

Incrementality in Compositional Distributional Semantics

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SemDial 2018
joint work with M. Purver, J. Hough, R. Kempson

SYCO2, Glasgow
December 2018

NLP in one slide

Formal
Grammar

structure preserving map



Semantic
Calculus

NLP in one slide

Formal
Grammar

structure preserving map



Models of
First Order
Logic

NLP in one slide

Formal
Grammar

structure preserving map



Distributions
of Linguistic
Data

Distributional Semantics

sugar, a sliced lemon, a tablespoonful of their enjoyment. Cautiously she sampled her first well suited to programming on the digital for the purpose of gathering data and

apricot
pineapple
computer.
information

preserve or jam, a pinch each of, and another fruit whose taste she likened In finding the optimal R-stage policy from necessary for the study authorized in the

	computer	data	pinch	result	sugar
apricot	0	0	2.25	0	2.25
pineapple	0	0	2.25	0	2.25
digital	1.66	0	0	0	0
information	0	0.57	0	0.47	0

$$\text{PPMI}(w, c) = \max\left(\log_2 \frac{P(w, c)}{P(w)P(c)}, 0\right)$$

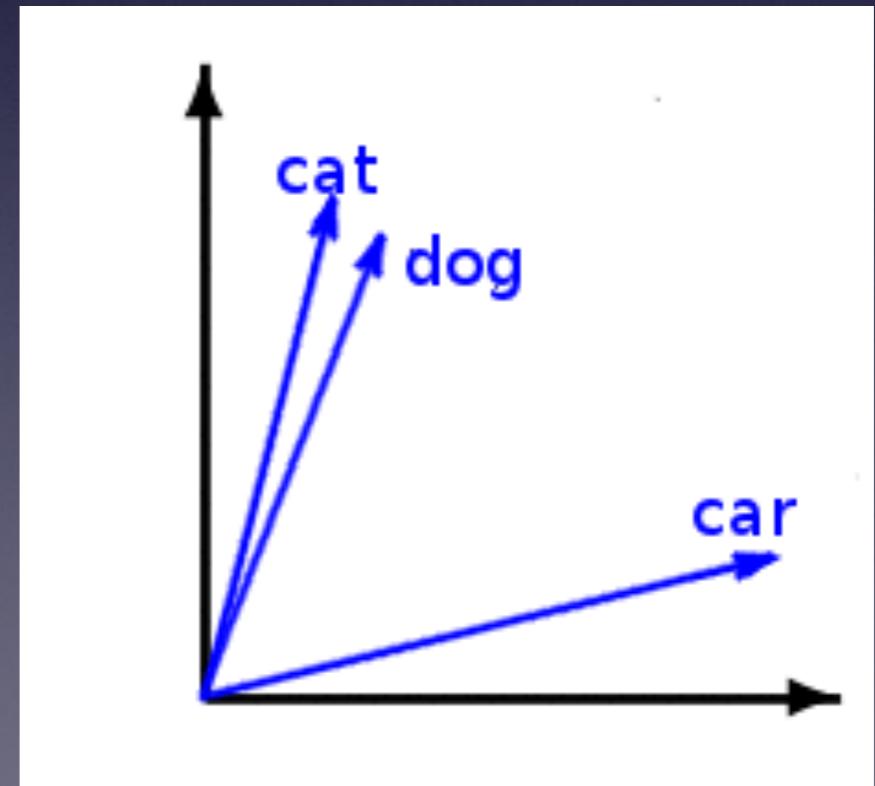
Speech and Language Processing,
Jurafsky and Martin

State of the art NLP packages

```
import spacy  
nlp = spacy.load('en_core_web_md')  
tokens = nlp(u'dog cat car')  
for token1 in tokens:  
    for token2 in tokens:  
        print(token1.text, token2.text, token1.similarity(token2))
```

```
dog  dog  1.0  
dog  cat  0.80168545  
dog  car  0.35629162  
cat  dog  0.80168545  
cat  cat  1.0  
cat  car  0.31907532  
car  dog  0.35629162  
car  cat  0.31907532  
car  car  1.0
```

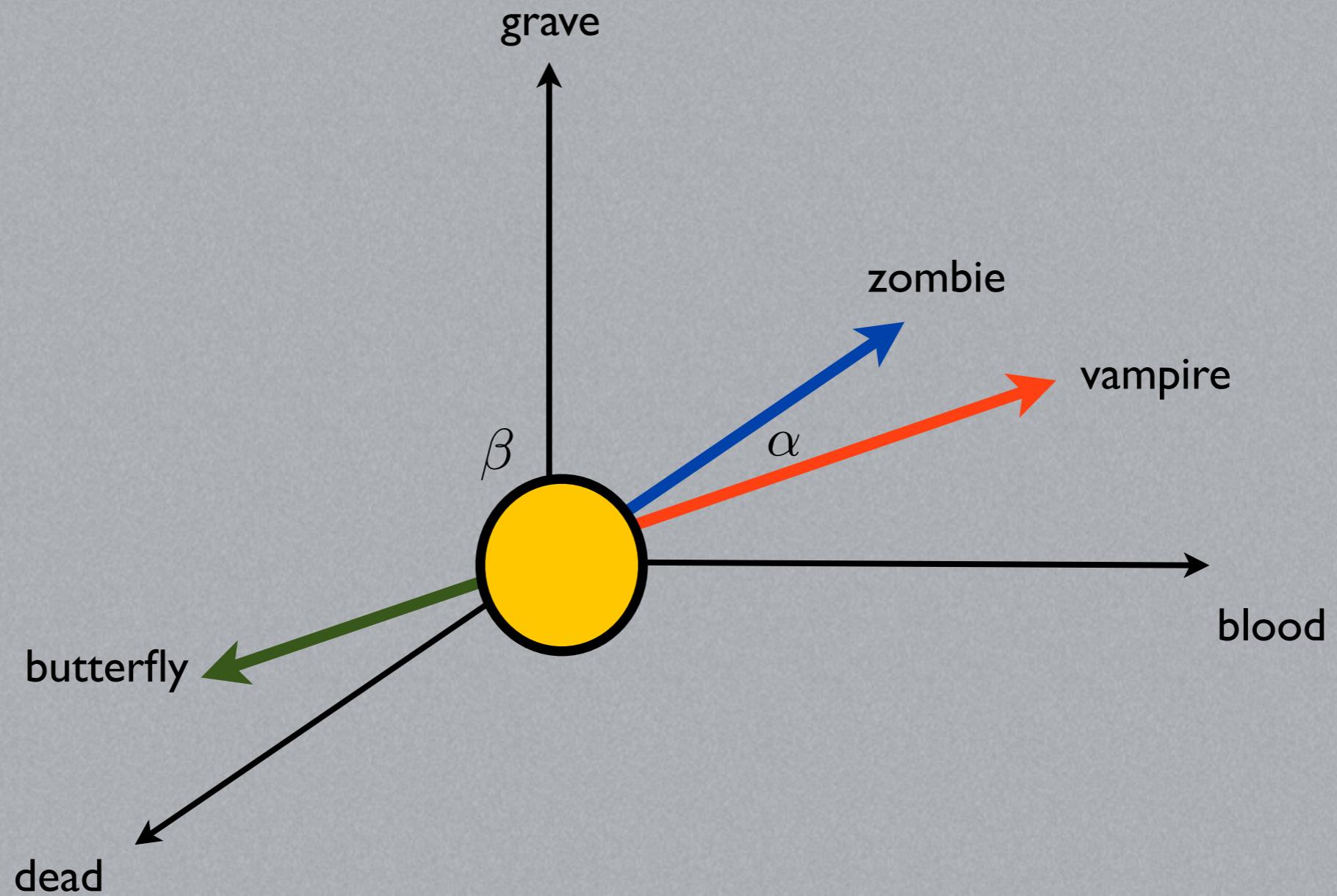
spaCy



Distributional Semantics

	dog	cat	car
dog	1	0.80	0.35
cat		1	0.31
car			1

Distributional Semantics



NLP in one slide

Formal
Grammar

structure preserving map



Distributions
of Linguistic
Data

???

structure preserving map



???

NLP in one slide

Formal
Grammar

structure preserving map



Distributions
of Linguistic
Data

Type
Grammars

structure preserving map



Multilinear
Algebra

CCG

Types

A
 X/Y
 $X \setminus Y$

NP, S
NP/NP, S\NP
(S\NP)/NP

noun phrase
adj, iTv
Tv

Rules

$X/Y \quad Y \Rightarrow X$
 $Y \quad X \setminus Y \Rightarrow X$

NP/NP NP => NP
NP S\NP => S

Multilinear Algebraic Semantics

Vectors

$A \mapsto \mathcal{A}$

$$\mathcal{A} = \{e_i\}_i \quad \ni T_i = \sum_i C_i e_i$$

Multilinear Algebraic Semantics

Matrices

$$A/B \mapsto \mathcal{A} \otimes \mathcal{B}$$

$$\mathcal{A} = \{e_i\}_i \quad \mathcal{B} = \{e_j\}_j$$

$$\mathcal{A} \otimes \mathcal{B} \ni T_{ij} = \sum_{ij} C_{ij} e_i \otimes e_j$$

Multilinear Algebraic Semantics

Cubes

$$A/(B/C) \mapsto \mathcal{A} \otimes (\mathcal{B} \otimes \mathcal{C}) \quad \mathcal{A} = \{e_i\}_i \quad \mathcal{B} = \{e_j\}_j \quad \mathcal{C} = \{e_k\}_k$$

$$\mathcal{A} \otimes \mathcal{B} \otimes \mathcal{C} \ni T_{ijk} = \sum_{ijk} C_{ijk} \ e_i \otimes e_j \otimes e_k$$

Multilinear Algebraic Semantics

Higher order tensors

$$\mathcal{A} \otimes \mathcal{B} \otimes \cdots \otimes \mathcal{Z} \ni T_{ij\cdots w} = \sum_{ij\cdots w} C_{ij\cdots w} e_i \otimes e_j \otimes \cdots \otimes e_w$$

Multilinear Algebraic Semantics

Matrix Multiplication

$$A/B \quad B \implies A \quad \mapsto \quad (\mathcal{A} \otimes \mathcal{B}) \quad \mathcal{B} \implies \mathcal{A}$$

$$T_{ij} \quad T_j \xrightarrow{\text{tensor contract}} T_i$$

$$\left(\sum_{ij} C_{ij} e_i \otimes e_j \right) \left(\sum_i C_j e_j \right) = \sum_i C_{ij} C_j e_i \langle e_j \mid e_j \rangle$$

Multilinear Algebraic Semantics

Higher order tensor contraction

$$\cdots \mapsto \mathcal{A} \otimes \mathcal{B} \otimes \cdots \otimes \mathcal{M} \quad \mathcal{M} \otimes \mathcal{N} \otimes \mathcal{P} \otimes \cdots \otimes \mathcal{W}$$

$$T_{ij\cdots m} \quad T_{mnp\cdots w} \quad \xrightarrow{\text{tensor contract}} \quad T_{ij\cdots np\cdots w}$$

$$\begin{aligned} & \left(\sum_{ij\cdots m} C_{ij\cdots m} e_i \otimes e_j \otimes \cdots \otimes e_m \right) \left(\sum_{mn\cdots w} C_{mn\cdots w} e_m \otimes e_n \otimes \cdots \otimes e_w \right) \\ &= \sum_{ij\cdots n\cdots w} C_{ij\cdots m} C_{mn\cdots w} e_i \otimes e_j \otimes \cdots \otimes e_n \otimes \cdots \otimes e_w \langle e_m \mid e_m \rangle \end{aligned}$$

Dogs Chase White Cats

NP $(S \setminus NP)/NP$ NP/NP NP

NP

$S \setminus NP$

S

Dogs Chase White Cats

\mathcal{N} $(\mathcal{S} \otimes \mathcal{N}) \otimes \mathcal{N}$ $\mathcal{N} \otimes \mathcal{N}$ \mathcal{N}

\mathcal{N}

$\mathcal{S} \otimes \mathcal{N}$

\mathcal{S}

Dogs Chase white Cats

T_i T_{ijk} T_{kl} T_l

T_k

T_{ij}

T_j

Pregroup Grammars

Types ...

$$XY^l$$

$$NPNP^l \quad NP^r S$$

$$Y^r X$$

$$NP^r S NP^l$$

Rules

$$XY^l Y \leq X$$

$$NPNP^l NP \leq NP$$

$$YY^r X \leq X$$

$$NPNP^r S \leq S$$

Catgorical Semantics

Formal
Grammar

structure preserving map



Distributions
of Linguistic
Data

Pregroup
Grammars

monoidal functor



FVect

Categorial Grammars + Distributional Semantics

Coecke, Sadrzadeh, Clark, 2010
Grefenstette and Sadrzadeh 2011, 2015
Maillard, Clark, Grefenstette, 2014
Krishnamurti and Mitchell, 2014
Baroni and Zamparelli 2010
Wijnholds (and Moortgat) 2015-16

Language Processing

Complete Sentences

Naturally Occurring Dialogue



- “*A: mary likes ...*” “*B: john?*”
- “*A: mary likes ...*” “*B: who?*”
- “*A: mary likes ...*” “*B: nobody really*”

Naturally Occurring Dialogue



A: Ray destroyed . . .

B: . . . the fuchsia. He never liked it. The roses he spared . . .

A: . . . this time.

Naturally Occurring Dialogue



A: *You are going to write the letter?*
B: *Only if you post it!*

Howes et al, 2011, Poesio and Reiser 2010

Computational Dialogue Systems

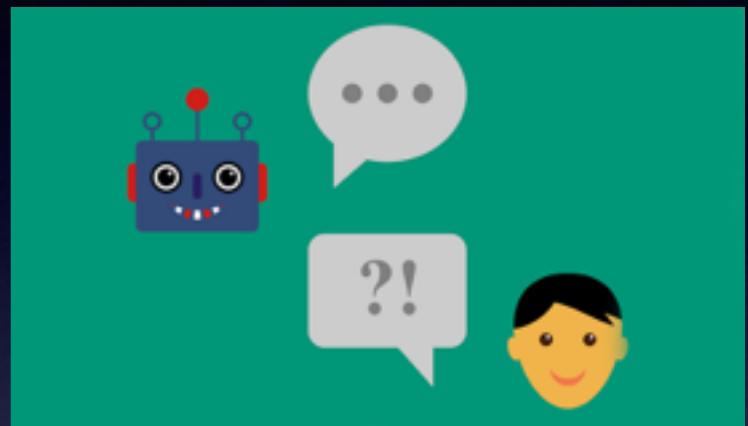
A: *I want to book a ticket ...*

B: *... from where?*

A: *London*

B: *... to where?*

A: *to Paris.*



Purver and Kempson 2011
Purver, Eshghi, Hough 2017

Psycholinguistic Analysis

A: *The footballer dribbled ...*

B (*thinking*) *it means controlling the ball*

A: *... the ball across the pitch*

A: *The baby dribbled ... the milk all over the floor.*



Pickering and Frisson 2001

Cognitive Neuroscience



Predictive Processing: agents incrementally generate expectations and judge the degree to which they are met.

Frisson and Frith 2001
Clarke 2015

- Incremental Language Processing

Dynamic Syntax

+

Type Theoretic Semantics

Ruth Kempson, Wilfried Meyer-Viol, and Dov Gabbay. 2001.

Hough 2015, Purver et al 2014.

Recent Contribution

Dynamic Syntax
+
Distributional Semantics

Sadrzadeh, Purver, Hough, Kempson

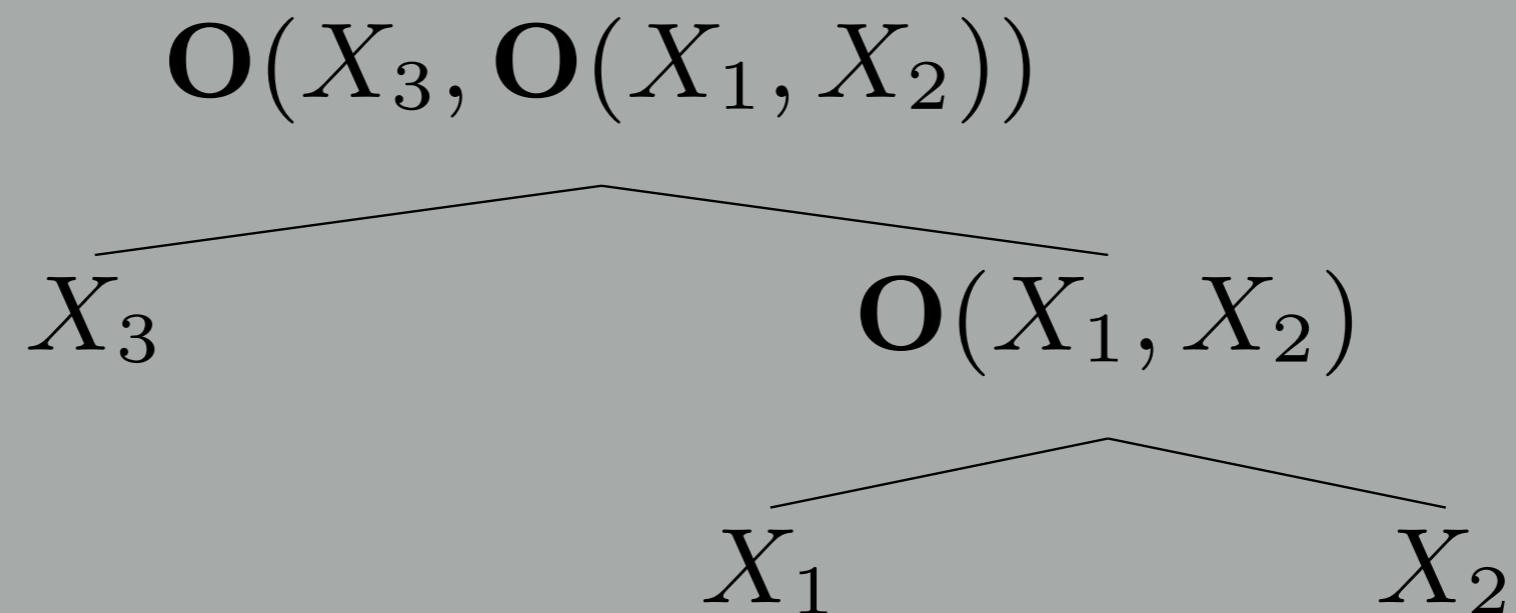
SemDial 2018

Outline

- Dynamic Syntax: DS
- CDS for DS
- Some Examples
- Some Experimental Results

Dynamic Syntax

Trees decorated with semantic formulae and applications



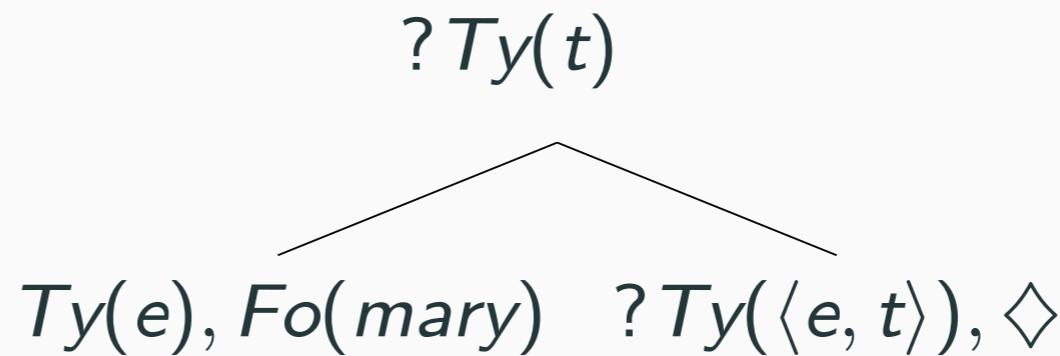
Dynamic Syntax

and with ...

- Ty: types of formulae
- ?: requirements for further development
- <>: node currently under development
- links: connect trees of arguments of conjunctives etc

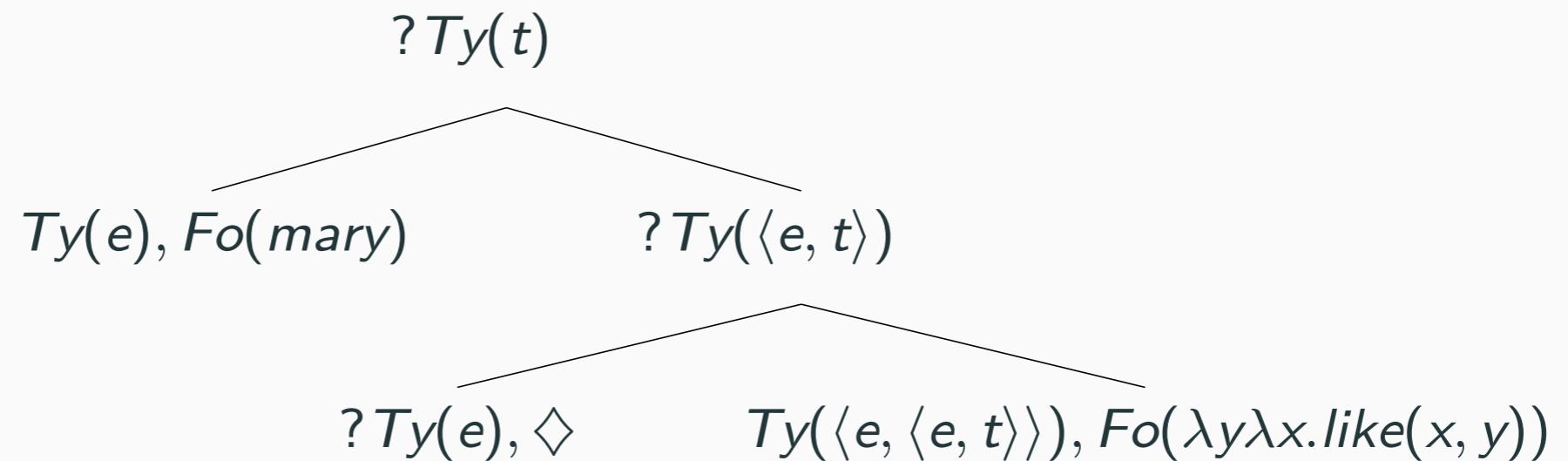
Dynamic Syntax

“*mary* ...”



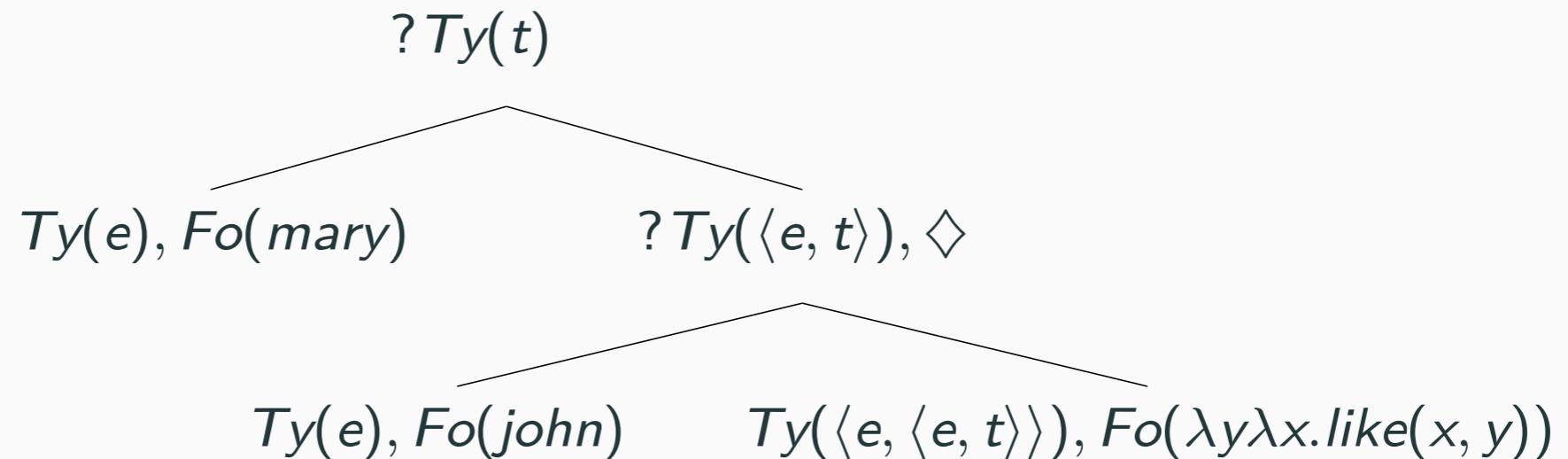
Dynamic Syntax

“mary likes ...”



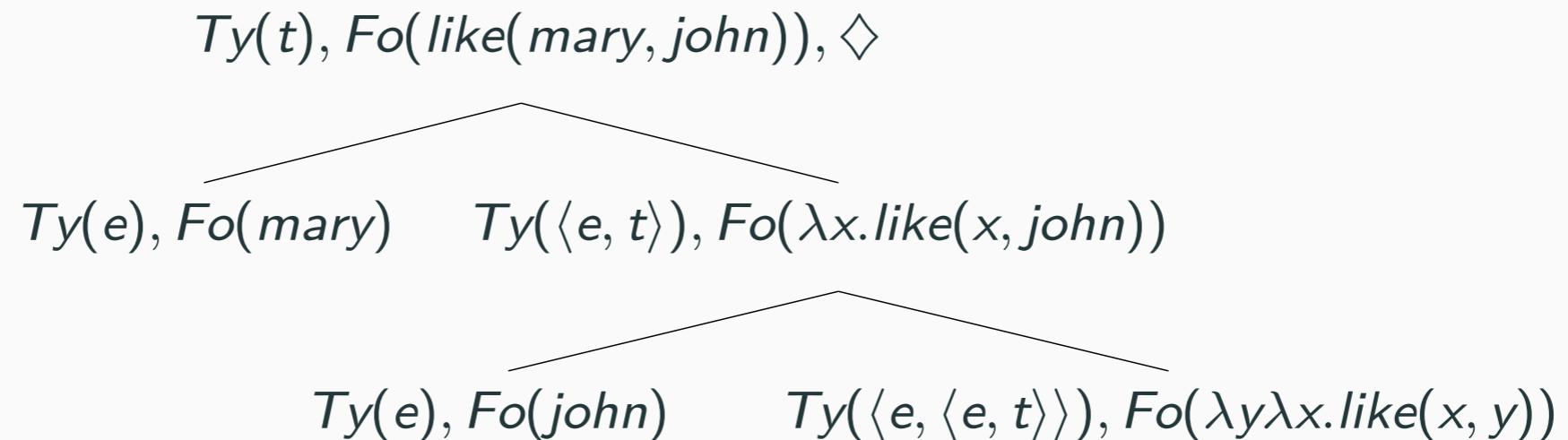
Dynamic Syntax

“mary likes john”

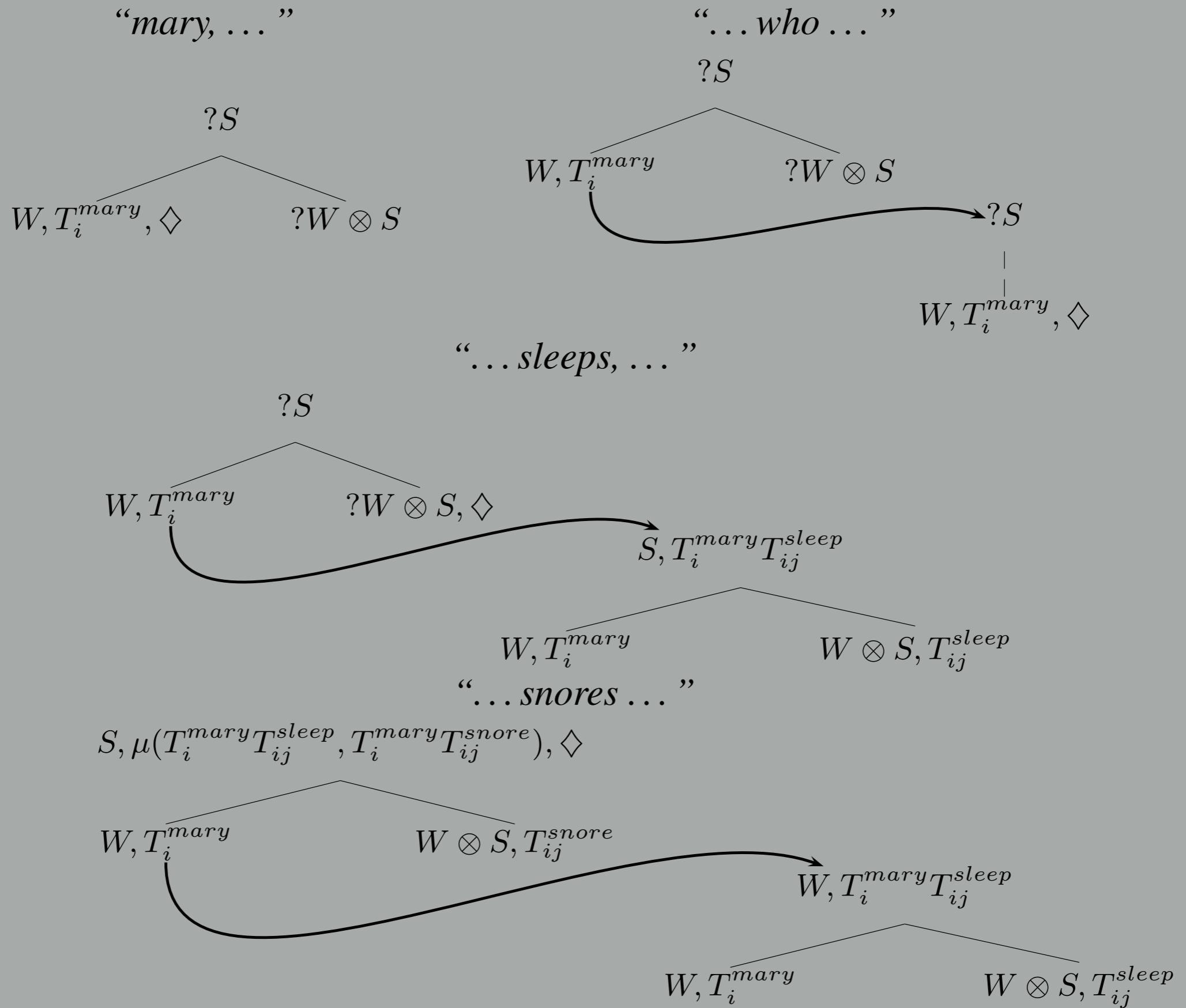


Dynamic Syntax

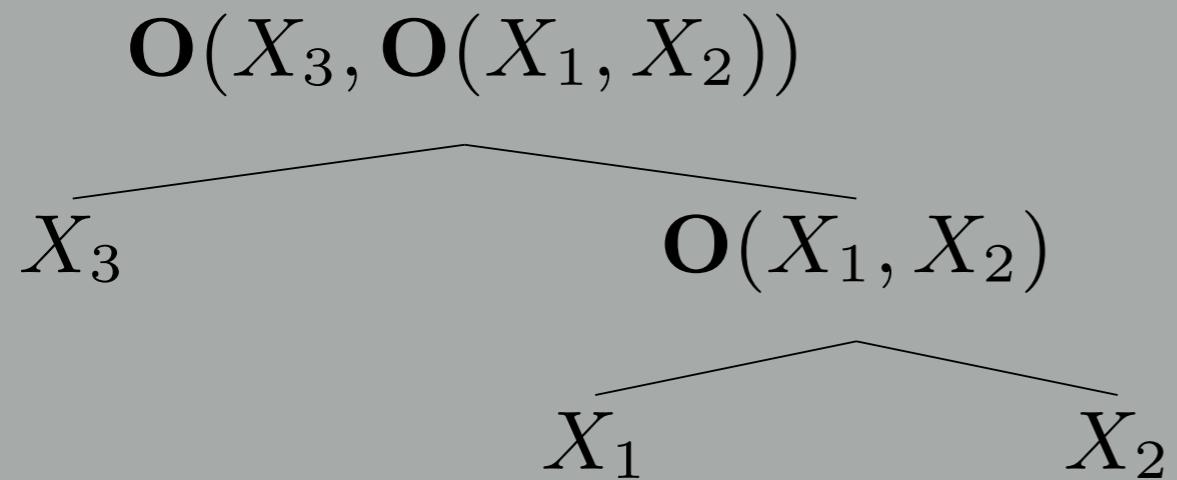
“mary likes john”



Mary who sleeps snores.



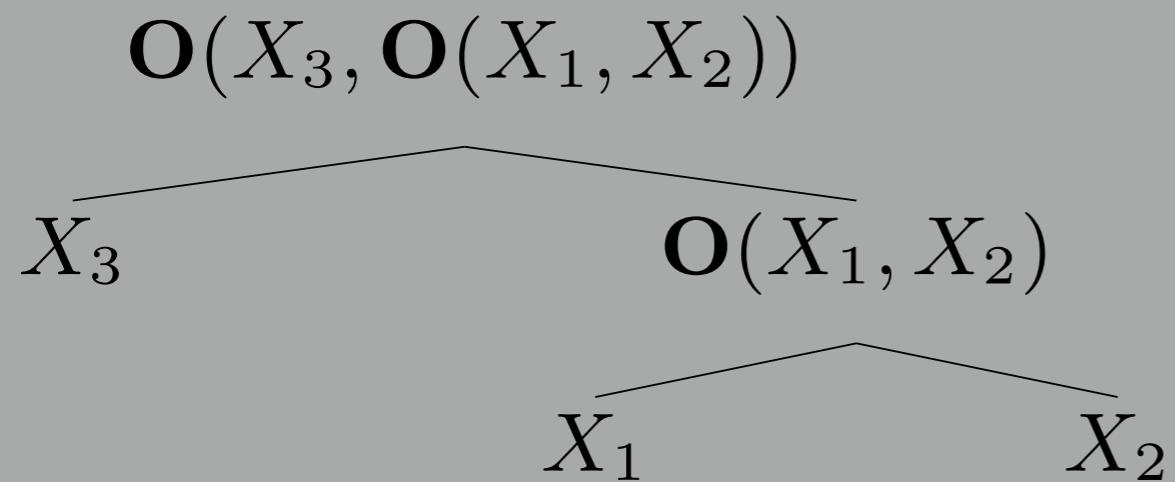
Multilinear Algebraic Semantics for DS



Simple Nodes

$$\begin{array}{llll} X_1 & \mapsto & T_{i_1 i_2 \dots i_n} & \in V_1 \otimes V_2 \otimes \cdots \otimes V_n \\ X_2 & \mapsto & T_{i_n i_{n+1} \dots i_{n+k}} & \in V_n \otimes V_{n+1} \otimes \cdots \otimes V_{n+k} \\ X_3 & \mapsto & T_{i_{n+k} i_{n+k+1} \dots i_{n+k+m}} & \in V_{n+k} \otimes V_{n+k+1} \otimes \cdots \otimes V_{n+k+m} \end{array}$$

Multilinear Algebraic Semantics for DS



Operations Nodes

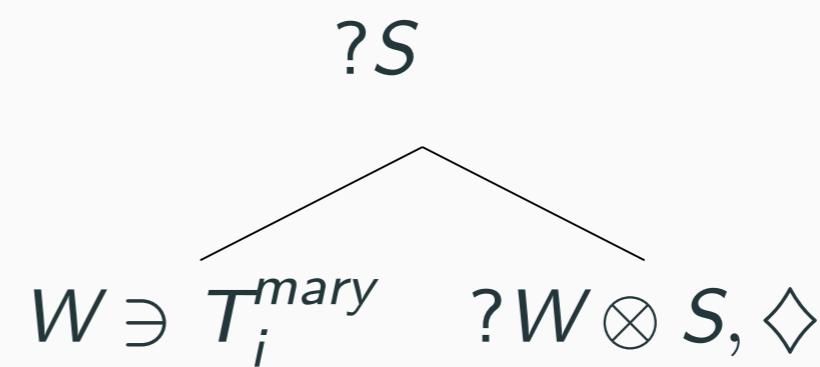
$$\begin{aligned} \mathbf{O}(X_1, X_2) &\mapsto T_{i_1 i_2 \dots i_n} T_{i_n i_{n+1} \dots i_{n+k}} \\ &\in V_1 \otimes V_2 \otimes \dots \otimes V_{n-1} \otimes V_{n+1} \otimes \dots \otimes V_{n+k} \end{aligned}$$

Multilinear Algebraic Semantics for DS

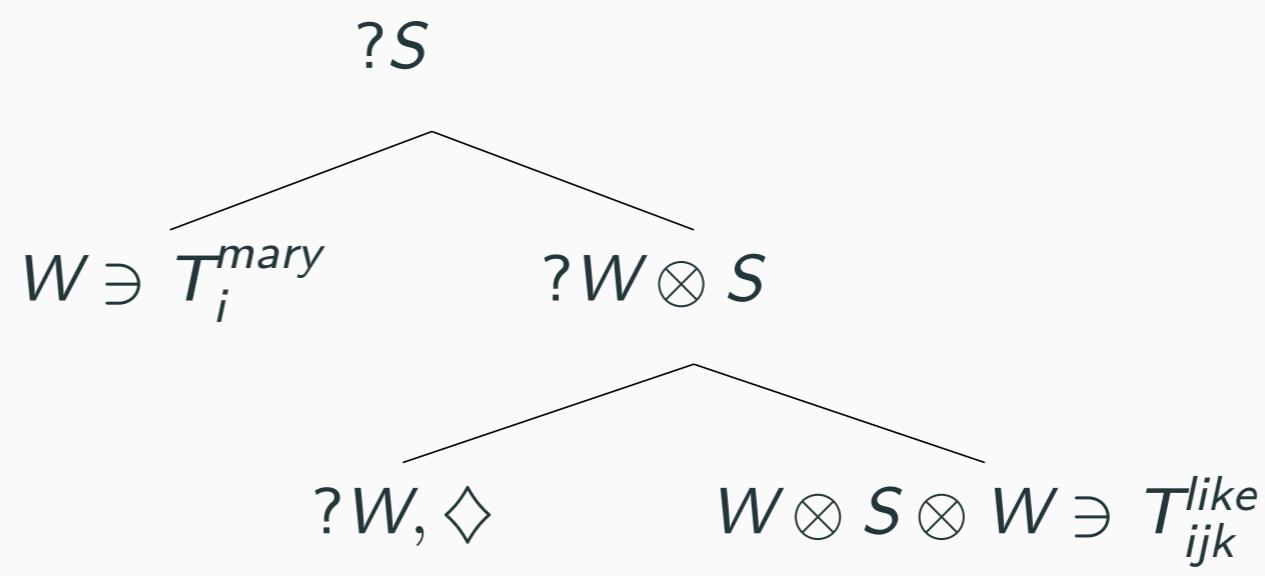
Extras

- $Ty(t) \longrightarrow S$
- $Ty(e) \longrightarrow W$
- $?X \longrightarrow$ sum or direct sum of the words and phrase with semantics in X and their probabilities
- $_ \longrightarrow$ a neutral element such as the identity in X
- $_ \longrightarrow$ a tensor full of 1's in X

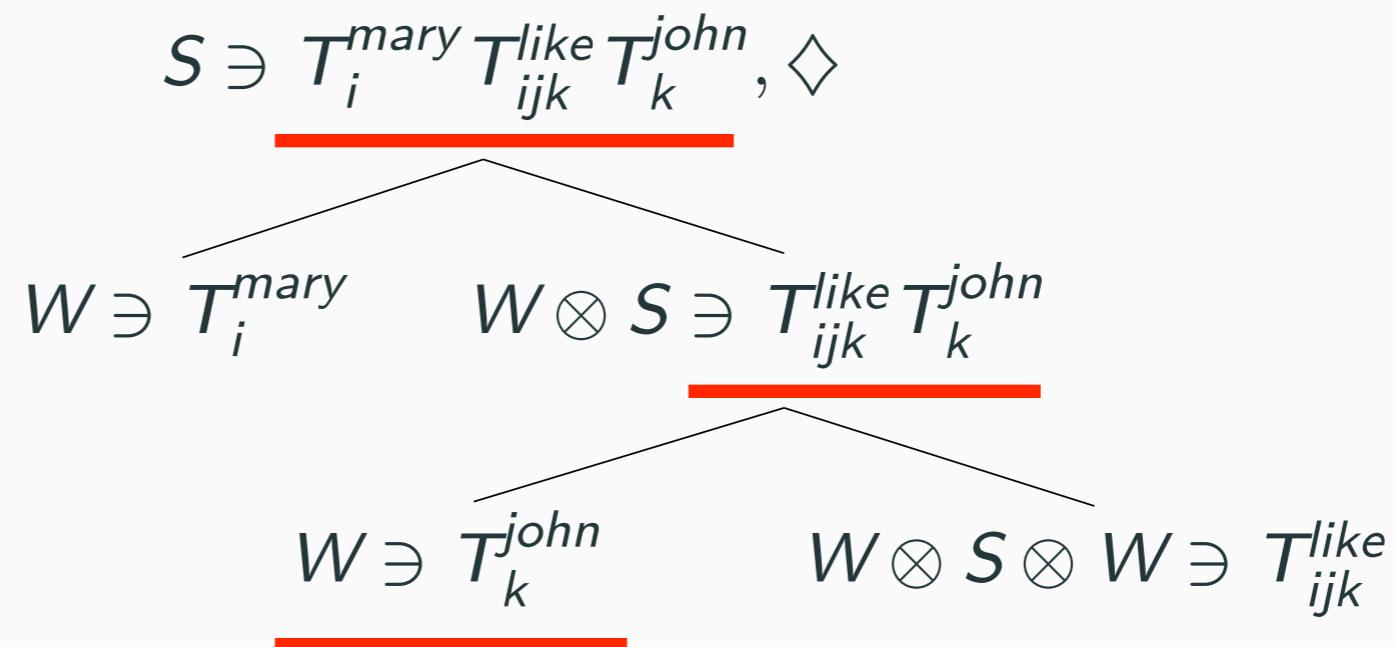
“*mary* ...”



“*mary likes ...*”

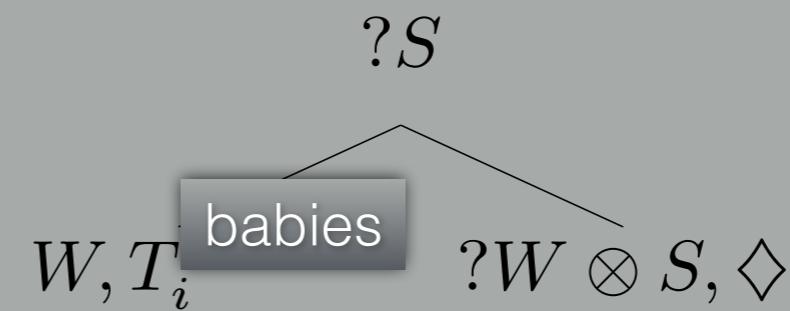


“*mary likes john*”



Incremental Utterances

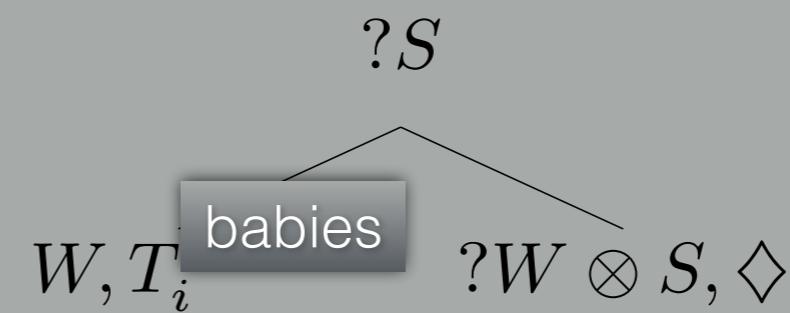
“Babies ...”



$T_i^{babies} T_{ij}^+$

Incremental Utterances

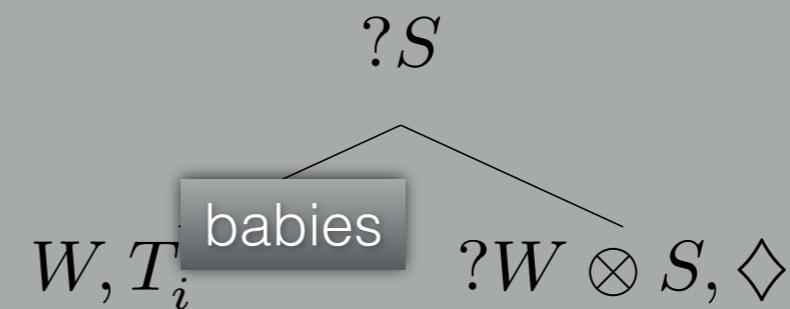
“Babies ...”



$$T_{ij}^+ =$$

Incremental Utterances

“Babies ...”



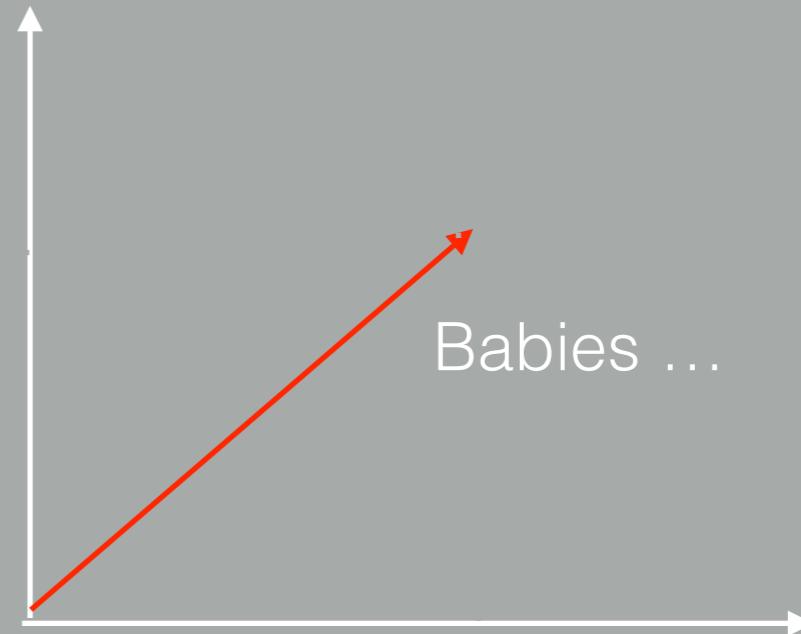
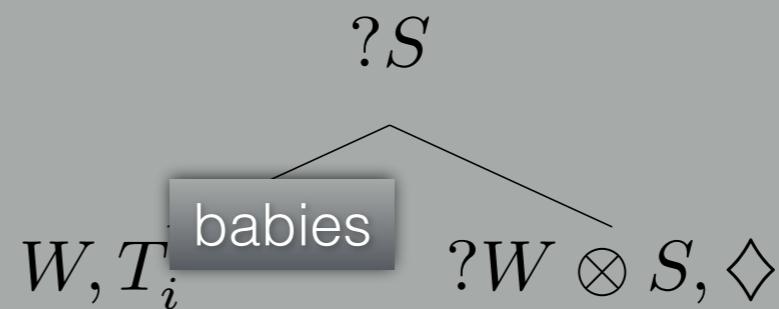
$$T_{ij}^+ =$$

$$T^{vomit} + T^{score} + T^{dribble} + T^{control\ baby} + T^{control\ milk} + T^{control\ footballer} + T^{control\ ball}$$

Incremental Utterances

“Babies ...”

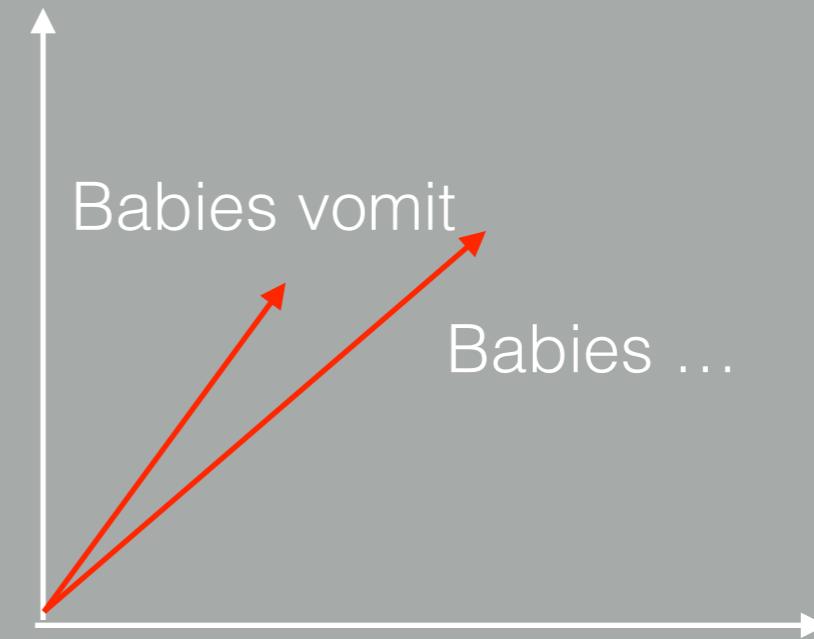
$T_i^{babies} T_{ij}^+$



Incremental Utterances

“Babies vomit”

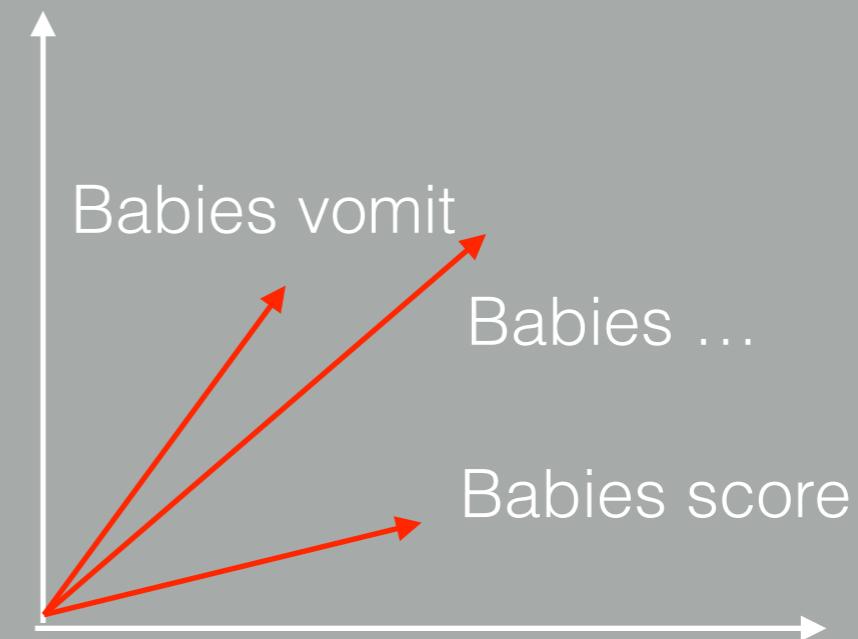
$$T_i^{babies} T_{ij}^{vomit}$$



Incremental Utterances

“Babies score”

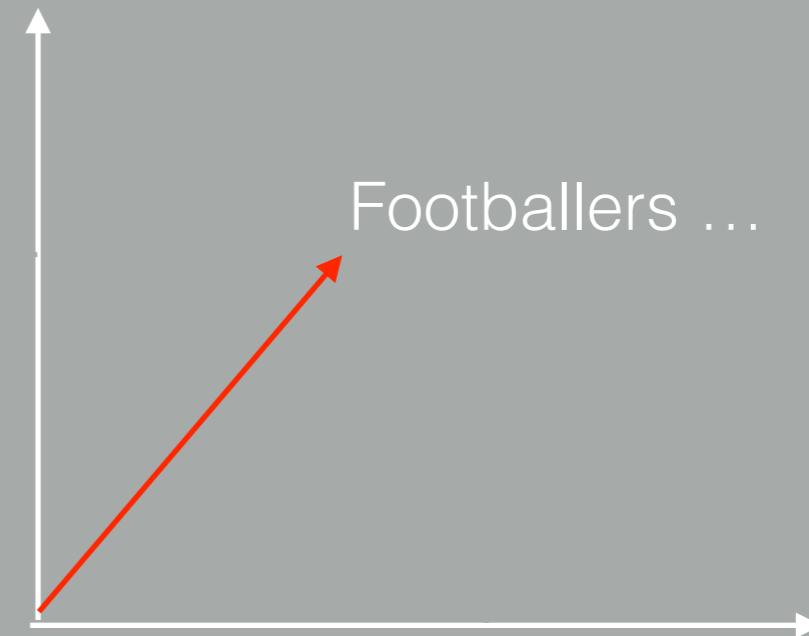
$$T_i^{babies} T_{ij}^{score}$$



Incremental Utterances

“Footballers ...”

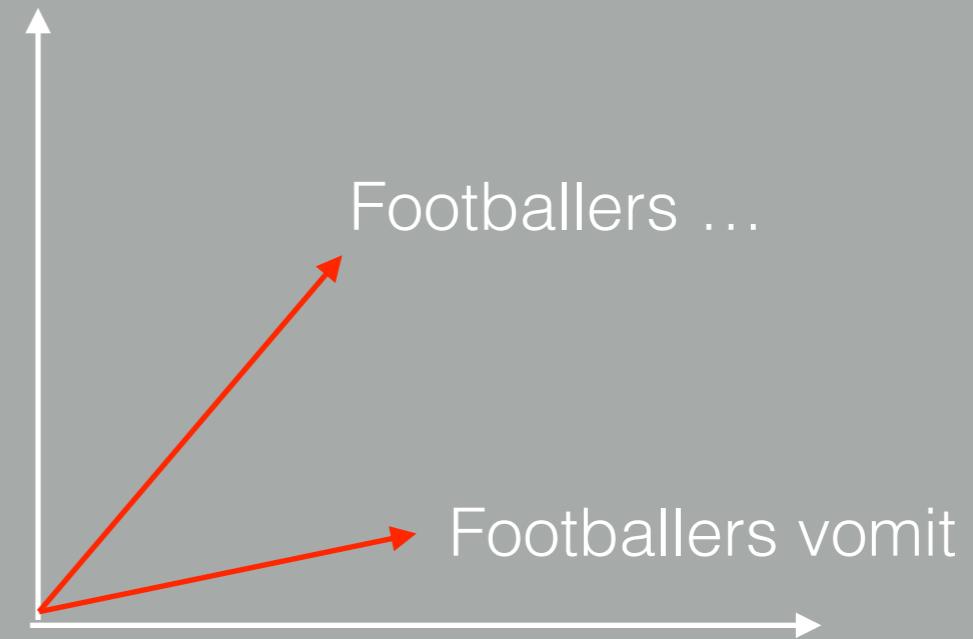
$T^{footballers} T^+$



Incremental Utterances

“Footballers vomit”

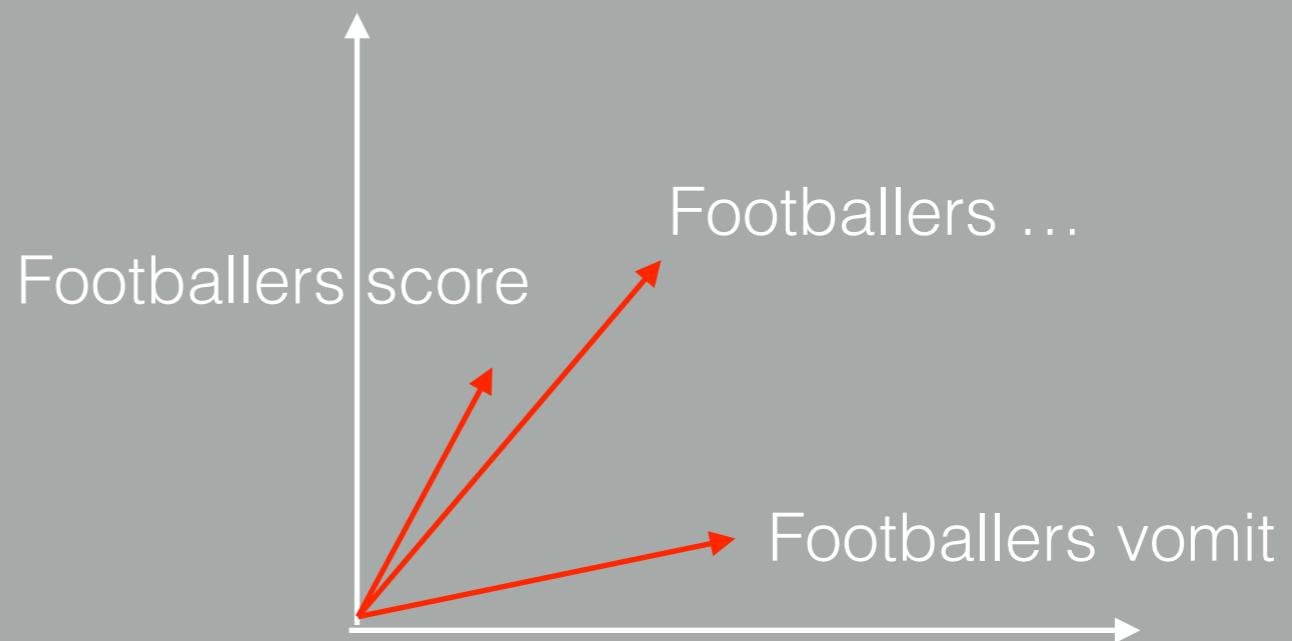
$$T^{footballers} T_{ij}^{vomit}$$



Incremental Utterances

“Footballers score”

$$T_i^{footballers} T_{ij}^{score}$$



Dataset

- Kartsaklis D., MS, Pulman S.: Separating disambiguation from composition in compositional distributional semantics.
- Chose ambiguous verbs and two landmark meanings from Pickering and Frisson 2001
- Picked subjects and objects for landmarks using most frequently occurring ones in the BNC

Dataset

- Pairs of subjects and complete sentences
 - (footballers ... , footballers dribble milk)
 - (footballers ... , footballers dribble ball)
- Pairs of subject+verb and complete sentences:
 - (footballers dribble ... , footballers dribble milk)
 - (footballers dribble ... , footballers dribble ball)
- Pairs of complete sentences:
 - (footballers dribble milk , footballers dribble ball)
 - (babies dribble milk , babies dribble ball)

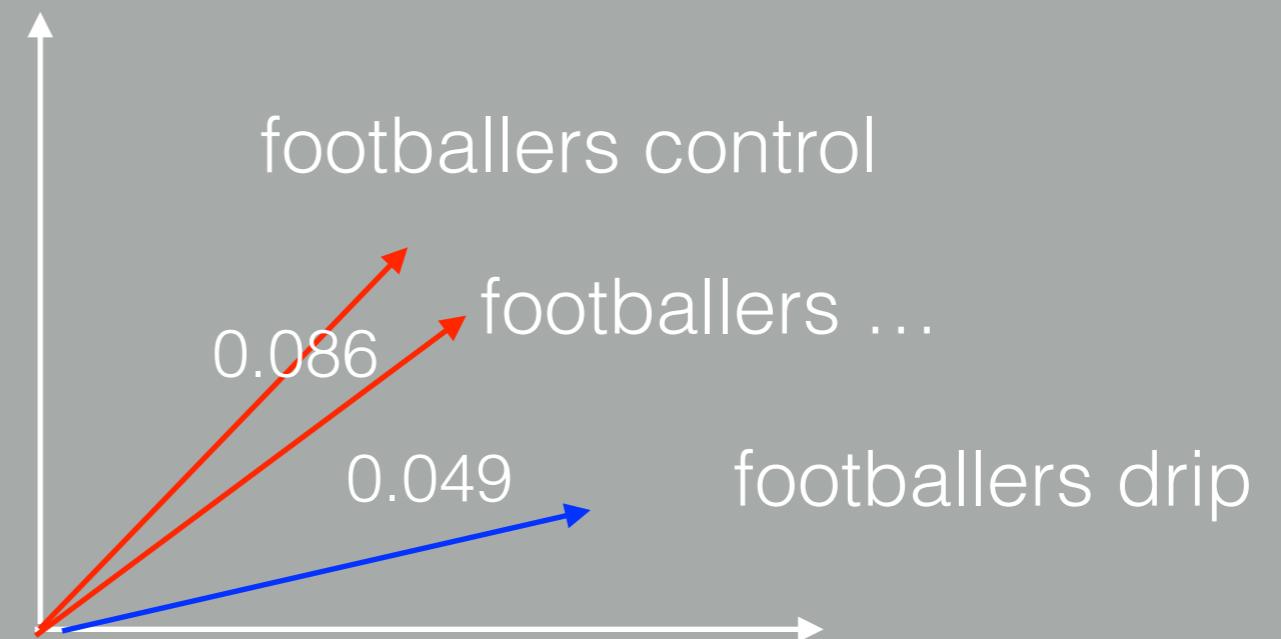
Data

Vectors: 300 Dim from Word2Vec,

Tensors: the G&S EMNLP 2011 method

Data

Just subject



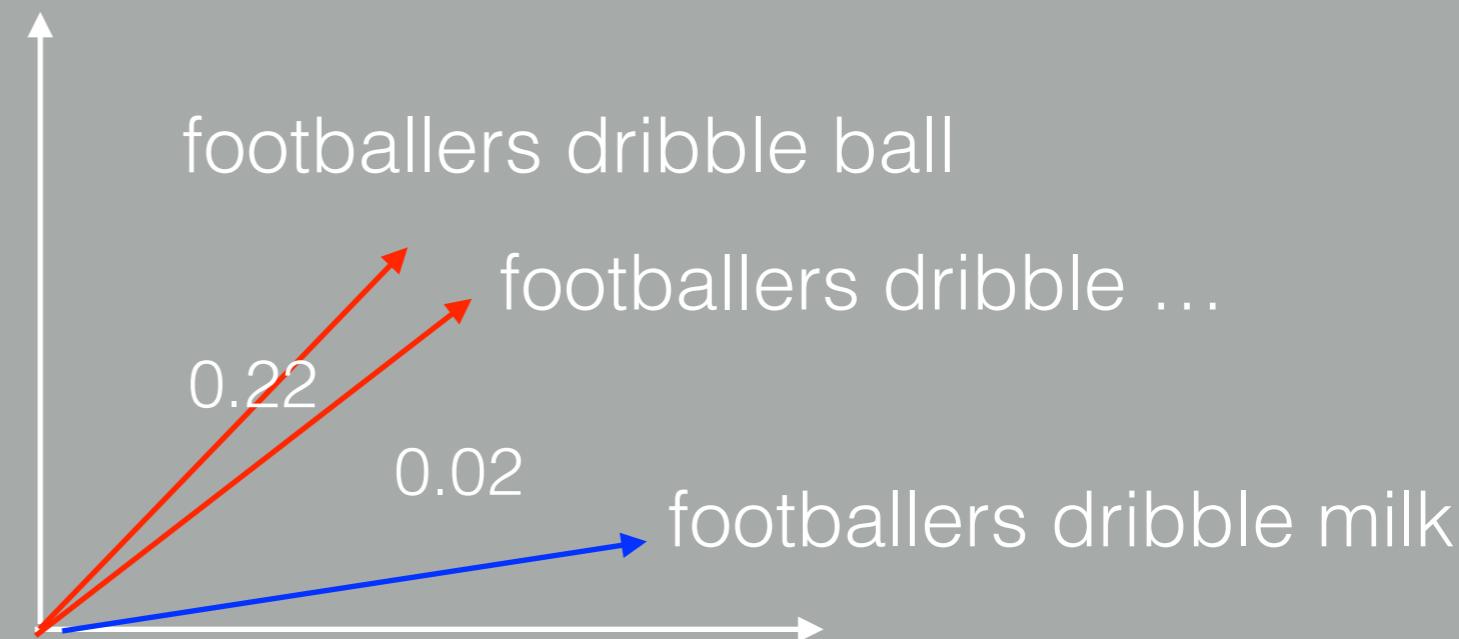
Data

Just Subject



Data

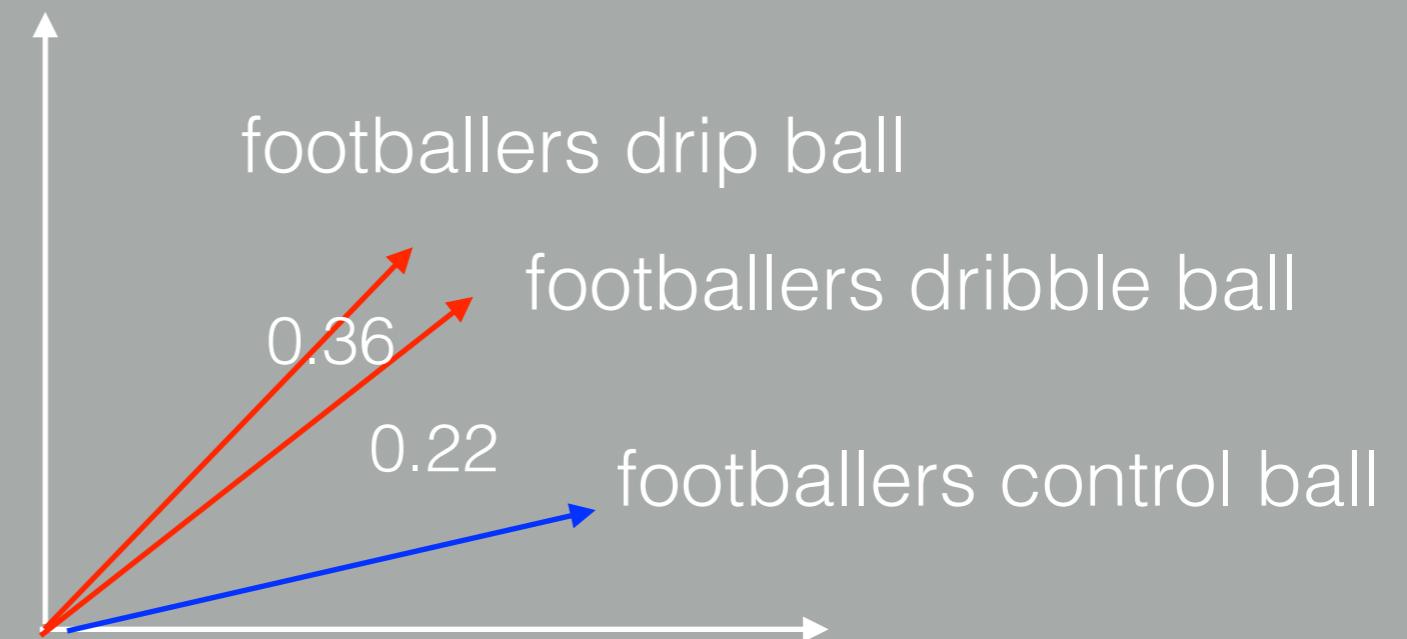
Subject + Verb



Data

Complete Sentences

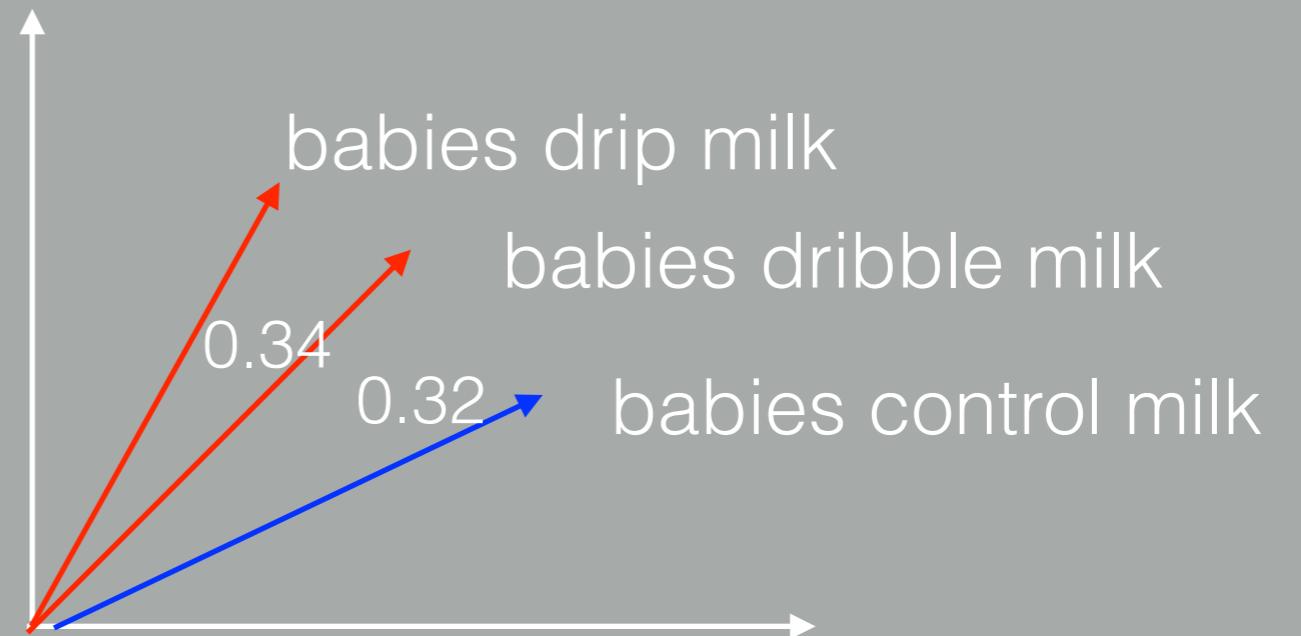
$0.22 < 0.36$



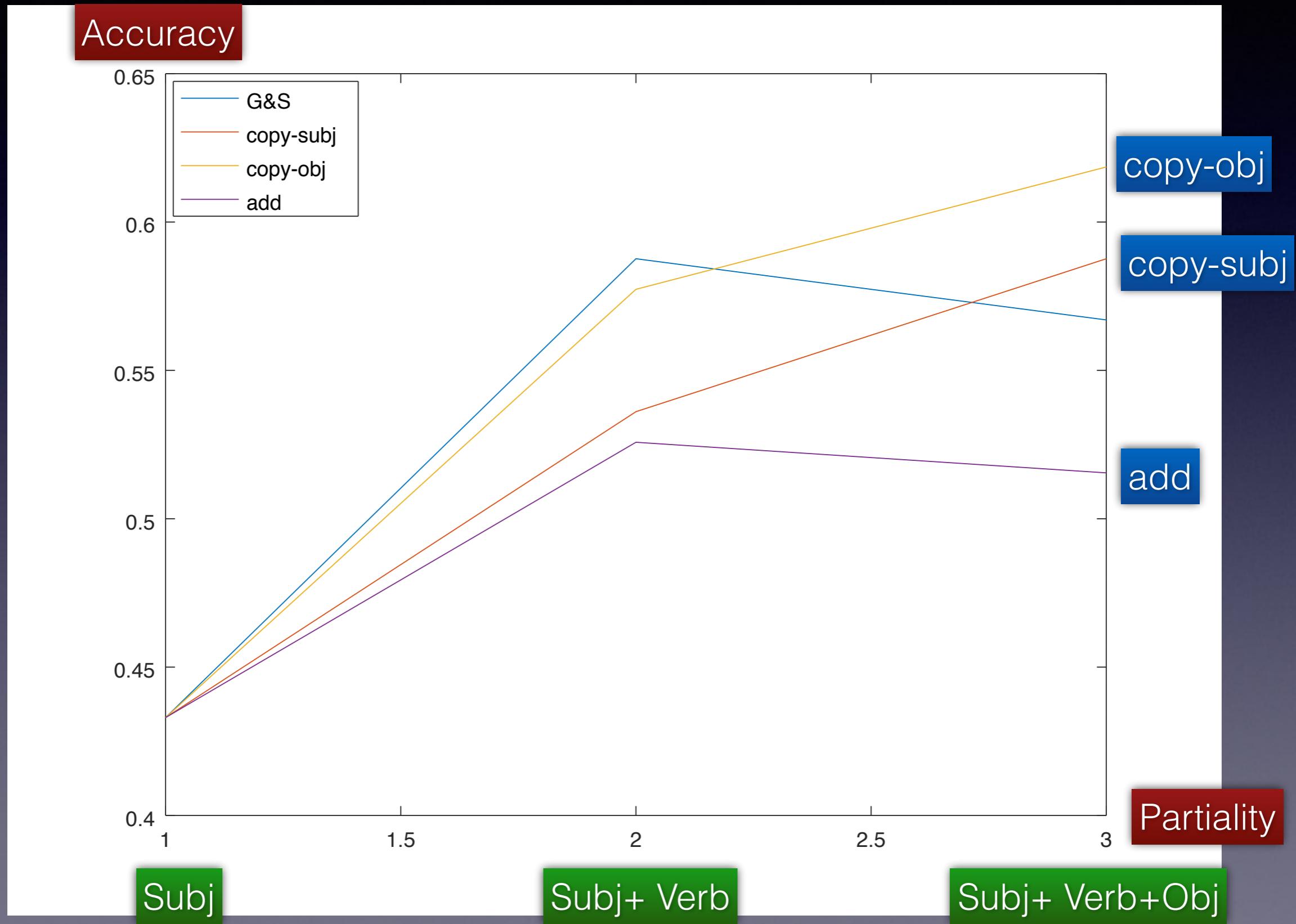
Data

Complete Sentences

$0.34 > 0.32$



Accuracy Results





Work in Progress

Implement the plausibilities model of Clark 2013, Polajnar
et al 2015
... under way ...

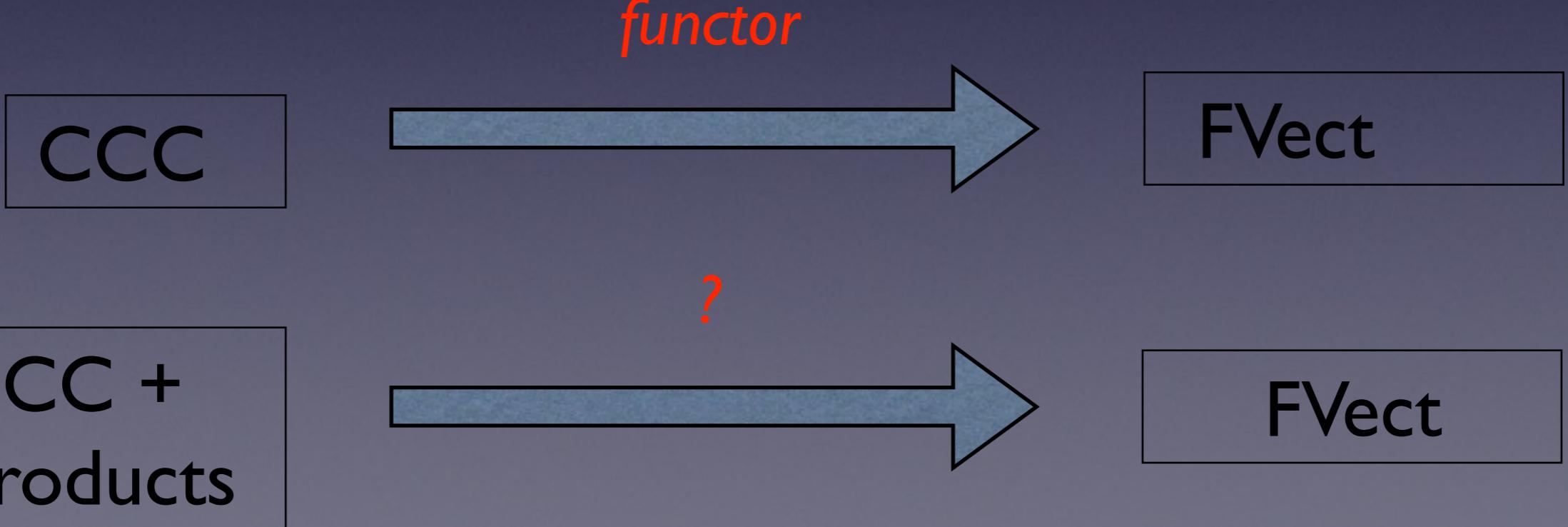
Extend it to experimental expectation predication

...

Incremental Understanding of Dialogue Content

...

Categorical Semantics



A: Thank ...



B: ... you!