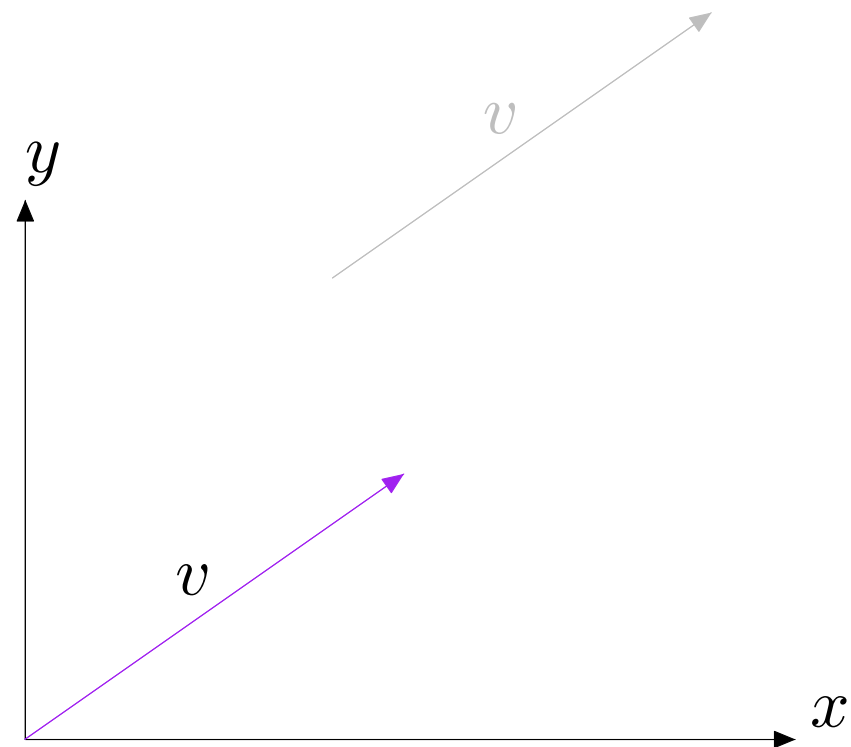
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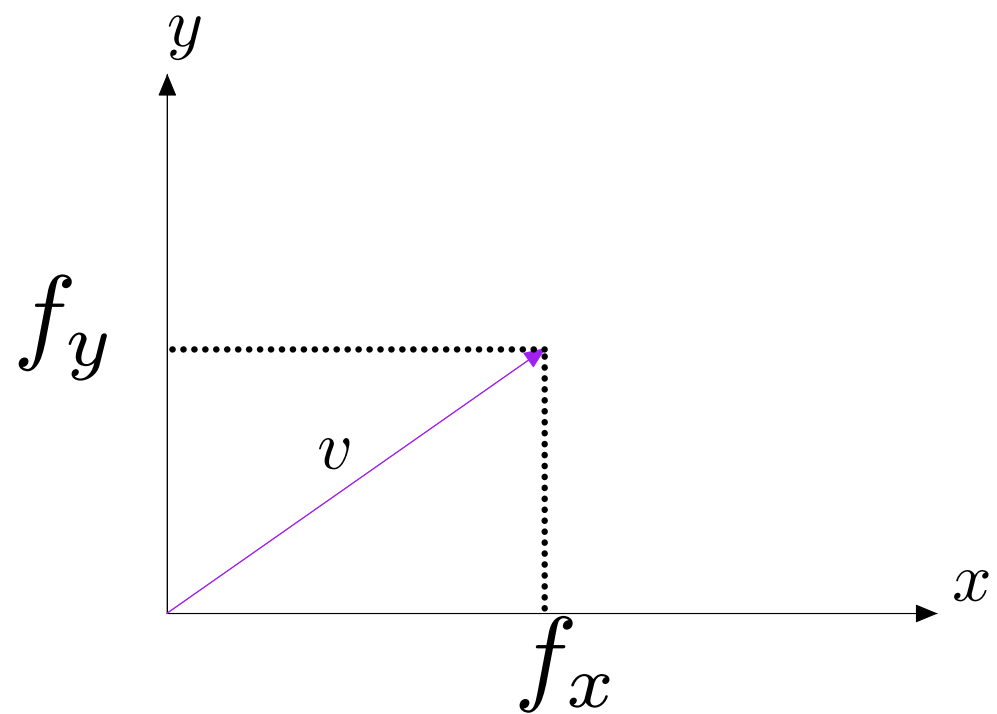
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$$\phi_{projx} = f_x$$

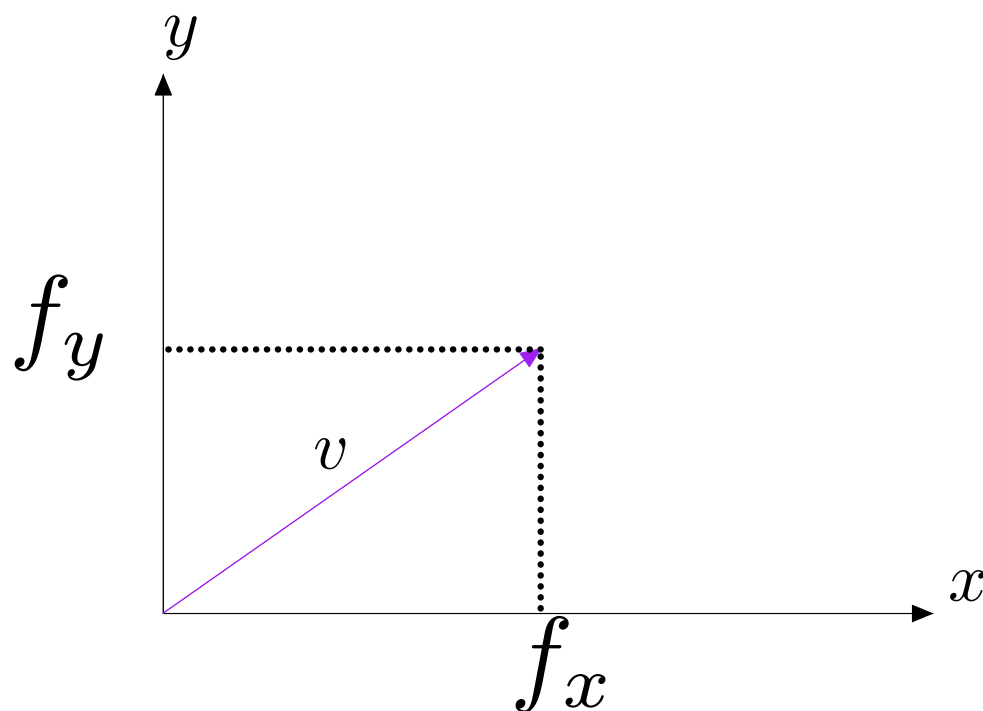
$$\phi_{projy} = f_y$$

$$\phi_{projz} = f_z$$

$$\phi_{proj1} = \mathbb{I}[f_x = \max\{f_x, f_y, f_z\}]$$

$$\phi_{proj2} = \mathbb{I}[f_y = \max\{f_x, f_y, f_z\}]$$

$$\phi_{proj3} = \mathbb{I}[f_z = \max\{f_x, f_y, f_z\}]$$





# Results

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Reflex	<sup>S</sup> <input type="checkbox"/> LEARNED	38.4%
Rational	<sup>S(L)</sup> <input type="checkbox"/> LEARNED	52.6%

# What's missing?

Two things are missing from the setup so far.

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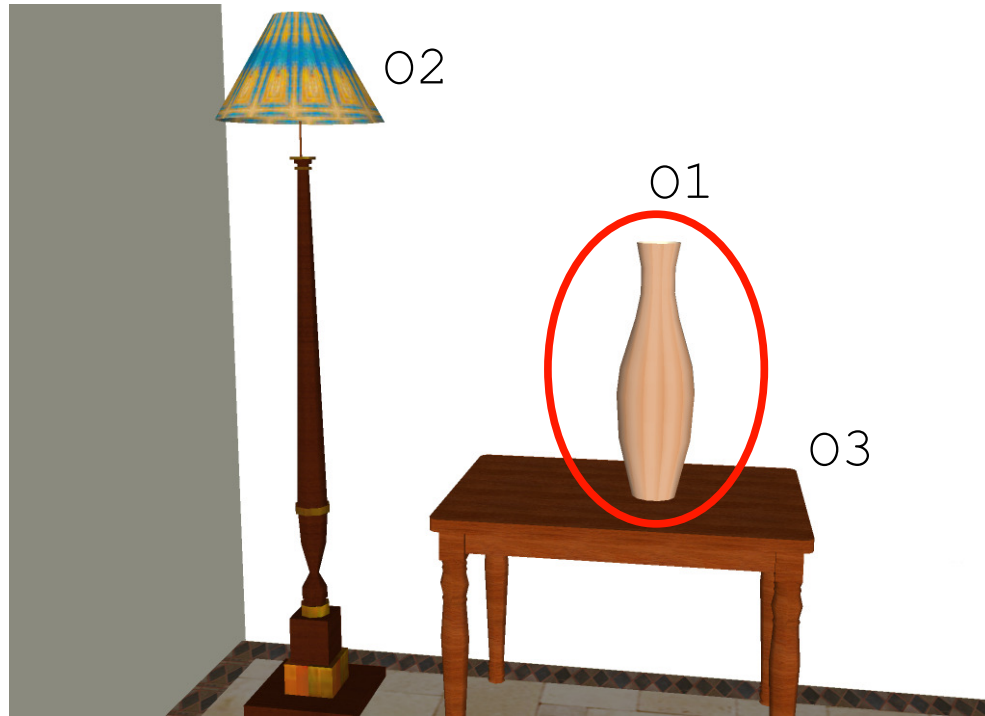
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1. Arbitrary descriptors

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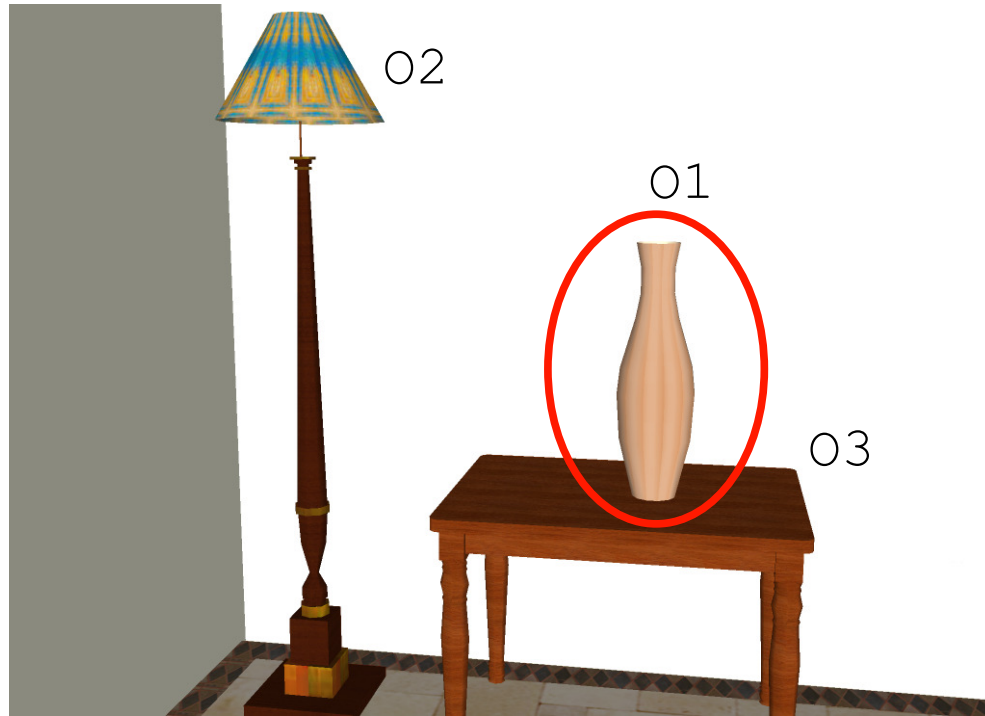
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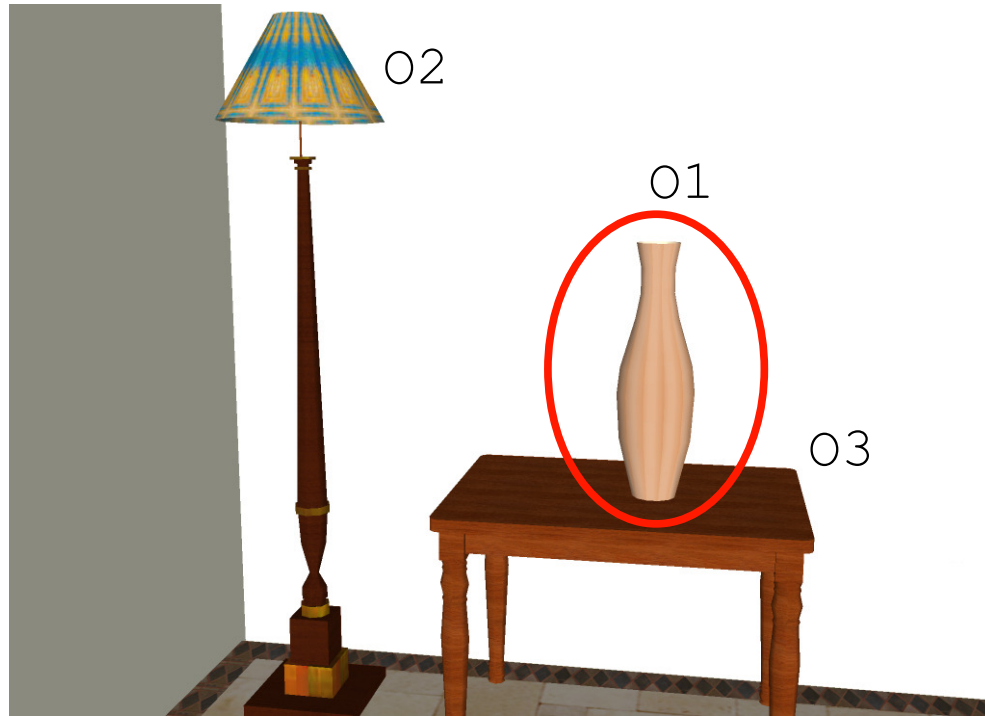


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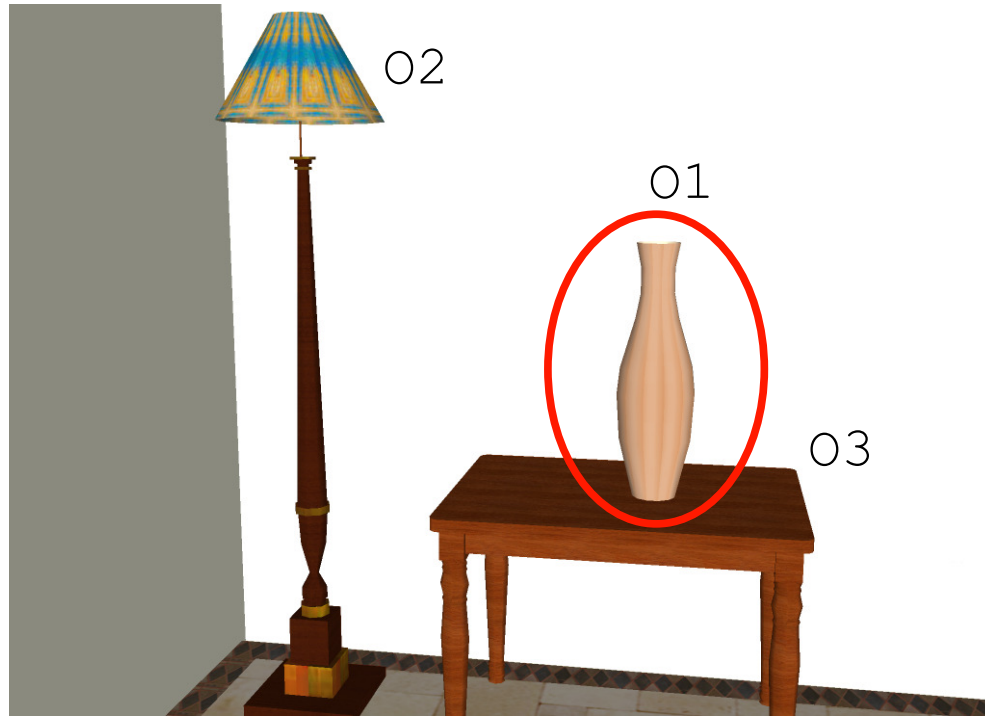
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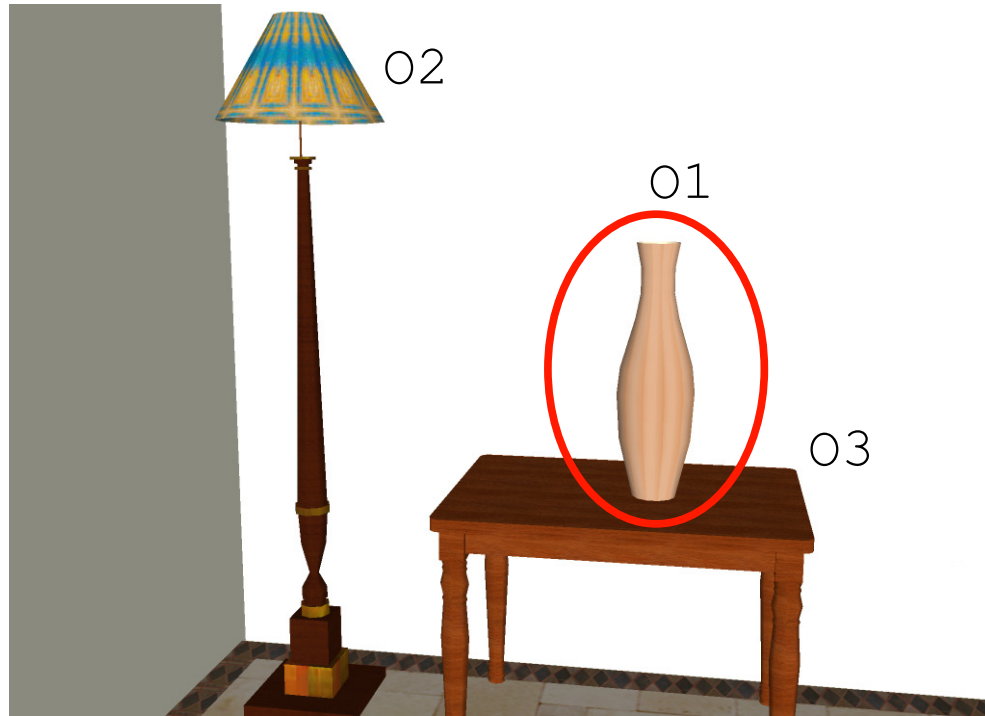
2. Complex utterances



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We will not be seeing 100% in this talk.

2. Complex utterances (coming up)

# Listener Extensions

- Training a Listener
- Generating Complex Utterances
- Modeling Listener Confusion

# Complex Utterances

Before: utterances were simple, such as:

*right of 02*

*on 03*

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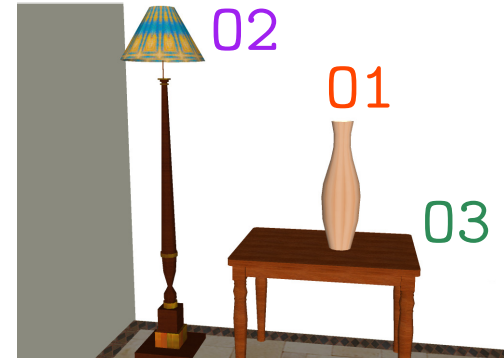
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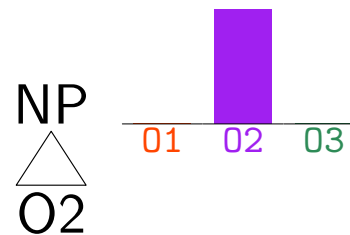
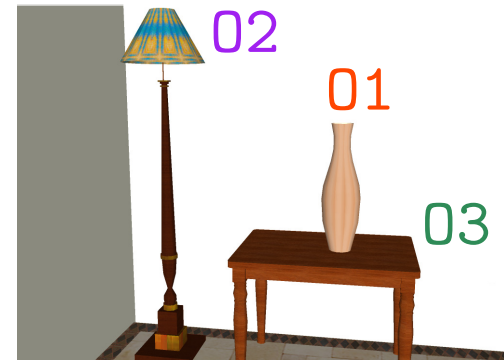
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Computing:  $p(g \mid \text{on something right of } 02)$



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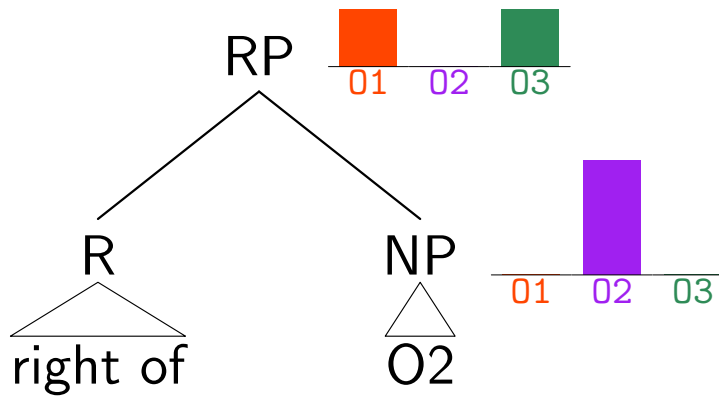
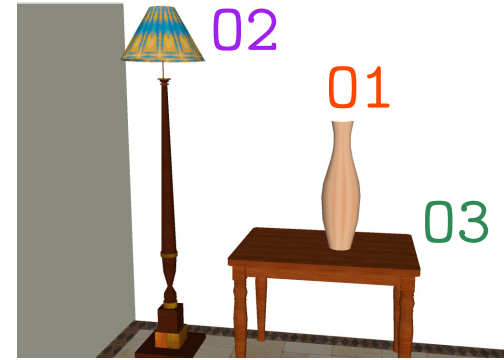
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If  $w$  is rooted at  $N$ ,  $p_L(g|w) = \mathbb{I}[w = g]$ .

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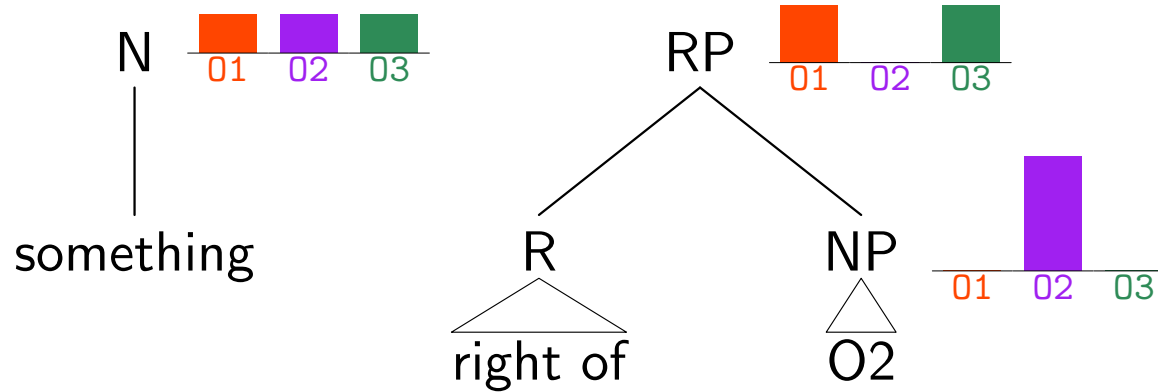
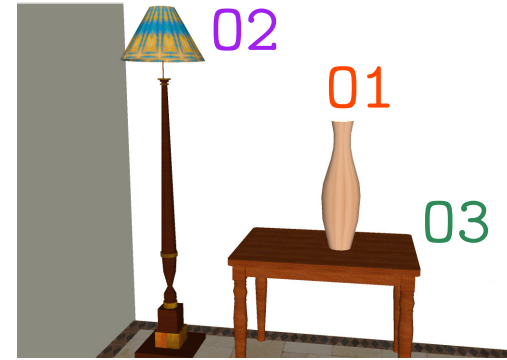
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If  $w$  is rooted at RP, recurse on NP subtree, use base listener.

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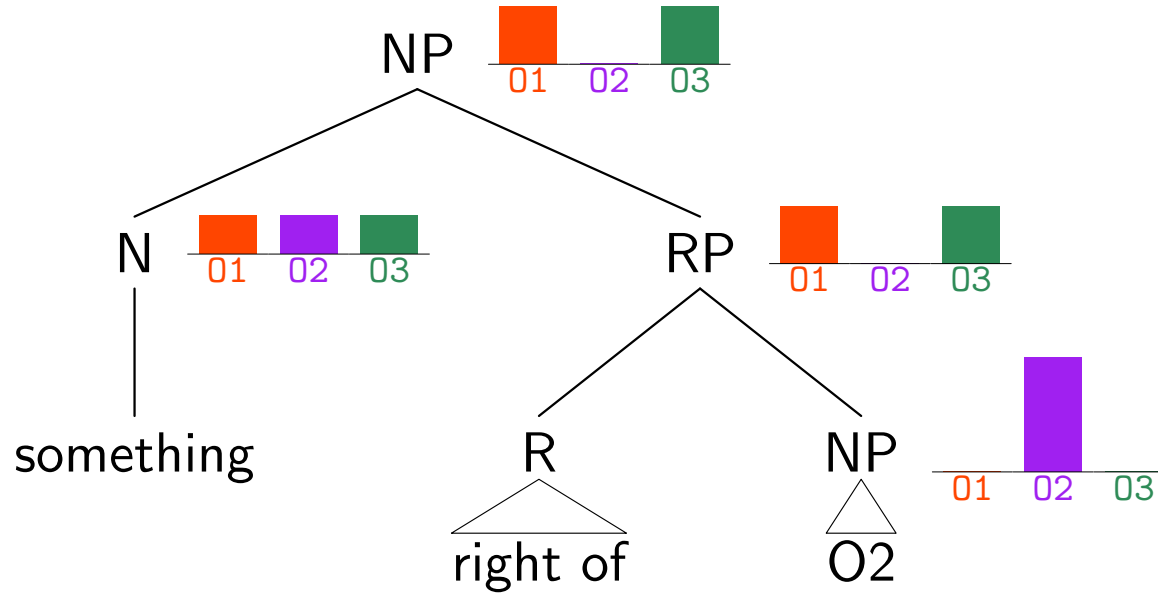
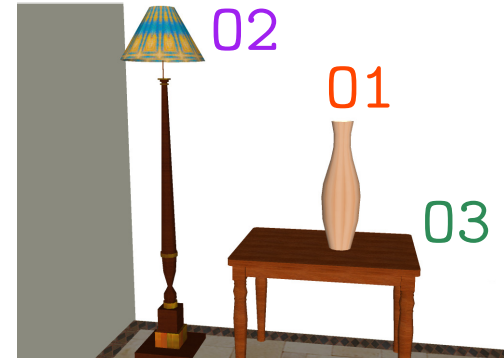
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If  $w$  is rooted at N and  $w = \text{something}$ ,  $p_L(g|w)$  is uniform.

# Example Interpretation

Computing:  $p(g \mid \text{on something right of } 02)$

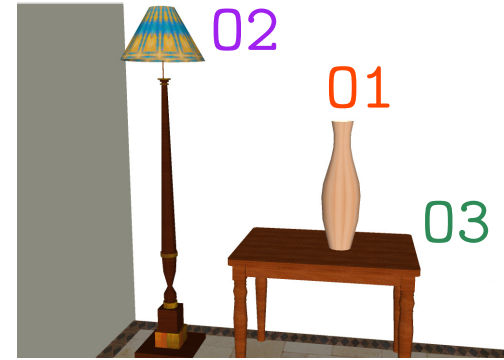
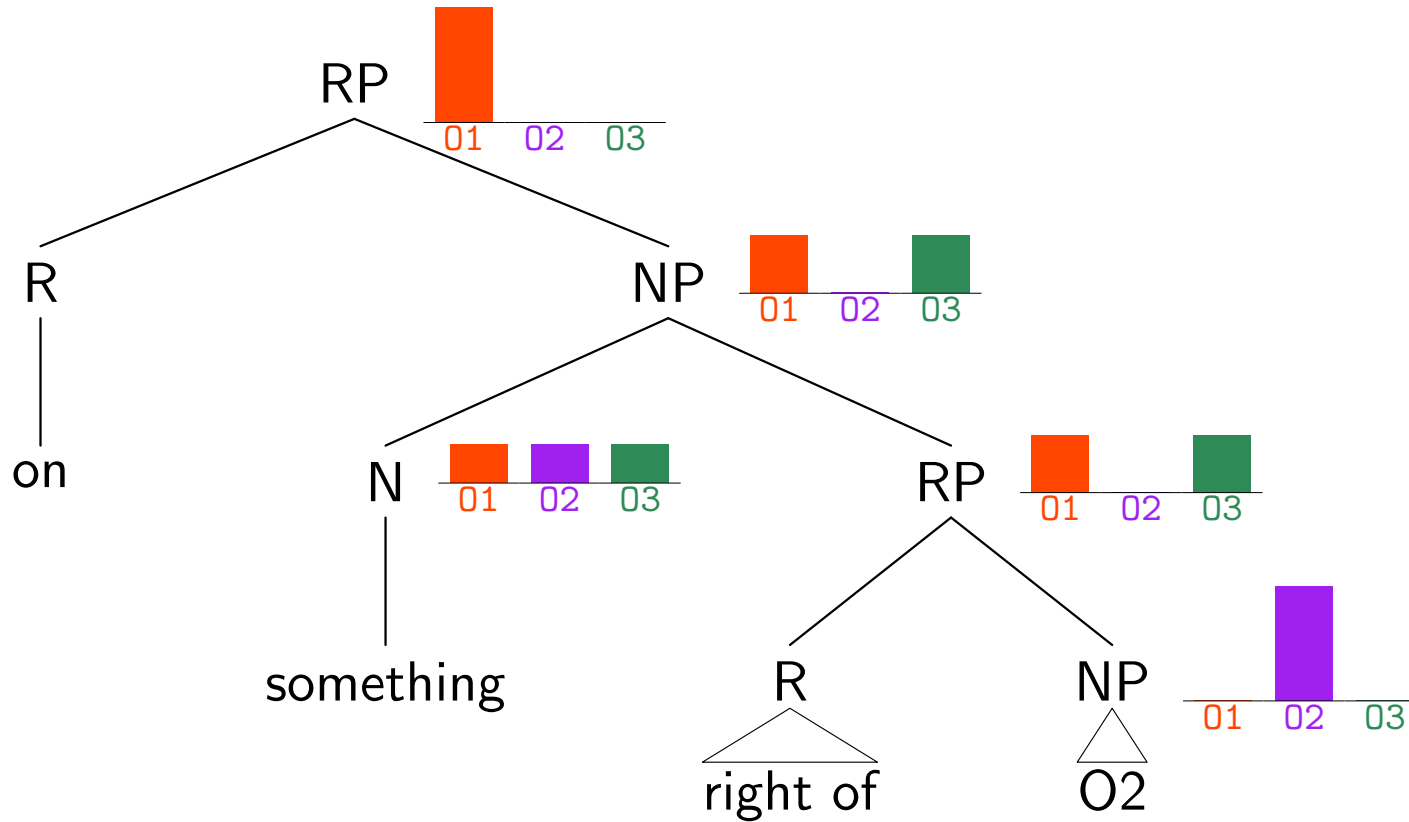


If  $w$  is rooted at NP, recurse on children, multiply and renormalize.



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# Results

	<u>Speaker</u>	<u>Success</u>
Reflex	<sup>S</sup> <input type="checkbox"/> LEARNED	38.4%
Rational	<sup>S(L)</sup> <input type="checkbox"/> LEARNED	52.6%

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Problem: introducing complex utterances hurts success

# Listener Confusion

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# Modeling Listener Confusion

**Problem:** our model does not match turkers

$|w|$

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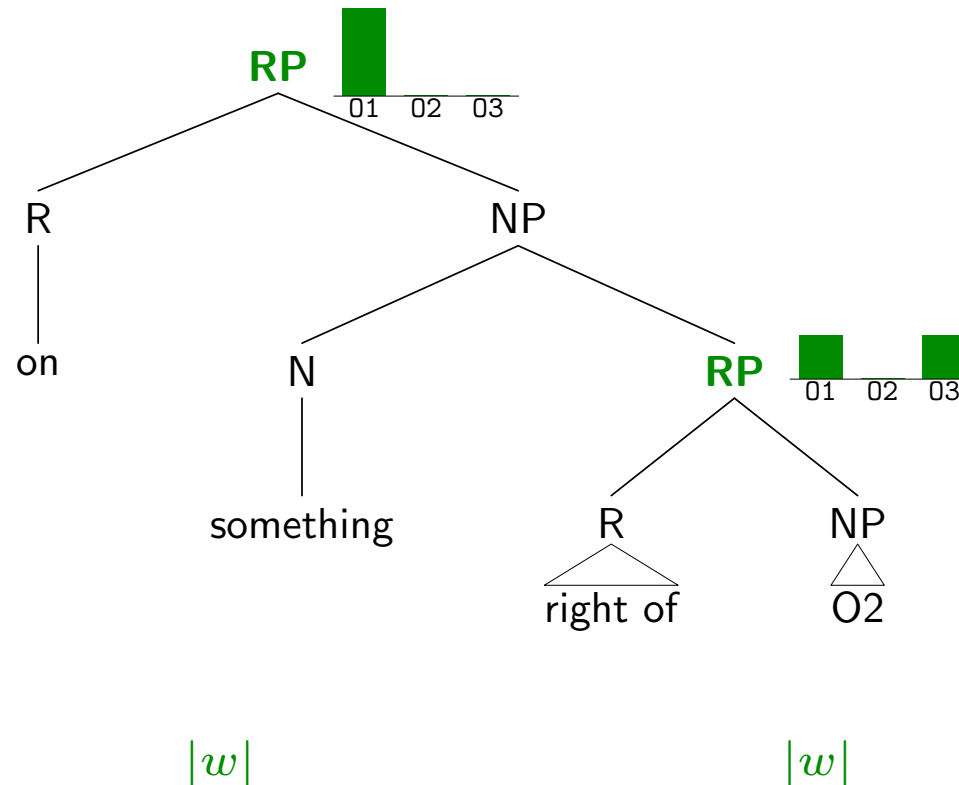
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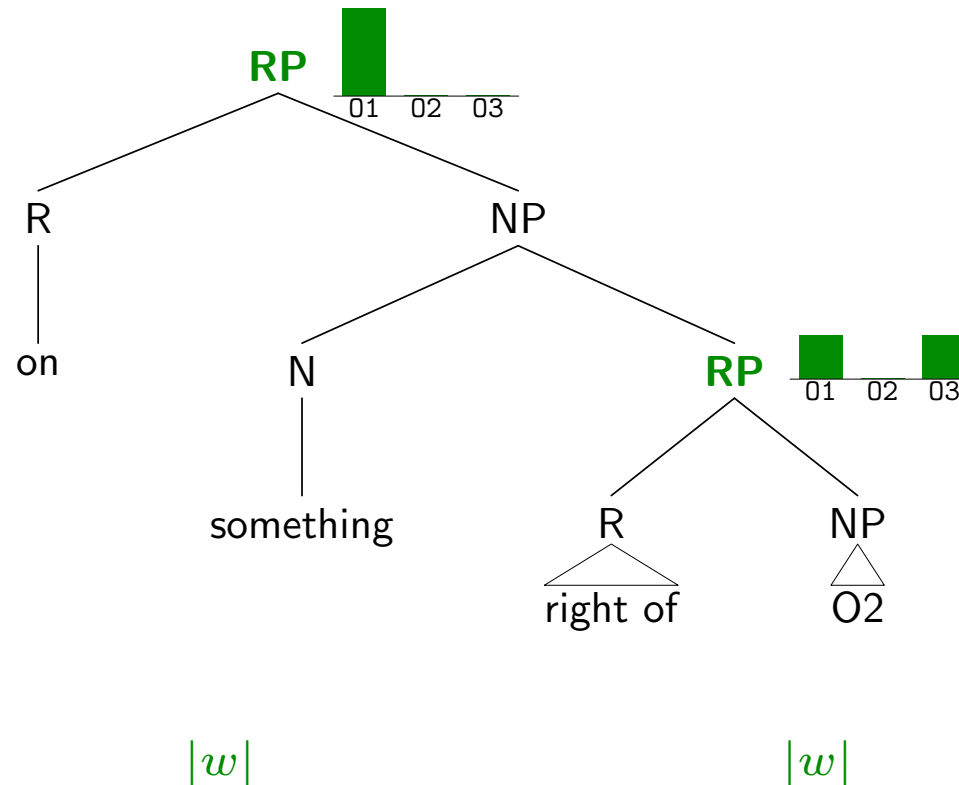
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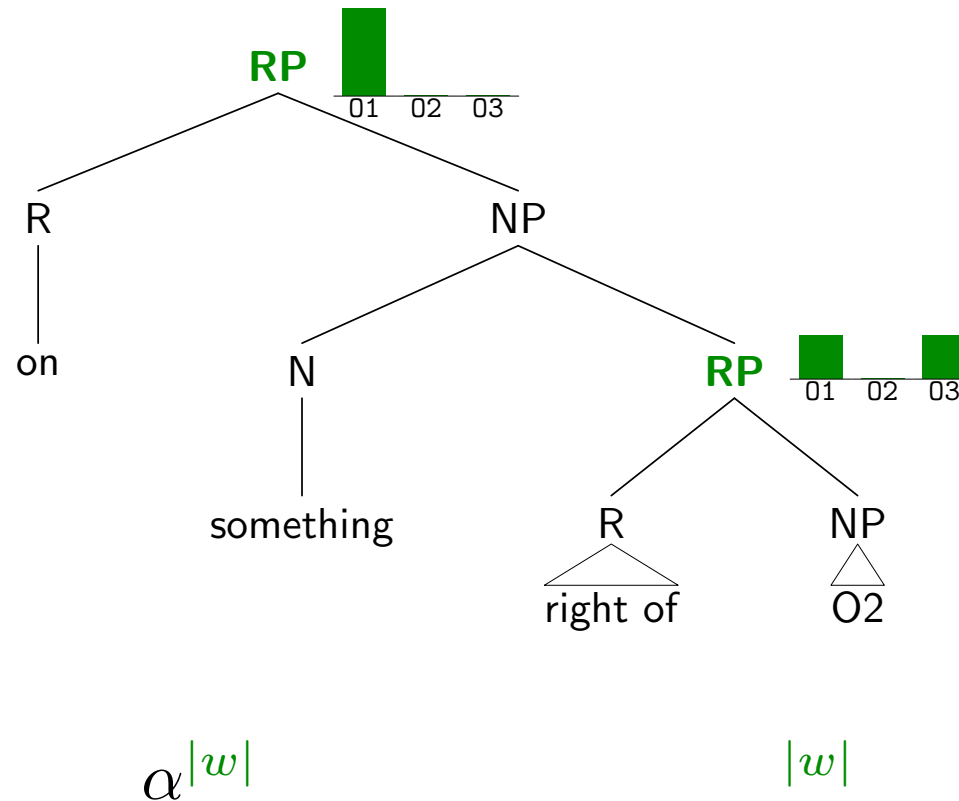
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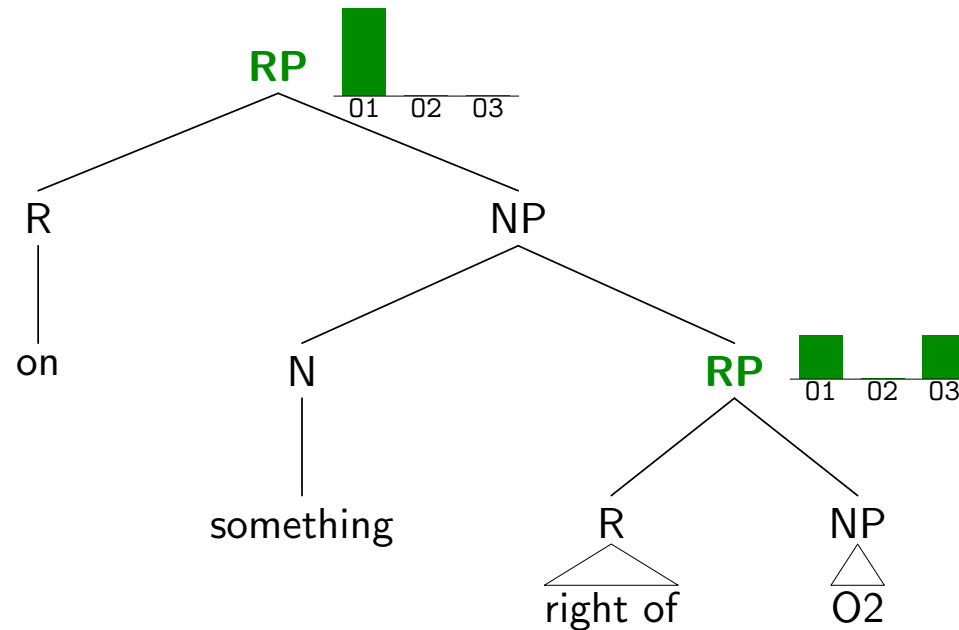
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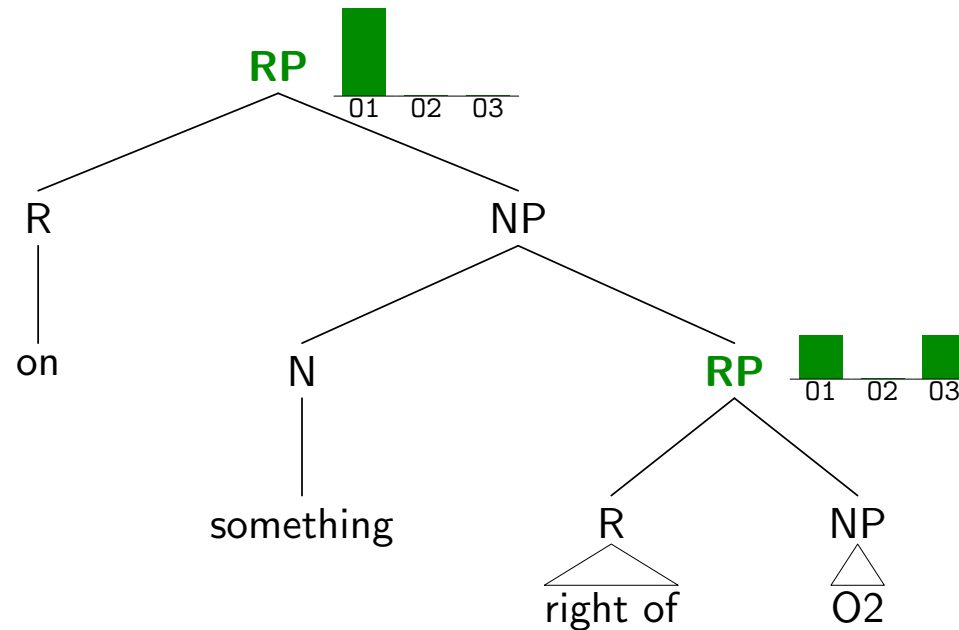


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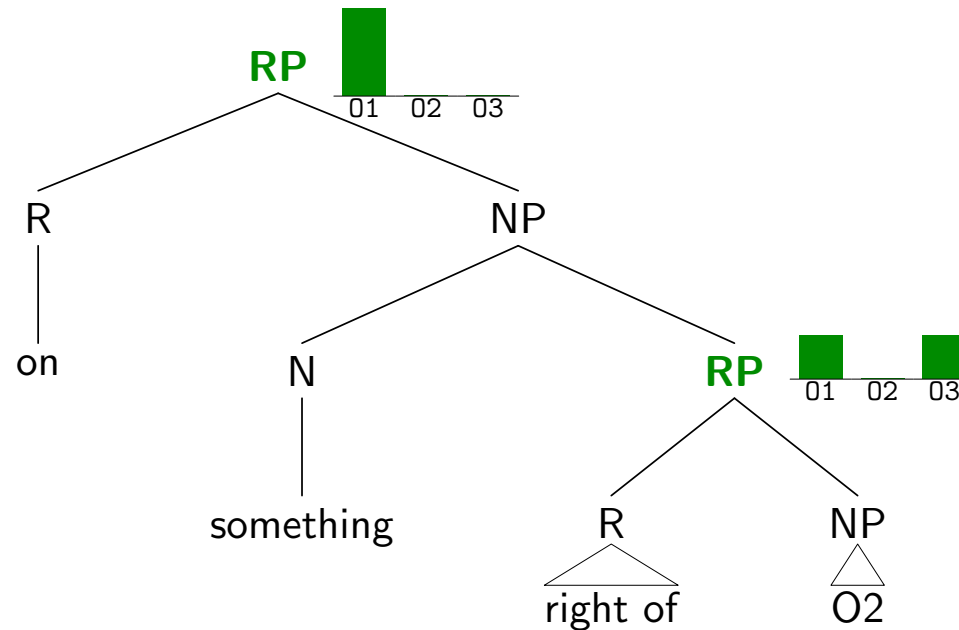
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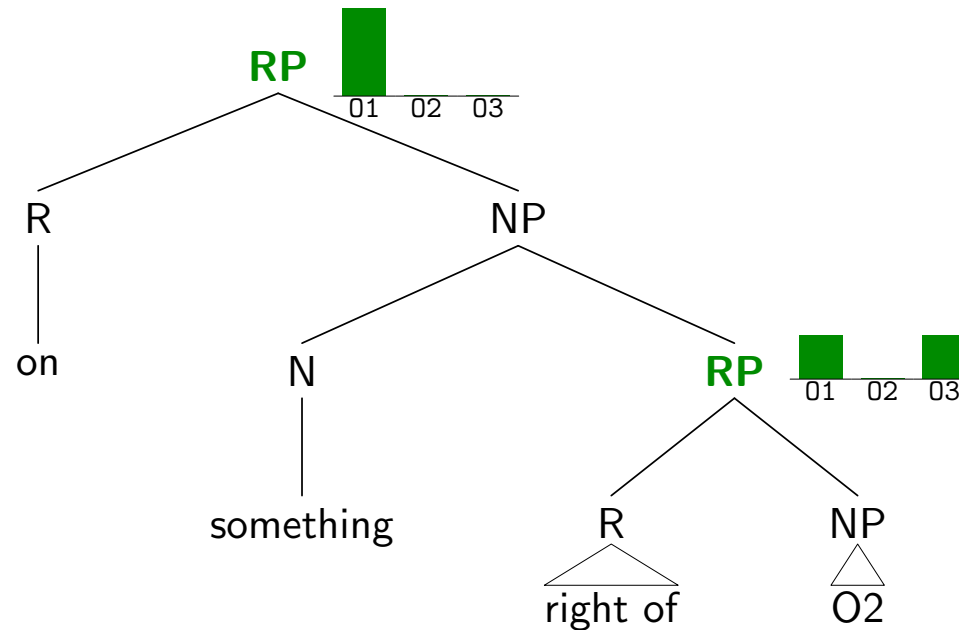


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


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**Our contribution:** we show how a game theoretic pragmatics model  
can be used to successfully generate spatial descriptions



Thank you!



slides compiled with rfig