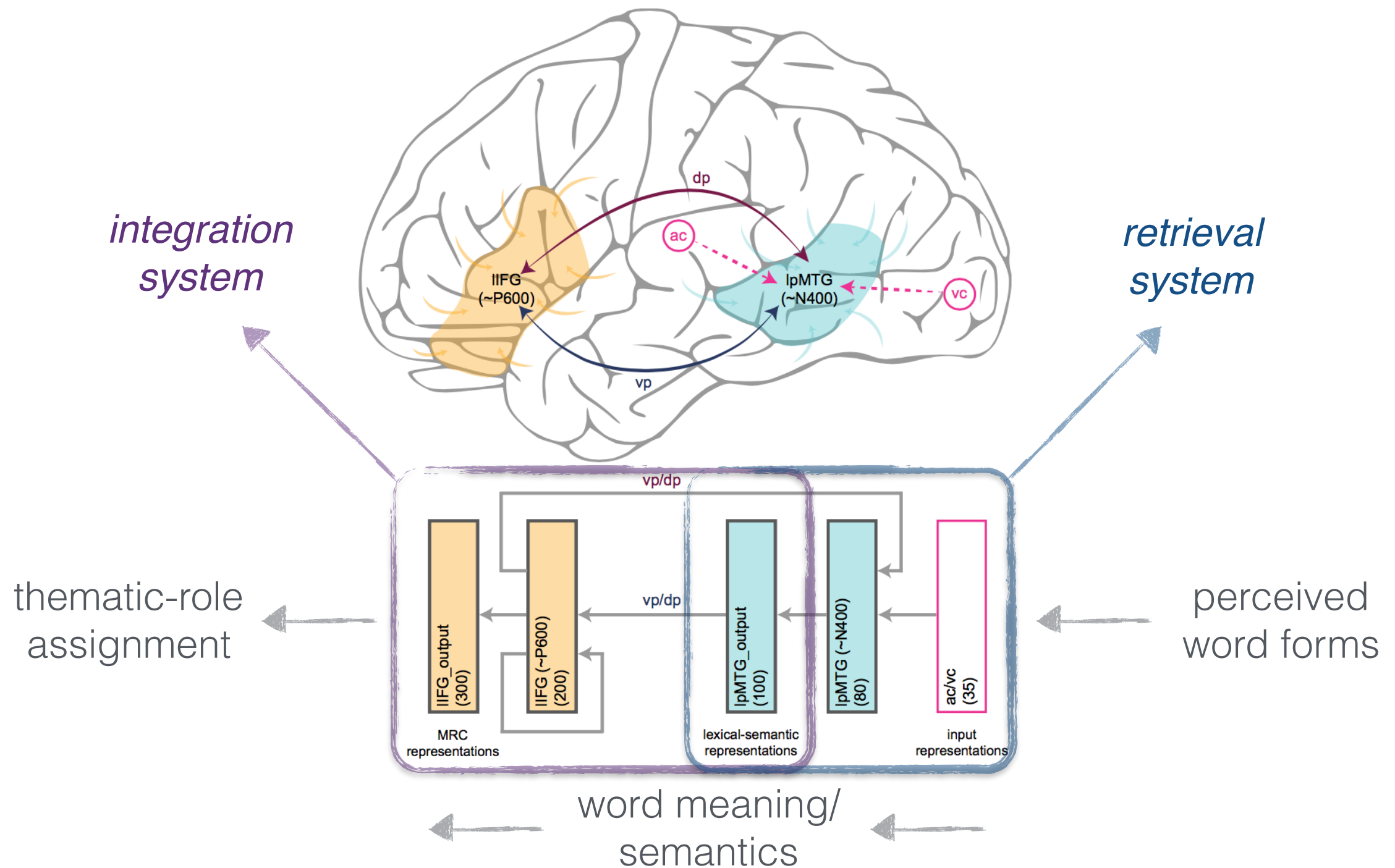


Connectionist Language Processing

Lecture 10: **Situation Modeling using Microworlds**

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A Neurocomputational Model



Sentence comprehension

“charlie plays soccer”

play(charlie,soccer)



Two requirements

A richer representational scheme

We need to *represent* that Charlie is outside, on a field, playing with a ball, and with others, etc.

Knowledge about the world

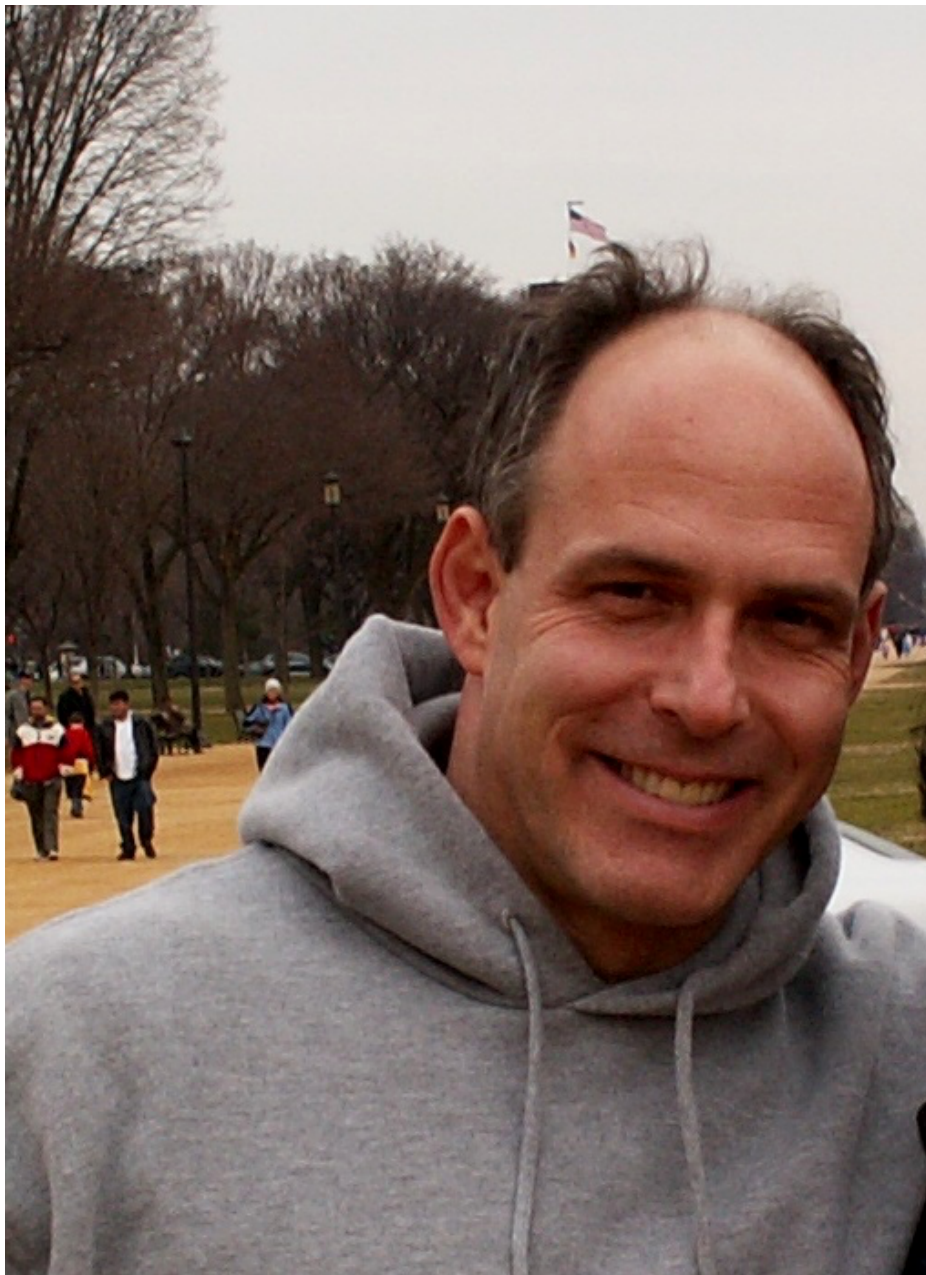
We need to *know* that Charlie is outside on a field, because soccer is typically played on a field, with a ball, with others, etc.

> **Solution:** the Distributed Situation Space (DSS) model

Distributed Situation Space (DSS)

- > A non-symbolic, distributed representational scheme for meaning
- > Situations are represented as vectors in a high-dimensional space called “*situation-state space*”
- > DSS vectors capture dependencies between situations, allowing for ‘world knowledge’-driven *direct inference*
- > To encode all world knowledge, DSS vectors are derived from observations of *states-of-affairs* (situations) in a *microworld*

Introducing... Golden & Rumelhart



DSS—The main idea

Take a snapshot of the world (“a sample”) at many different times, and for each snapshot write down the *full state-of-affairs* in the world



Next: extract regularities—*world knowledge*—from the full set of observations, and construct meaning representations (vectors) that encode this world knowledge

Problem: How to record full state-of-affairs in the world for each snapshot?

> use a confined *microworld* (which limits the scope of the world)

Defining a Microworld

A *state-of-affairs* (observation) in a microworld is defined in terms of *atomic events* that can be assigned a state (i.e., they can be *the case* or not *the case*)

Class	Variable	Class members (concepts)	#	Event name	#
People	p	charlie, heidi, sophia	3	$\text{play}(p, g)$	$3 \times 3 = 9$
Games	g	chess, hide&seek, soccer	3	$\text{play}(p, t)$	$3 \times 3 = 9$
Toys	t	puzzle, ball, doll	3	$\text{win}(p)$	3
Places	x	bathroom, bedroom, playground, street	4	$\text{lose}(p)$	3
Manners of playing	m_{play}	well, badly	2	$\text{place}(p, x)$	$3 \times 4 = 12$
Manners of winning	m_{win}	easily, difficultly	2	$\text{manner}(\text{play}(p), m_{\text{play}})$	$3 \times 2 = 6$
Predicates	—	play, win, lose, place, manner	5	$\text{manner}(\text{win}, m_{\text{win}})$	2
					Total 44

More specifically, states-of-affairs are combinations of these 44 atomic events

Example—“heidi loses at chess”: $\text{play}(\text{heidi}, \text{chess}) \wedge \text{lose}(\text{heidi})$

> 2^{44} ($\approx 10^{13}$) possible situations, but world knowledge precludes many

Microworld knowledge

World knowledge enforces constraints on event co-occurrence. Some examples:

Personal characteristics—each person has a specialty, a preferred toy, and some persons frequent specific places

Games and toys—each game/toy can only be played (with) in specific places, and has a number of possible player configurations; soccer is played with a ball

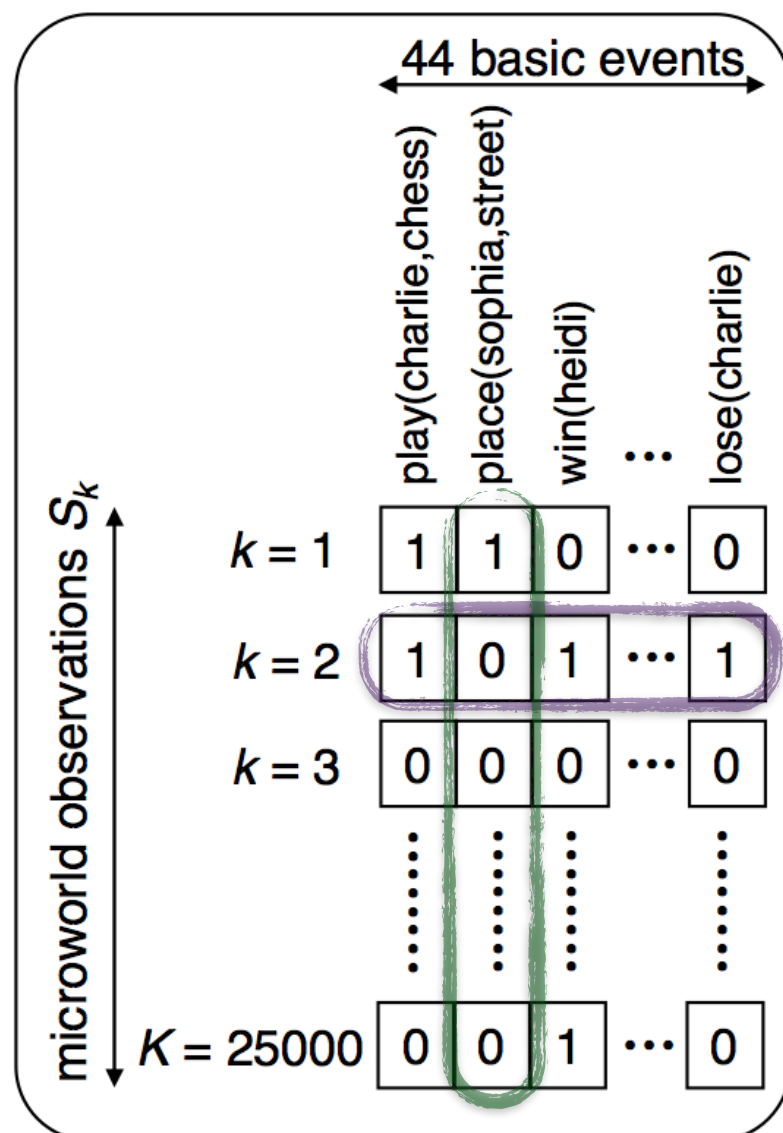
Being there—everybody is exactly at one place; if hide&seek is played in the playground, all players are there; all chess players are in the same place

Winning and losing—only one can win, and one cannot win and lose; if someone wins, all other players lose

Note: there are *hard* (being there) and *probabilistic* (preferences) constraints

Situation-state space

Many samples of microworld observations constitute a “*situation-state space*”



Rows represent observations (states-of-affairs)

Columns represent situation vectors for atomic events:

$$\vec{v}(a) = (\vec{v}_1(a), \dots, \vec{v}_n(a)) \quad (\text{a point in situation space})$$

Using (fuzzy) logic, *complex event* vectors can be derived:

$$\vec{v}(\neg a) = 1 - \vec{v}(a)$$

$$\vec{v}(a \wedge b) = \vec{v}(a)\vec{v}(b) \quad \text{where} \quad \vec{v}(a \wedge a) = \vec{v}(a)$$

which gives *functional completeness*:

$$\vec{v}(a \uparrow b) = \vec{v}(\neg \vec{v}(a \wedge b))$$

Situation vectors

Situation vectors encode events by means of *co-occurrence probabilities*

Prior belief in atomic event a (= estimate of its probability):

$$B(a) = \frac{1}{k} \sum_i \vec{v}_i(a) \approx Pr(a)$$

Prior conjunction belief of atomic events a and b :

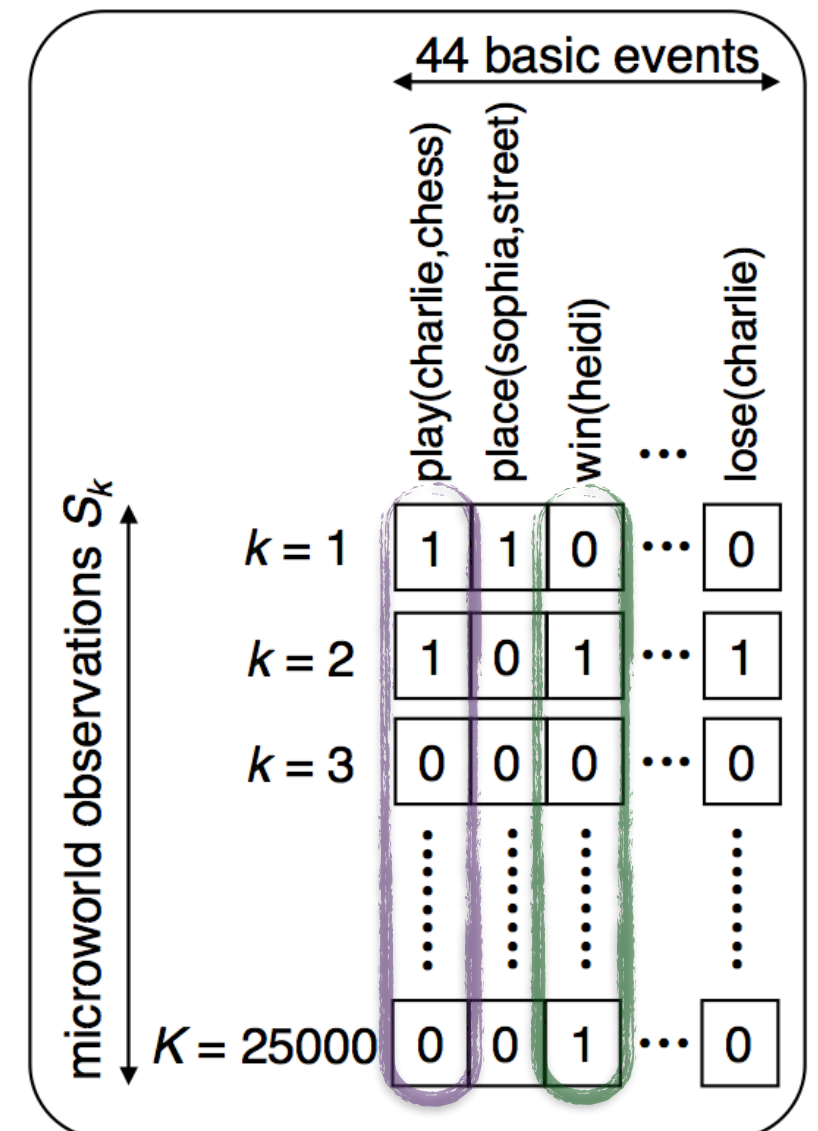
$$B(a \wedge b) = \frac{1}{k} \sum_i \vec{v}_i(a) \vec{v}_i(b) \approx Pr(a \wedge b) \quad \text{where} \quad B(a \wedge a) = B(a)$$

Prior conditional belief of atomic event a given b :

$$B(a|b) = \frac{B(a \wedge b)}{B(b)} \approx Pr(a|b)$$

Critically, either a and/or b can be atomic or complex events

$B(a|b) \approx Pr(a|b)$ means $\vec{v}(b)$ encodes b and all that depends upon b ; this allows 'world knowledge'-driven inference



Quantifying “comprehension”

Beyond conditional belief—how much is *a* ‘understood’ from *b*?

If *a* is **understood to be the case** from *b*, the conditional belief $B(a|b)$ should be higher than the prior belief $B(a)$: knowing *b* increases belief in *a*

If *a* is ***understood not to be the case*** from *b*, the conditional belief $B(a|b)$ should be lower than the prior belief $B(a)$: knowing *b* decreases belief in *a*

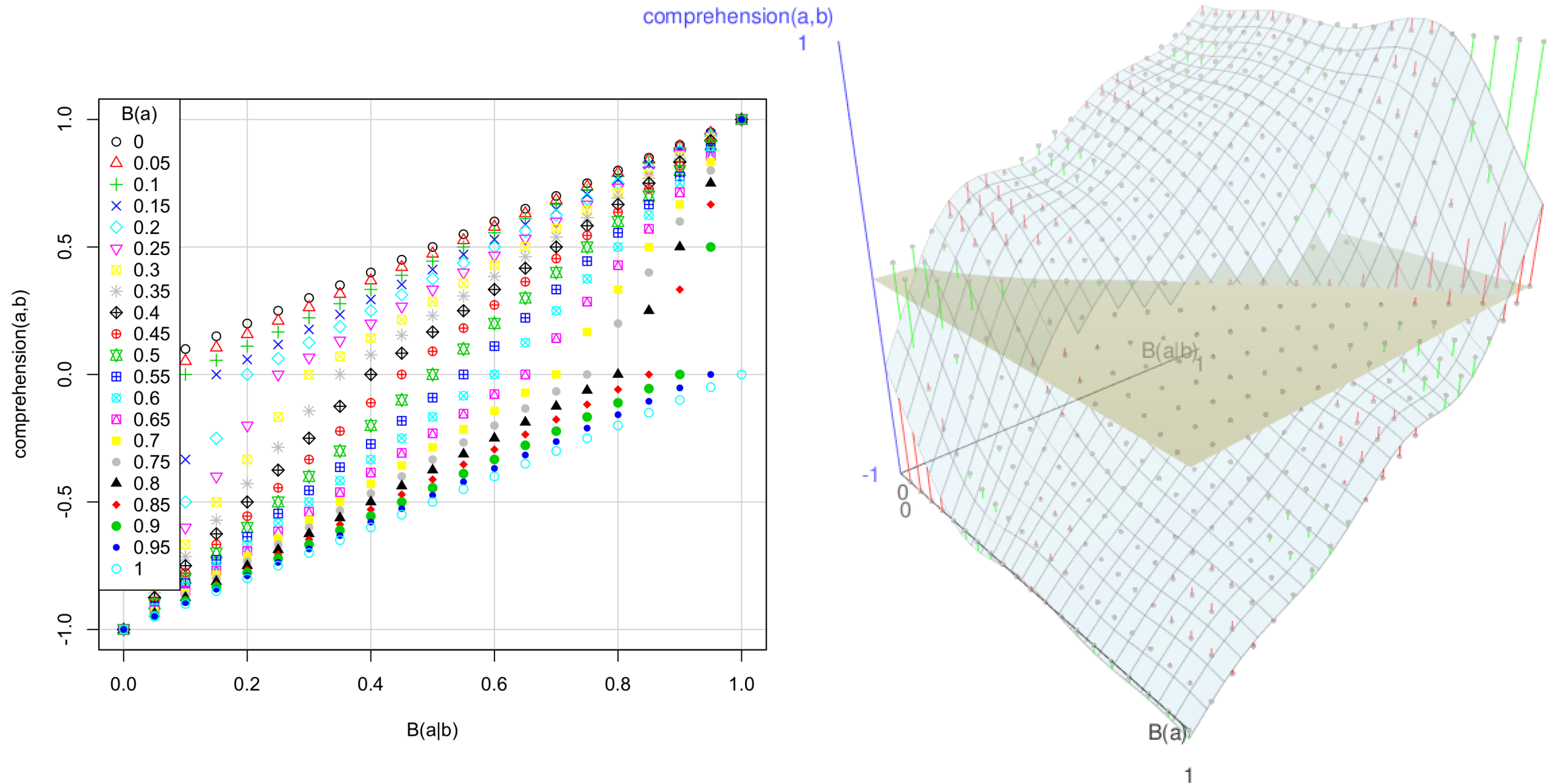
$$\textit{comprehension}(a, b) = \begin{cases} \frac{B(a|b) - B(a)}{1 - B(a)} & \text{if } B(a|b) > B(a) \\ \frac{B(a|b) - B(a)}{B(a)} & \text{otherwise} \end{cases}$$

$-1 \leq \textit{comprehension}(a, b) \leq +1$:

+1 indicates perfect **positive** comprehension: *b* took away all uncertainty in *a*

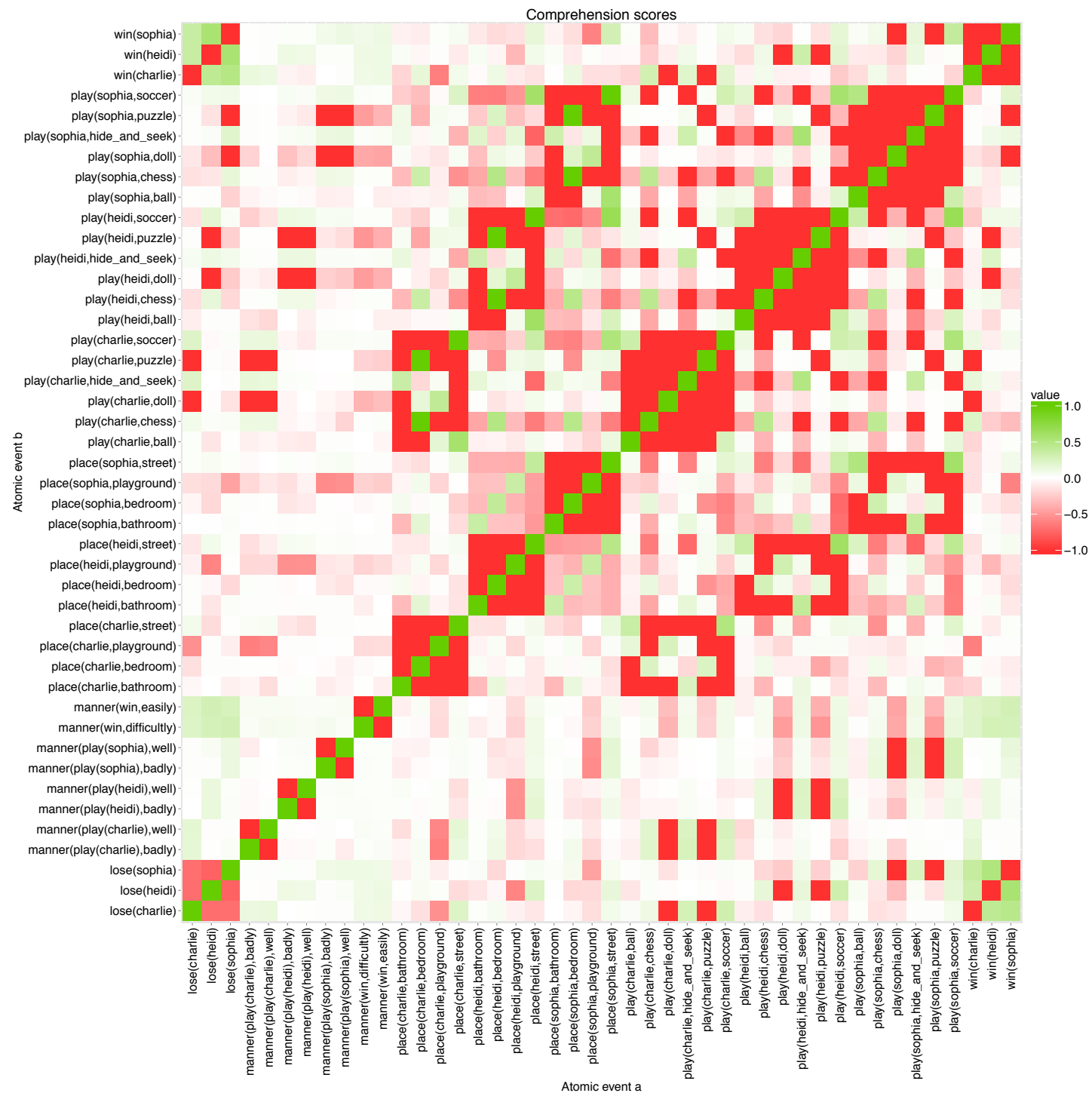
-1 indicates perfect **negative** comprehension: *b* took away all certainty in *a*

Comprehension scores



The higher $B(a)$ the more difficult it is to *increase certainty* in a , and the lower $B(a)$ the more difficult it is to *increase uncertainty* in a

Map of the World



Zooming in: Observation sampling

Q: how to efficiently sample k observation from 2^{44} possibilities, such that *no observation violates world knowledge* and the *set of samples reflects the probabilistic nature of the world*?

> inference-driven, incremental sampling algorithm using three-valued logic (0.5: Unknown)

Step 0—start with a completely undefined observation (all n atomic event states set to 0.5);

Step 1—pick a random, undecided atomic event e ;

Step 2—set e to be the case (1) or not (0) on basis of its probability given the observation so far;

Step 3—draw all inferences that follow from deciding the state of event e ;

(a) randomly pick the next, undecided event e' ;

(b) construct two observations: $s1$ in which e' is the case (1), and $s2$ in which it is not (0)

(c) check for $s1$ and $s2$ if they violate any hard world knowledge constraints:

i. both $s1$ and $s2$ are felicitous: state of e' cannot be inferred (and remains 0.5)

ii. only $s1$ is felicitous: infer e' to be the case (its state is set to 1)

iii. only $s2$ is felicitous: infer e' not to be the case (its state is set to 0)

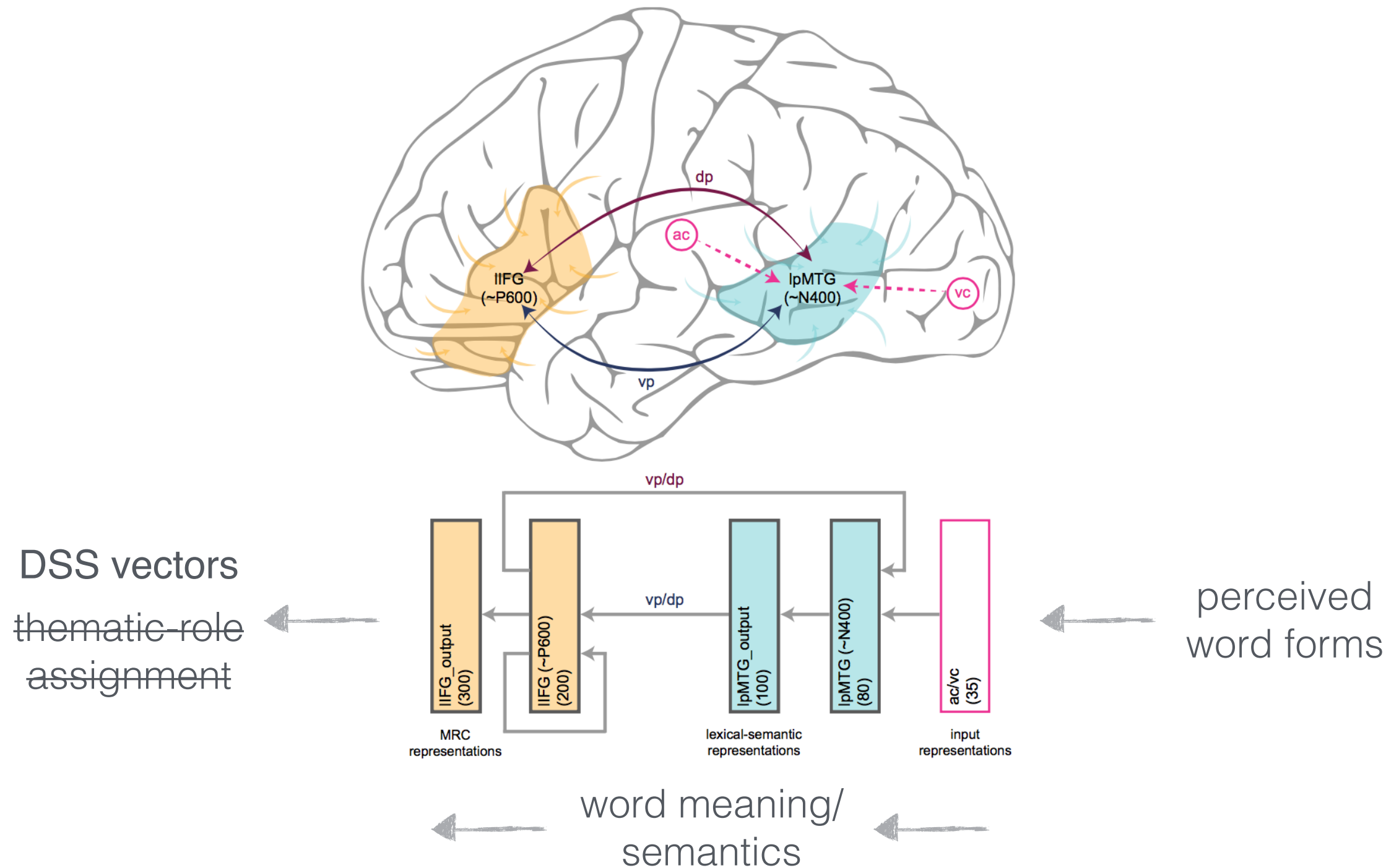
iv. both $s1$ and $s2$ are infelicitous: prior observation is infelicitous (restart from **step 0**)

(d) repeat (b) until it has been tried to infer each undecided event

Step 4—repeat **step 1** until there are no more undecided events

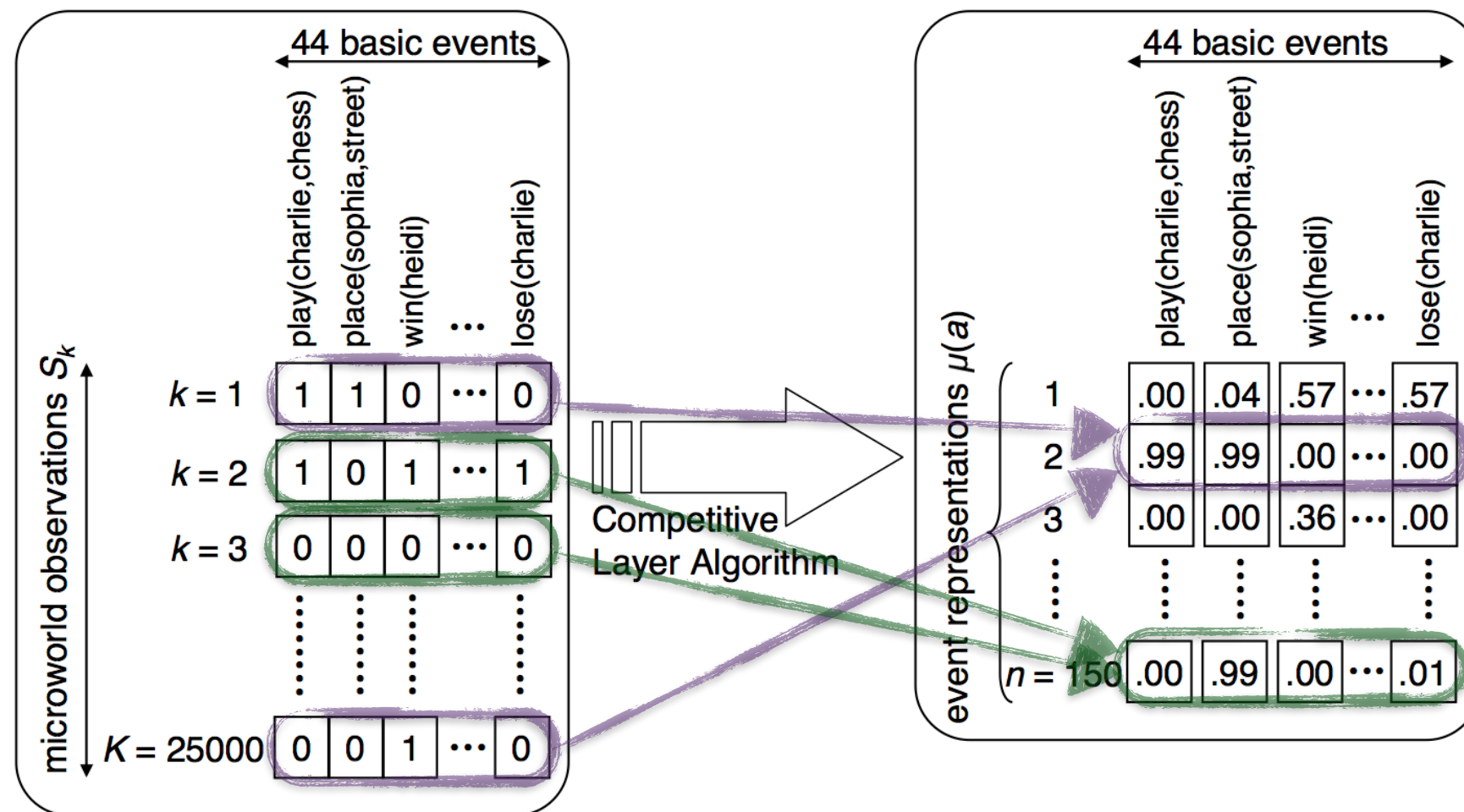
Prize of incrementality: need to deal with undecidedness in checking world knowledge violations

DSS vectors—Plug and Play?



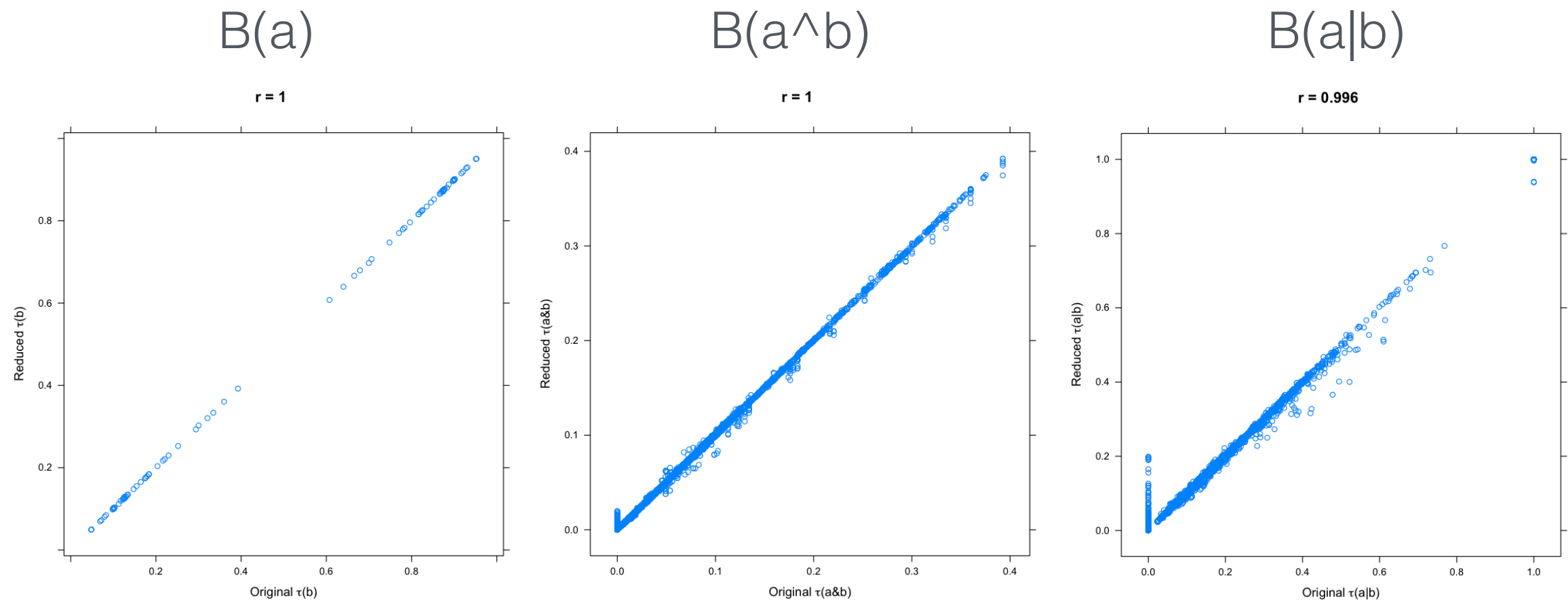
Reduced situation vectors

Problem: Situation vectors are 25K-dimensional, and are hence far larger than the thematic-role assignment vectors (300D) in our neurocomputational model



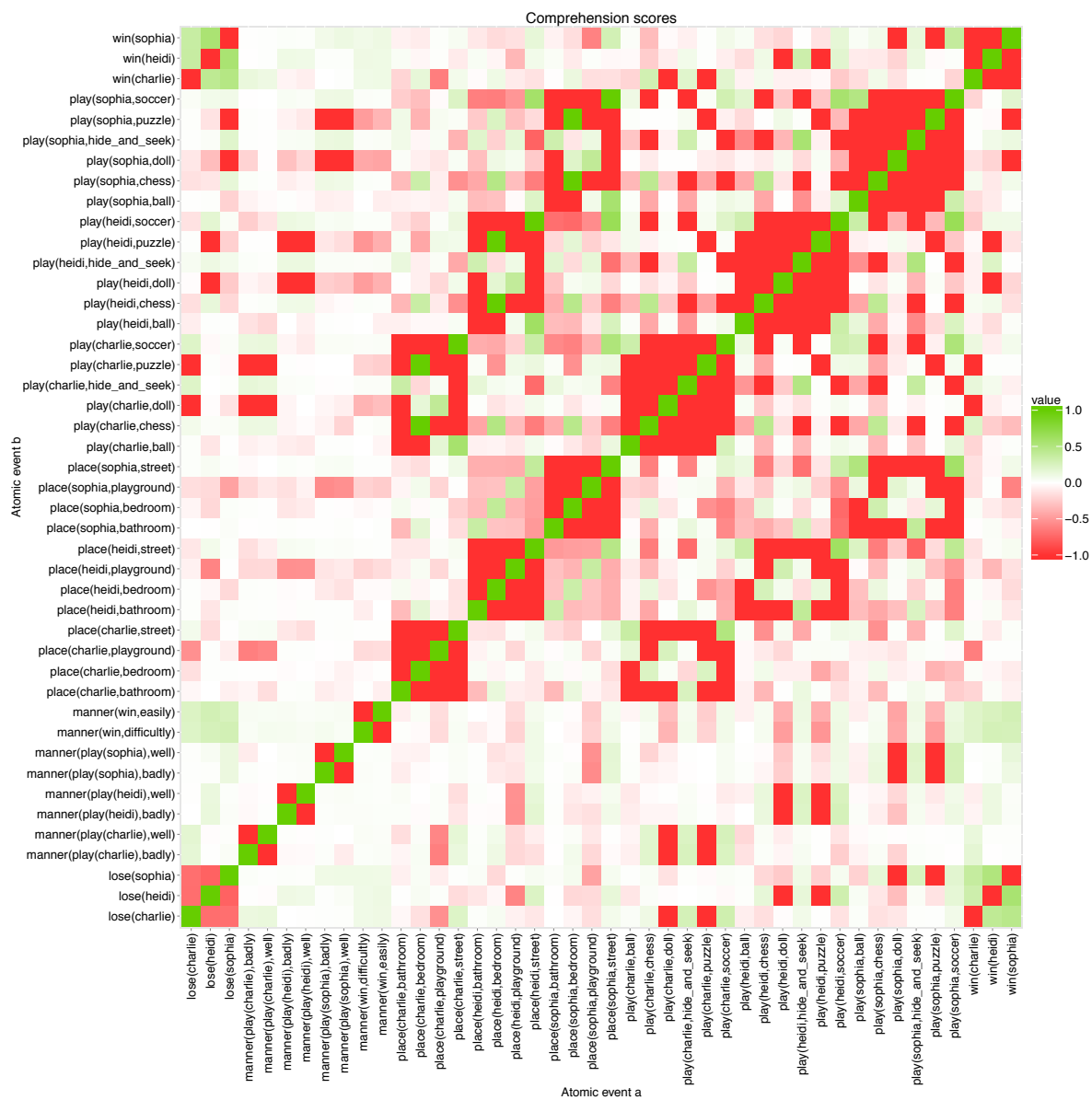
> justified if belief values estimated from original and reduced vectors are similar

Correlating belief values



Comparing world maps

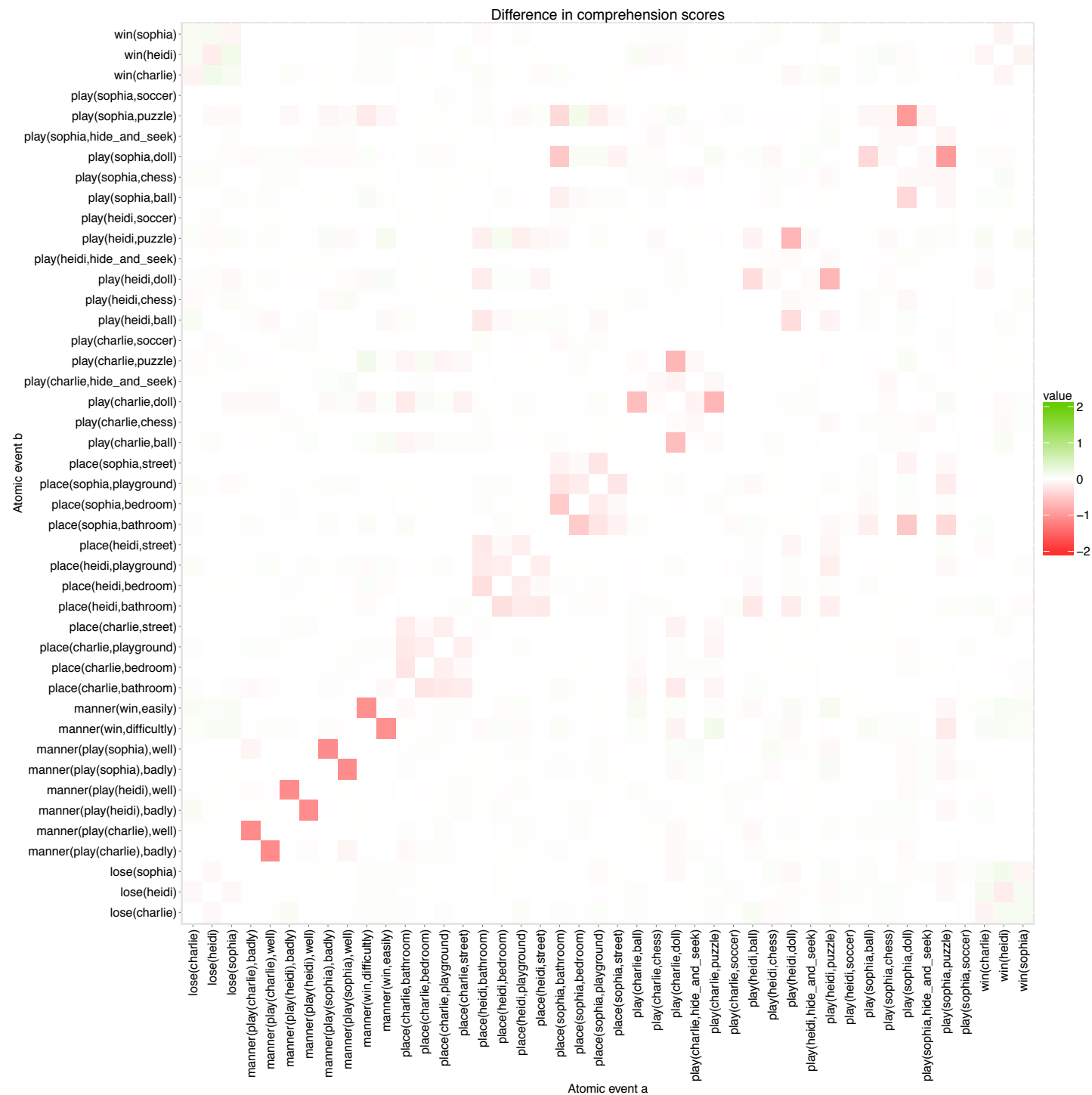
Original



Reduced



The “difference” world



Zooming in: Reducing dimensionality

Q: How to go from a $m \times n$ (25K x 44) to a $k \times n$ (k=150) situation-state space?

> employ a competitive layer algorithm to classify the m observations into k classes

Step 0—define a $k \times n$ matrix with all cells set to 0.5, and assign each row a bias $b=1$;

Step 1—for each observation m_i in the original situation-state space, repeat:

(a) determine the cityblock distance between row m_i and each row k_j :

$$\text{dist}(m_i, k_j) = \sum_c |m_{i,c} - k_{j,c}| \quad \text{where } c \text{ is a column index}$$

(b) determine row k_w with the shortest, biased distance to m_i :

$$k_w = \text{argmin}_j (\text{dist}(m_i, k_j) - b_j)$$

(c) and update row k_w by means of

$$\Delta k_w = \alpha(m_i - k_w) \quad \text{where } \alpha \text{ is a learning rate parameter}$$

(d) next, decrease the bias b_w of row k_w (to a minimum of 1):

$$\Delta b_w = \beta b_w (1 - b_w) \quad \text{where } \beta \text{ is a learning rate parameter}$$

(d) and, increase the bias of all other rows $k_j \neq k_w$:

$$\Delta b_j = \beta b_j$$

Step 2—repeat **step 1** for N training epochs

From sentences to vectors

We want to train the model to map sequences of words constituting a sentence onto the DSS vector representing the meaning of this sentence

Q: How to go from sentences to DSS vectors?

Path: **sentence** —> **propositional logic form** —> **DSS vector**

Defining a Microlanguage—Lexicon

Class	Words	#
proper nouns	<i>charlie, heidi, sophia</i>	3
(pro)nouns	<i>boy, girl, someone, chess, hide-and-seek, soccer, football, game, puzzle, ball, doll, jigsaw, toy, ease, difficulty, bathroom, bedroom, playground, shower, street</i>	20
verbs	<i>wins, loses, beats, plays, is, won, lost, played</i>	8
adverbs	<i>well, badly, inside, outside</i>	4
prepositions	<i>with, to, at, in, by</i>	5
Total		40

Defining a Microlanguage—Grammar

S	→	N _n VP _{n,v} APP _{n,v}	APP _{person, play}	→	[N _{game}] [Manner] [Place] PP _{toy} [Place] Place PP _{toy}
N _{person}	→	<i>charlie</i> <i>heidi</i> <i>sophia</i> <i>someone</i> <i>boy</i> <i>girl</i>	APP _{person, win}	→	[PP _{manner}] [PP _{game}] [Place] PP _{game} PP _{manner} Place PP _{game}
N _{game}	→	<i>chess</i> <i>hide-and-seek</i> <i>soccer</i> <i>football</i> <i>game</i>	APP _{person, lose}	→	[PP _{game}] [Place] Place PP _{game}
N _{toy}	→	<i>puzzle</i> <i>ball</i> <i>doll</i> <i>jigsaw</i> <i>toy</i>	APP _{game, play}	→	[Manner] [PP _{person}] [Place]
VP _{person, play}	→	<i>plays</i>	APP _{game, win}	→	[PP _{manner}] [PP _{person}] [Place]
VP _{person, win}	→	<i>wins</i> <i>beats</i> N _{person}	APP _{game, lose}	→	[PP _{person}] [Place]
VP _{person, lose}	→	<i>loses</i> <i>loses to</i> N _{person}	APP _{toy, play}	→	[PP _{person}] [Place] Place PP _{person}
VP _{game, play}	→	<i>is played</i>	Manner	→	<i>well</i> <i>badly</i>
VP _{game, win}	→	<i>is won</i>	Place	→	<i>inside</i> <i>outside</i> PP _{place}
VP _{game, lose}	→	<i>is lost</i>	PP _{place}	→	<i>in bathroom</i> <i>in shower</i> <i>in bedroom</i> <i>in street</i> <i>in playground</i>
VP _{toy, play}	→	<i>is played with</i>	PP _{person}	→	<i>by</i> N _{person}
			PP _{game}	→	<i>at</i> N _{game}
			PP _{toy}	→	<i>with</i> N _{toy}
			PP _{manner}	→	<i>with ease</i> <i>with difficulty</i>

> this grammar generates 13.556 different sentences

Defining a Microlanguage—Semantics

<i>charlie plays chess</i>	$\text{play}(c, \text{chess})$
<i>chess is played by charlie</i>	$\text{play}(c, \text{chess})$
<i>girl plays chess</i>	$\text{play}(h, \text{chess}) \vee \text{play}(s, \text{chess})$
<i>heidi plays game</i>	$\text{play}(h, \text{chess}) \vee \text{play}(h, \text{hide\&seek}) \vee \text{play}(h, \text{soccer})$
<i>heidi plays with toy</i>	$\text{play}(h, \text{puzzle}) \vee \text{play}(h, \text{ball}) \vee \text{play}(h, \text{doll})$
<i>sophia plays soccer well</i>	$\text{play}(s, \text{soccer}) \wedge \text{manner}(\text{play}(s), \text{well})$
<i>sophia plays with ball in street</i>	$\text{play}(s, \text{ball}) \wedge \text{place}(s, \text{street})$
<i>someone plays with doll</i>	$\text{play}(c, \text{doll}) \vee \text{play}(h, \text{doll}) \vee \text{play}(s, \text{doll})$
<i>doll is played with</i>	$\text{play}(c, \text{doll}) \vee \text{play}(h, \text{doll}) \vee \text{play}(s, \text{doll})$
<i>charlie plays</i>	$\text{play}(c, \text{chess}) \vee \text{play}(c, \text{hide\&seek}) \vee \text{play}(c, \text{soccer})$ $\vee \text{play}(c, \text{puzzle}) \vee \text{play}(c, \text{ball}) \vee \text{play}(c, \text{doll})$

Logic forms —> Situation vectors

The situation vectors of *atomic events* are the columns of the situation-state matrix

The situation vectors of *complex events* can be found through *fuzzy logic*:

$$\vec{v}(\neg a) = 1 - \vec{v}(a)$$

$$\vec{v}(a \wedge b) = \vec{v}(a)\vec{v}(b) \quad \text{where} \quad \vec{v}(a \wedge a) = \vec{v}(a)$$

Which gives us $\vec{v}(a \uparrow b) = \vec{v}(\neg \vec{v}(a \wedge b))$ and hence *functional completeness*:

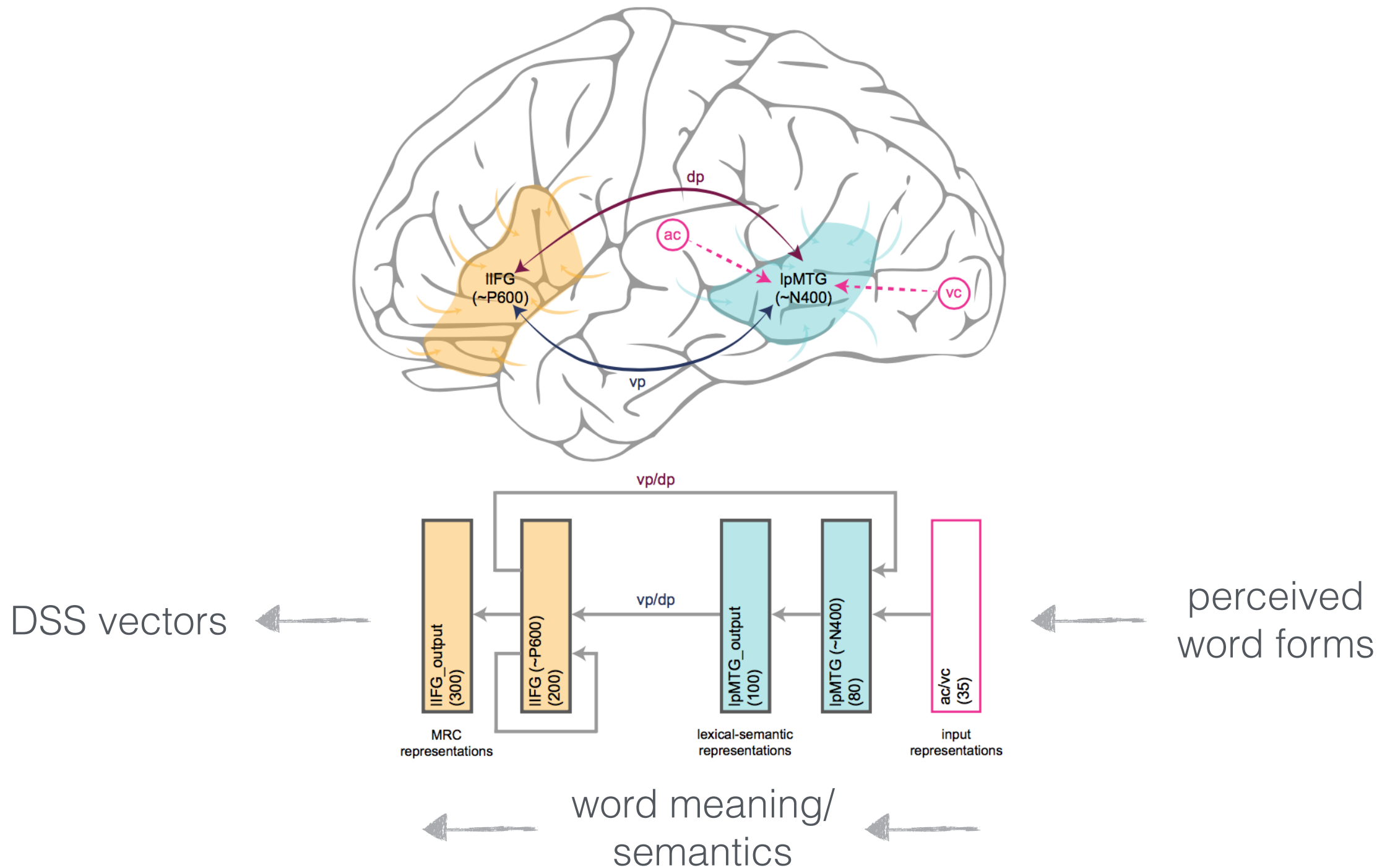
$$\vec{v}(a \vee b) = \vec{v}(\vec{v}(a \uparrow a) \uparrow \vec{v}(b \uparrow b))$$

$$\vec{v}(a \rightarrow b) = \vec{v}(a \uparrow \vec{v}(b \uparrow b)) = \vec{v}(a \uparrow \vec{v}(a \uparrow b))$$

$$\vec{v}(a \underline{\vee} b) = \vec{v}(\vec{v}(a \uparrow \vec{v}(a \uparrow b)) \uparrow \vec{v}(b \uparrow \vec{v}(a \uparrow b)))$$

> allows to derive vectors for events of arbitrary logical complexity

DSS vectors—Plug and Play!



What does the model ‘understand’?

Given a sentence describing an atomic or complex event e , the model will construct an output vector $\vec{v}(e')$ that is at best an approximation of $\vec{v}(e)$

> how well $\vec{v}(e')$ approximates $\vec{v}(e)$ is quantifiable through $comprehension(e, e')$

[if the model has ‘understood’ e , the conditional belief $B(e|e')$ should be higher than the prior belief $B(e)$, yielding a positive comprehension score, and vice versa; moreover if $B(e|e') = 1$ iff $e = e'$]

> we can also probe the state-of-affairs as ‘understood’ by the model, by computing $comprehension(a, e')$ for any other atomic or complex event a

[for instance, for all the atomic events in the microworld]

Note—we can do both of these things after the processing of each word, and hence investigate how a state-of-affairs unfolds on word-by-word basis

Putting it all together ...

```
model:all_sents> dssScores basic_events "charlie plays chess"

**** Sentence: charlie plays chess
**** Semantics: play(charlie,chess)
****
****                                     charlie           plays           chess
****
****                                +0.08693   -0.01071   +0.07622   +0.72407   +0.80029
****
**** play(charlie,chess)          +0.08693   -0.01071   +0.07622   +0.72407   +0.80029   play(charlie,chess)
**** play(charlie,hide_and_seek) +0.04279   -0.01568   +0.02710   -0.77386   -0.74676   play(charlie,hide_and_seek)
**** play(charlie,soccer)        +0.12169   -0.03329   +0.08841   -0.87402   -0.78562   play(charlie,soccer)
**** play(heidi,chess)           +0.01111   +0.00486   +0.01597   +0.41746   +0.43343   play(heidi,chess)
**** play(heidi,hide_and_seek)   -0.08301   -0.00589   -0.08890   -0.77328   -0.86218   play(heidi,hide_and_seek)
**** play(heidi,soccer)          +0.00730   -0.00709   +0.00021   -0.88450   -0.88429   play(heidi,soccer)
**** play(sophia,chess)          +0.00809   +0.00035   +0.00844   +0.35767   +0.36612   play(sophia,chess)
**** play(sophia,hide_and_seek)  -0.11984   -0.00710   -0.12694   -0.71115   -0.83810   play(sophia,hide_and_seek)
**** play(sophia,soccer)         +0.03251   -0.02039   +0.01211   -0.90027   -0.88816   play(sophia,soccer)
**** play(charlie,puzzle)        -0.03140   +0.04140   +0.01001   -0.69627   -0.68626   play(charlie,puzzle)
**** play(charlie,ball)          +0.01103   -0.00114   +0.00989   -0.81493   -0.80505   play(charlie,ball)
**** play(charlie,doll)          -0.23204   +0.11227   -0.11977   -0.67018   -0.78995   play(charlie,doll)
**** play(heidi,puzzle)          -0.12599   +0.04865   -0.07733   -0.12012   -0.19746   play(heidi,puzzle)
**** play(heidi,ball)            +0.00373   -0.00318   +0.00055   -0.37581   -0.37527   play(heidi,ball)
**** play(heidi,doll)            -0.00341   -0.00959   -0.00618   -0.20651   -0.21269   play(heidi,doll)
**** play(sophia,puzzle)         -0.04606   -0.03433   -0.08039   +0.08060   +0.00021   play(sophia,puzzle)
**** play(sophia,ball)           +0.00915   -0.00004   +0.00912   -0.40594   -0.39683   play(sophia,ball)
**** play(sophia,doll)           -0.03115   +0.02611   -0.00504   -0.07983   -0.08487   play(sophia,doll)
**** win(charlie)                +0.09208   -0.03814   +0.05393   +0.19446   +0.24839   win(charlie)
**** win(heidi)                  +0.01076   -0.10392   -0.09316   -0.11974   -0.21290   win(heidi)
**** win(sophia)                 +0.01201   -0.06818   -0.05617   -0.34537   -0.40154   win(sophia)
**** lose(charlie)               +0.08350   -0.04772   +0.03578   -0.00125   +0.03453   lose(charlie)
**** lose(heidi)                 +0.01591   -0.01308   +0.00283   +0.05884   +0.06167   lose(heidi)
**** lose(sophia)                +0.02748   -0.02746   +0.00001   +0.06213   +0.06214   lose(sophia)
**** place(charlie,bathroom)      -0.45570   -0.04047   -0.49617   -0.40434   -0.90052   place(charlie,bathroom)
**** place(charlie,bedroom)       +0.11246   +0.01725   +0.12972   +0.71078   +0.84050   place(charlie,bedroom)
**** place(charlie,playground)    -0.24599   +0.07566   -0.17033   -0.65517   -0.82550   place(charlie,playground)
**** place(charlie,street)        +0.07425   -0.02316   +0.05109   -0.86119   -0.81011   place(charlie,street)
**** place(heidi,bathroom)        -0.01684   +0.02099   +0.00415   -0.44904   -0.44489   place(heidi,bathroom)
**** place(heidi,bedroom)         -0.00530   +0.00981   +0.00451   +0.45972   +0.46424   place(heidi,bedroom)
**** place(heidi,playground)      -0.02113   -0.01805   -0.03918   -0.29745   -0.33664   place(heidi,playground)
**** place(heidi,street)          +0.01236   -0.00930   +0.00307   -0.57345   -0.57038   place(heidi,street)
**** place(sophia,bathroom)       -0.03025   +0.03582   +0.00558   -0.34721   -0.34164   place(sophia,bathroom)
**** place(sophia,bedroom)        -0.01583   -0.00878   -0.02461   +0.43567   +0.41106   place(sophia,bedroom)
**** place(sophia,playground)     -0.05556   +0.01669   -0.03886   -0.15951   -0.19837   place(sophia,playground)
**** place(sophia,street)         +0.03195   -0.01436   +0.01759   -0.60041   -0.58283   place(sophia,street)
**** manner(play(charlie),well)   +0.04509   -0.00978   +0.03531   +0.05307   +0.08838   manner(play(charlie),well)
**** manner(play(charlie),badly) +0.04845   -0.01236   +0.03610   +0.05926   +0.09535   manner(play(charlie),badly)
**** manner(play(heidi),well)     -0.01155   +0.00267   -0.00888   +0.01330   +0.00442   manner(play(heidi),well)
**** manner(play(heidi),badly)    -0.02207   +0.00191   -0.02016   +0.01801   -0.00215   manner(play(heidi),badly)
**** manner(play(sophia),well)    +0.00287   -0.01083   -0.00797   -0.13684   -0.14481   manner(play(sophia),well)
**** manner(play(sophia),badly)   +0.00391   -0.00347   +0.00044   -0.02596   -0.02552   manner(play(sophia),badly)
**** manner(win,easily)           +0.01757   -0.01269   +0.00488   +0.03096   +0.03583   manner(win,easily)
**** manner(win,difficultly)      +0.01483   -0.01074   +0.00409   +0.01167   +0.01576   manner(win,difficultly)

model:all_sents>
```

Discussion

- > The DSS model provides a powerful framework for simulating rich situation model representations
- > We have employed the DSS representations in a model of interpretation-level Surprisal
- > We have also successfully used DSS vectors in a model of language production
- > The use of DSS vectors paves way towards modeling pragmatic phenomena