## The typology of veridicality inferences

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Slides available at aaronstevenwhite.io

Data available at megaattitude.io

## Introduction

## Overarching question

## How are a verb's semantic properties related to its syntactic distribution? Gruber 1965; Fillmore 1970; Zwicky 977; Iackendoff 1972;

Grimshaw 1979, 1990; Pesetsky 1982, 1991; Pinker 1989; Levin 1993

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## Semantic

Properties
$\left[\begin{array}{cc}+ & \text { Telic } \\ - & \text { durative } \\ - & \text { Stative } \\ & \cdots\end{array}\right]$

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Grimshaw 1979, 1990; Pesetsky 1982, 1991; Pinker 1989; Levin 1993


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## Factors claimed to affect the distribution of nominals

Sensitive to event structural properties like stativity, telicity, durativity, causativity, transfer, etc. (see Levin and Rappaport Hovav 2005)

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Factors claimed to affect the distribution of clauses
Sensitive to 'content-dependent' properties like representationality, preferentiality, factivity/veridicality, communicativity, etc. Bolinger 1968;
Hintikka 1975; Hooper 1975; Stalnaker 1984; Farkas 1985; Villalta 2000, 2008; Kratzer 2006; Egré 2008;
Scheffler 2009; Moulton 2009; Anand and Hacquard 2013; Rawlins 2013; Portner and Rubinstein
2013; Anand and Hacquard 2014; Spector and Egré 2015; Bogal-Allbritten 2016; Theiler et al. 2017

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2013; Anand and Hacquard 2014; Spector and Egré 2015; Bogal-Allbritten 2016; Theiler et al. 2017
Possibly indirectly, via e.g. neo-Davidsonian event decomposition
Kratzer 2006; Hacquard 2006; Moulton 2009; Anand and Hacquard 2013, 2014; Rawlins 2013;
Bogal-Allbritten 2016; White and Rawlins 2016b a.o.

## Our prior work

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How direct is the relationship between content-dependent properties and syntactic distribution?

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## Focus

Two content-dependent properties - factivity and veridicality - that are argued to determine selection of interrogatives \& declaratives

## Veridicality and factivity

## Veridicality

A verb $v$ is veridical iff NP V S entails S Karttunen 1971a; Egré 2008; Kartunen 2012;
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But there are strong empirical reasons to believe they do not.

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But there is substantial variability in the veridicality inferences generated with different complements - even for the same verb.

## Variability in veridicality

(3) a. Jo forgot that she $i_{i}$ bought tofu.

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(3) a. Jo ${ }_{i}$ forgot that she $e_{i}$ bought tofu. $\rightarrow$ Jo bought tofu.

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b. Jo knew to buy tofu. $\nrightarrow$ Jo \{bought, didn't buy\} tofu.

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1. Dataset capturing the variability of factivity and veridicality across finite and infinitival complement types.
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Analytical contributions

1. Inference pattern typology explains some parts of syntactic distribution reasonably well, but far from perfect.

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1. Dataset capturing the variability of factivity and veridicality across finite and infinitival complement types.
2. Data-driven typology of inference patterns across comp. types.

## Analytical contributions

1. Inference pattern typology explains some parts of syntactic distribution reasonably well, but far from perfect.
2. More likely that the veridicality-distribution relationship is indirect, mediated by fine-grained verb class.

## Outline

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A new veridicality dataset

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MegaVeridicality dataset (white and Rawlins, 2018)
Veridicality judgments for 517 verbs from the MegaAttitude based on their acceptability in the [NP _ that S] and [NP was _ed that S] frames

## Veridicality judgment task

61. Someone knew that a particular thing happened.

Did that thing happen?
no
maybe or maybe
not yes

How acceptable is the bolded sentence?
terrible $\begin{array}{lllllll}2 & 3 & 4 & 5 & 6 & \text { perfect }\end{array}$

## Veridicality judgment task

68. Someone didn't know that a particular thing happened.

Did that thing happen?

> maybe or maybe not
yes

How acceptable is the bolded sentence?

terrible | 2 | 3 | 4 | 5 | 6 | perfect |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |

## Stimuli

Expand MegaVeridicality with 603 verb types from MegaAcceptability based on acceptability in 7 frames involving infinitival complements:

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- [NP _ed for NP to VP] (184 verbs)


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## NP _ed for NP to VP

(5) a. Someone wanted for a particular thing to happen.
b. Someone didn't want for a particular thing to happen.

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- [NP _ed NP to VP[+ev]] (197 verbs)


## Stimuli

## NP _ed for NP to VP

(5) a. Someone wanted for a particular thing to happen.
b. Someone didn't want for a particular thing to happen.

NP _ed NP to VP[+ev]
(6) a. Someone told a particular person to do a particular thing.
b. Someone didn't tell a particular person to do a particular thing.

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(5) a. Someone wanted for a particular thing to happen.
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NP _ed NP to VP[+ev]
(6) a. Someone told a particular person to do a particular thing.
b. Someone didn't tell a particular person to do a particular thing.

NP _ed NP to VP[-ev]
(7) a. Someone believed a particular person to have a particular thing.
b. Someone didn't believe a particular person to have a particular thing.

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- [NP was _ed NP to VP[+ev]] (278 verbs)


## Stimuli

## NP was _ed to VP[+ev]

(8) a. A particular person was ordered to do a particular thing.
b. A particular person wasn't ordered to do a particular thing.

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- [NP was _ed NP to VP[-ev]] (256 verbs)


## Stimuli

NP was _ed to VP[+ev]
(8) a. A particular person was ordered to do a particular thing.
b. A particular person wasn't ordered to do a particular thing.

NP was _ed to VP[-ev]
(9) a. A particular person was overjoyed to have a particular thing.
b. A particular person wasn't overjoyed to have a particular thing.

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- [NP _ed to VP[+ev]] (217 verbs)


## Stimuli

NP _ed to VP[+ev]
(10) a. A particular person decided to do a particular thing.
b. A particular person didn't decide to do a particular thing.

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- [NP _ed to VP[+ev]] (217 verbs)
- [NP _ed to VP[-ev]] (165 verbs)


## Stimuli

NP _ed to VP[+ev]
(10) a. A particular person decided to do a particular thing.
b. A particular person didn't decide to do a particular thing.

NP _ed to VP[-ev]
(11) a. A particular person hoped to have a particular thing.
b. A particular person didn't hope to have a particular thing.

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2,850 items randomly partitioned into 50 lists of 57

## Results

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Mixed-effects ordinal model-based normalization to control for variability in how participants use the response scale. (see Agresti, 2014)

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## Intuition

Like z-scoring, but better models response behavior.

## Data overview

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## Example: $x$-axis

A particular person didn't forget to do a particular thing.

## Results



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## Example: $x$-axis

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## Example: $x$-axis

A particular person didn't forget to do a particular thing.

## Example: $y$-axis

A particular person forgot to do a particular thing.

## Results



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## Predicting distribution using veridicality

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Regularized censored factor analysis with loss weighted by normalized acceptability and scores constrained to ( $-1,1$ ).

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(Ask about specifics after the talk.)

## Inference patterns



Inference polarity
Matrix polarity $\square$ negative $\square$ positive

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Pattern 3

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## Inference patterns: factivity/veridicality



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## Inference patterns: implicatives



Pattern 1

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$\begin{array}{rrrrr}-1.0 & -0.5 & 0.0 & 0.5 & 1.0\end{array}$

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Learn optimal mapping from veridicality inference patterns to syntactic distribution using cross-validated ridge regression.

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Approach
Learn optimal mapping from veridicality inference patterns to syntactic distribution using cross-validated ridge regression.

Finding
Across all frames in MegaAcceptability, this mapping explains about $20 \%$ of the variance in the acceptability judgments.

## Predicting distribution from inference



Syntactic structure

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2. Not nearly enough information to base a generalization on.

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## Motivation

Relationship may be mediated by non-contentful properties of contentful events Kratzer 2006; Hacquard 2006; Moulton 2009; Anand and Hacquard 2013, 2014; Rawlins 2013; Bogal-Allbritten 2016; White and Rawlins 2016b a.o.

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## Approach

Use Uniform Manifold Approximation and Projection (UMAP) to visualize the topological structure of the distribution and veridicality data. McInnes and Healy 2018

## Exploratory analysis



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1. Dataset capturing the variability of factivity and veridicality across finite and infinitival complement types.
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## Conclusion

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Analytical contributions

1. Inference pattern typology explains some parts of syntactic distribution reasonably well, but far from perfect.
2. More likely that the veridicality-distribution relationship is indirect, mediated by fine-grained verb class.

## Future directions

Big remaining question
How are inference patterns represented in the lexicon?

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## Possibility 1

Verb class-specific rules (possibly sensitive to content-dependent properties, like veridicality and factivity).

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Possibility 2
More abstract semantic properties relevant to thematic roles - e.g. affectedness, existence, creation/destruction, ...

Thanks!

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Data available at megaattitude.io

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