

Misunderstanding the Concrete, Disagreeing about the Abstract

A Closer Look at Word Meaning Negotiation Triggers

Bill Noble, Staffan Larsson, Jenny Myrendal

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What is WMN?

Research questions

Data

Descriptive Statistics

Experiments

What is WMN?

Word Meaning Negotiation (WMN)

WMN is a conversational sequence in which two or more speakers explicitly discuss the meaning of a word or phrase (Myrendal, 2015).

DIN originating in **disagreement** over the meaning of the trigger term

NON originating in **non-understanding** the meaning of the trigger term

WMN Example 1

Non-understanding (NON)

Myrendal (2015) – familjeliv.se forum, translated from Swedish

S1: I'm going to the doctor to get a **full body scan** tomorrow.

S2: What do you mean by **full body scan**?

S1: I mean a kind of X-ray where they can see all of the inflamed parts

S2: I've never had that done, was curious as to what that was I've had my hands examined by ultrasound.

WMN Example 1

Non-understanding (NON)

Myrendal (2015) – familjeliv.se forum, translated from Swedish

S1: I'm going to the doctor to get a **full body scan** tomorrow.

S2: What do you mean by **full body scan**?

S1: I mean a kind of X-ray where they can see all of the inflamed parts

S2: I've never had that done, was curious as to what that was I've had my hands examined by ultrasound.

- ▶ **trigger** – *full body scan* (S1)
- ▶ **indicator** – explicit clarification request (S2)
- ▶ **response** – explication of the in-context trigger (S2)

WMN Example 2

Disagreement (DIN)

Norén and Linell (2007) – translated from Swedish

S1: Telling children about Santa Claus is straight up **lying** to them.

S2: That's not what **lying** means at all!

S1: Of course it is, **lying** means not telling the truth and everyone knows Santa doesn't exist.

WMN Example 2

Disagreement (DIN)

Norén and Linell (2007) – translated from Swedish

S1: Telling children about Santa Claus is straight up **lying** to them.

S2: That's not what **lying** means at all!

S1: Of course it is, **lying** means not telling the truth and everyone knows Santa doesn't exist.

- ▶ **trigger** – straight up *lying* (S1)
- ▶ **indicator** – challenge usage in given context (S2)
- ▶ **response** – elaborates their understanding, connecting to situation (S2)

Situated meaning vs. Meaning potential¹

Situated meaning – the **in context** meaning of a lexical item

- ▶ *disambiguated* and *enriched* by interactive context
- ▶ related to **speaker meaning**
- ▶ the *type of meaning* used to compute Montague-style compositional utterance meanings

¹Norén and Linell (2007)

Situated meaning vs. Meaning potential¹

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Meaning potential – the **abstract** semantic context of the lexical item

- ▶ grounded in a particular **speech community**
- ▶ a function that produces situated meanings, given context

¹Norén and Linell (2007)

WMN Examples 3

situated meaning

noun: *bell* | J9P/J9P_760

A: Oh no not an alarm it's it's be too expensive, no just an internal **bell** to frighten the hell out of them.

[...]

B: So what do you mean by a **bell**, [UNCLEAR] trying to visualize what you mean

A: Yes. Well like [UNCLEAR] I mean we all know what a **bell** is, a **bell** which is set off by— by a human body coming in.

meaning potential

noun: *invisible fencing* | 4179-0/4179-0_4179-74/

A: I see other people out there and they hit their dogs and try to— and those horrible collars that they put on them with invinc— **invisible fencing**, least I—

B: Invisible what?

A: **Invisible fencing**, have you heard of that?

B: No, what is that?

A: It's— uh, it's a system you can put in your yard where you bury these little uh, transducers or emitters in your yard—

Research questions

The background features a white central area with teal-colored geometric shapes. Two large teal triangles point towards each other from the left and right sides, meeting at a point at the bottom. A smaller, darker teal triangle is positioned at the bottom center, overlapping the meeting point of the two larger triangles.

Research questions

RQ1

What trigger word features are predictive of the **type of WMN** (NON or DIN)?

RQ2

What trigger word features are predictive of the **kind of meaning** (situated meaning or meaning potential) that is the focus of a WMN?

Research questions

RQ1

What trigger word features are predictive of the **type of WMN** (NON or DIN)?

RQ2

What trigger word features are predictive of the **kind of meaning** (situated meaning or meaning potential) that is the focus of a WMN?

What features?

Research questions

RQ1

What trigger word features are predictive of the **type of WMN** (NON or DIN)?

RQ2

What trigger word features are predictive of the **kind of meaning** (situated meaning or meaning potential) that is the focus of a WMN?

What features?

- ▶ Part of speech
- ▶ Concreteness/abstractness
- ▶ Lexical sentiment

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Data

The NeWMe² corpus

- ▶ **Negotiation of Word Meaning**
- ▶ The first large-scale annotated corpus of WMNs
- ▶ Spoken and written interaction: BNC, Switchboard, r/ChangeMyView

Interaction Type	#	%
Online (r/ChangeMyView)	216	55%
Spoken (total)	176	45%
Spoken (BNC)	141	36%
Spoken (SW)	35	9%
Total	392	100%

²<https://dev.clasp.gu.se/newme/> (Garí Soler et al., 2025)

Lexical concreteness ratings

Concreteness list (Brysbaert et al., 2014)

- ▶ Crowd-sourced ratings for 40 000 English words
- ▶ 5-point likert scale
 - ▶ ≥ 25 judgments for most words
- ▶ For multi-word WMN triggers not in the concreteness dataset, we use the statistics for the syntactic headword

Concreteness features

conc_μ mean of judgements for a given word

conc_σ standard deviation of judgements for a given word

Sentiment metrics

SentiWordNet v. 3.0 (Baccianella et al., 2010)

- ▶ Includes three metrics: **positivity**, **negativity**, and *objectivity*
- ▶ Constructed such that all three sum to 1
- ▶ We use the first 2 as features:

Sentiment features

$sent^+$ SentiWordNet positivity score

$sent^-$ SentiWordNet negativity score

Other lexical features

Annotated in NewMe / by one of the authors

Other lexical features

`pos` part of speech – *noun, adjective, verb, adverb, or acronym*

`lexform` lexical form – *single word, or multi-word expression*

Descriptive Statistics

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Concreteness of trigger expression

Type	NeWMe		Brysaert	
	#	%	#	%
Abstract	130	33%	15,447	39%
Mixed	116	30%	10,913	27%
Concrete	120	31%	13,594	34%
Unknown	26	7%	0	0%
Total	392	100%	39,954	100%

Abstract $\text{conc}_\mu \leq 2.5$

Concrete $\text{conc}_\mu \geq 3.5$

Mixed $2.5 < \text{conc}_\mu < 3.5$

Unknown Word or phrase not found in the dataset

Sentiment of trigger expression

Sentiment	#	%
Positive	50	13%
Negative	39	10%
Both	35	9%
Neither / Not Included	268	68%
Total	392	100%

Positive $\text{sent}^+ > 0$ and $\text{sent}^- = 0$

Negative $\text{sent}^+ = 0$ and $\text{sent}^- > 0$

Both $\text{sent}^+ > 0$ and $\text{sent}^- > 0$

Neither $\text{sent}^+ = 0$ and $\text{sent}^- = 0$

Other variables

Lexical Form	#	%
Single	266	68%
Compound	121	31%
Acronym	5	1%
Total	392	100%

POS	#	%
Noun	281	72%
Adjective	55	14%
Verb	50	13%
Adverb	6	2%
Total	392	100%

Experiments

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Research questions

Remember these?

RQ1

What trigger word features are predictive of the **type of WMN** (NON or DIN)?

RQ2

What trigger word features are predictive of the **kind of meaning** (situated meaning or meaning potential) that is the focus of a WMN?

Summary of features

Features of WMNs

corpus *BNC, SwitchBoard, ChangeMyView*

type *non-understanding, disagreement, other*

situated *did the WMN discuss situated meaning*

potential *did the WMN discuss meaning potential*

Features of trigger words

pos *part of speech – noun, adjective, verb, adverb, or acronym*

lexform *single word, multi-word expression*

conc_μ / conc_σ *mean/standard deviation of concreteness judgements*

sent⁺ / sent⁻ *SentiWordNet positivity/negativity score*

Model 1

Predicting type of WMN (NON or DIN)

$$\begin{aligned} \text{type} \sim & 1 + \text{conc}_{\mu} + \text{conc}_{\sigma} \\ & + \text{sent}^{+} + \text{sent}^{-} + (\text{sent}^{+} * \text{sent}^{-}) \\ & + \text{pos} + \text{lexform} + (1|\text{corpus}) \end{aligned}$$

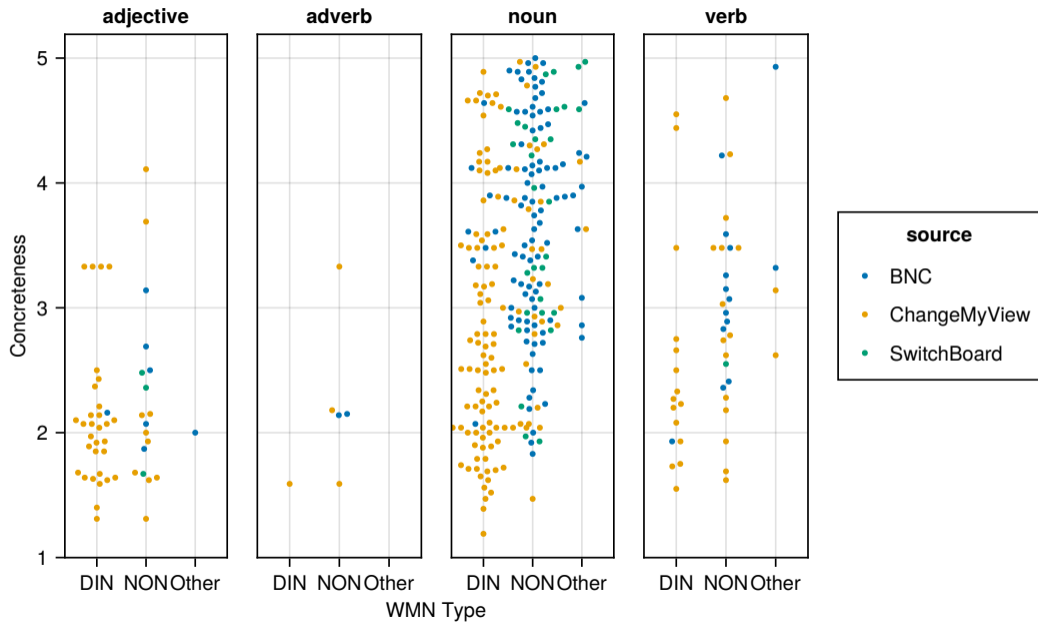
- ▶ `type` is a Bernoulli response variable (1 for *DIN* and 0 for *NON*)
 - ▶ We excluded `type = Other`
- ▶ A random intercept is included for `corpus`
- ▶ `sent+ * sent-` captures the sentiment magnitude

Model 1

Predicting type of WMN (NON or DIN)

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- ▶ DINs are more likely to be more abstract words; NONs about more concrete words
 - ▶ $\beta = -0.597; p = 0.0012$ (conc_{μ})
- ▶ Verbs are more likely to appear in NONs
 - ▶ $\beta = -1.306; p = 0.0030$ ($\text{pos} = \text{verb}$)
- ▶ Multi-word expressions are more associated with NONs
 - ▶ $\beta = -0.785; p = 0.0226$ ($\text{lexform} = \text{multi-word}$)



Model 2

Predicting *situated meaning* aspect

$$\begin{aligned} \text{situated} \sim & 1 + \text{type} + \text{conc}_{\mu} + \text{conc}_{\sigma} \\ & + \text{sent}^{+} + \text{sent}^{-} + (\text{sent}^{+} * \text{sent}^{-}) \\ & + \text{pos} + \text{lexform} + (1|\text{corpus}) \end{aligned}$$

- ▶ `situated` is a Bernoulli response variable
 - ▶ Recall that `situated` and `potential` are not mutually exclusive
- ▶ `type` is included as a predictor
- ▶ `sent+ * sent-` captures the sentiment magnitude

Model 2

Predicting *situated meaning* aspect

$$\begin{aligned} \text{situated} \sim & 1 + \text{type} + \text{conc}_{\mu} + \text{conc}_{\sigma} \\ & + \text{sent}^{+} + \text{sent}^{-} + (\text{sent}^{+} * \text{sent}^{-}) \\ & + \text{pos} + \text{lexform} + (1|\text{corpus}) \end{aligned}$$

- ▶ Adjective and verb triggers are *more likely* in situated meaning WMNs (vs. nouns)
 - ▶ $\beta = 1.305; p = 0.0111$ ($\text{pos} = \textit{adjective}$)
 - ▶ $\beta = 2.121; p = 0.0050$ ($\text{pos} = \textit{verb}$)
- ▶ Not quite significant results for `type`
 - ▶ $\beta = -0.688; p = 0.0684$ ($\text{type} = \textit{DIN}$)

Model 3

Predicting *meaning potential* aspect

$$\begin{aligned} \text{potential} \sim & 1 + \text{type} + \text{conc}_{\mu} + \text{conc}_{\sigma} \\ & + \text{sent}^{+} + \text{sent}^{-} + (\text{sent}^{+} * \text{sent}^{-}) \\ & + \text{pos} + \text{lexform} + (1|\text{corpus}) \end{aligned}$$

- ▶ Exact same model but with **potential** as the Bernoulli response variable

Model 3

Predicting *meaning potential* aspect

$$\begin{aligned} \text{potential} \sim & 1 + \text{type} + \text{conc}_\mu + \text{conc}_\sigma \\ & + \text{sent}^+ + \text{sent}^- + (\text{sent}^+ * \text{sent}^-) \\ & + \text{pos} + \text{lexform} + (1|\text{corpus}) \end{aligned}$$

- ▶ Adjective and verb triggers are *less likely* in meaning potential WMNs (vs. nouns)
 - ▶ $\beta = -1.562; p = 0.0016$ ($\text{pos} = \text{adjective}$)
 - ▶ $\beta = -1.386; p = 0.0020$ ($\text{pos} = \text{verb}$)
- ▶ DINs are significantly more likely to address meaning potential
 - ▶ $\beta = 2.133; p < 1e-9$ ($\text{type} = \text{DIN}$)

Summary

- ▶ There are features of the **trigger word** that predict **type of WMN (RQ1)**.
 - ▶ More abstract → disagreement
 - ▶ Verbs → non-understanding
 - ▶ Multi-word expressions → non-understanding

Summary

- ▶ There are features of the **trigger word** that predict **type of WMN (RQ1)**.
 - ▶ More abstract → disagreement
 - ▶ Verbs → non-understanding
 - ▶ Multi-word expressions → non-understanding
- ▶ There are features of the **trigger word** that predict what **aspect of meaning** is negotiated (**RQ2**).
 - ▶ Adjectives / verbs → situated meaning
 - ▶ Nouns → meaning potential
 - ▶ **Additionally:** disagreement → meaning potential
- ▶ No significant results for sentiment in this preliminary study

Discussion

and future work

- ▶ Abstract words may leave more room for *individual variation* in interpretation (thus more disagreement)
- ▶ The part of speech results point to the *flexibility* of adjectives and verbs in comparison to nouns
- ▶ Disagreements are more likely to address meaning potential, pointing to the *normative stakes* of certain WMNs

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- ▶ We only looked at WMNs here — how do the triggers compare to words that *don't* trigger WMNs?
- ▶ The *type of interaction* (task-orientation, relationship of interlocutors, etc.) may factor into what kinds of words get negotiated

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Thank you!

References

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- Stefano Baccianella, Andrea Esuli, and Fabrizio Sebastiani. 2010. SentiWordNet 3.0: An enhanced lexical resource for sentiment analysis and opinion mining. In *Proceedings of the Seventh International Conference on Language Resources and Evaluation (LREC'10)*, Valletta, Malta. European Language Resources Association (ELRA).
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situated meaning

noun: *bell* | J9P/J9P_760

A: Oh no not an alarm it's it's be too expensive, no just an internal **bell** to frighten the hell out of them.

[...]

B: So what do you mean by a **bell**, [UNCLEAR] trying to visualize what you mean

A: Yes. Well like [UNCLEAR] I mean we all know what a **bell** is, a **bell** which is set off by— by a human body coming in.

adjective: *recent* | FME/FME_18

A: Is that is that **recent** or is that the old stuff the Venn diagrams?

B: What do you mean by **recent**?

A: Have you done it in the last sort of few weeks?

B: Oh yeah it's the last few weeks.

meaning potential

noun: *invisible fencing* | 4179-0/4179-0_4179-74/

A: I see other people out there and they hit their dogs and try to— and those horrible collars that they put on them with invinc— **invisible fencing**, least I—

B: Invisible what?

A: **Invisible fencing**, have you heard of that?

B: No, what is that?

A: It's— uh, it's a system you can put in your yard where you bury these little uh, transducers or emitters in your yard—

adjective: *anthropogenic* | F8E/F8E_21

A: Okay ? So we know that so far about fifty percent of our **anthropogenic** C O two has been locked away

B: What does **anthropogenic** mean?

A: From human sources . For example can we continue burning fossil fuel [...]

Model 1 results: type = *DIN*

	Coef.	Std. Error	z	Pr(> z)
(Intercept)	0.353722	1.93684	0.18	0.8551
conc _{μ}	-0.596865	0.184348	-3.24	0.0012
conc _{σ}	-0.204307	0.564173	-0.36	0.7173
sent ⁺	0.664035	1.15001	0.58	0.5637
sent ⁻	0.199255	1.1104	0.18	0.8576
pos: <i>adjective</i>	-0.727312	0.507808	-1.43	0.1521
pos: <i>verb</i>	-1.306430	0.439714	-2.97	0.0030
lexform: <i>multi-word</i>	-0.784656	0.344007	-2.28	0.0226
sent ⁺ * sent ⁻	-5.356570	5.5816	-0.96	0.3372

Model 2 results: situated = 1

	Coef.	Std. Error	z	Pr(> z)
(Intercept)	1.58082	1.02038	1.55	0.1213
type: DIN	-0.687528	0.377333	-1.82	0.0684
conc _{μ}	-0.00794446	0.16794	-0.05	0.9623
conc _{σ}	-0.332505	0.496781	-0.67	0.5033
sent ⁺	-1.52334	1.00917	-1.51	0.1312
sent ⁻	-0.345484	0.936351	-0.37	0.7122
pos: <i>adjective</i>	1.30509	0.513597	2.54	0.0111
pos: <i>verb</i>	2.12148	0.755799	2.81	0.0050
lexform: <i>multi-word</i>	0.360319	0.322338	1.12	0.2636
sent ⁺ * sent ⁻	5.32303	5.34836	1.00	0.3196

Model 3 results: potential = 1

	Coef.	Std. Error	z	Pr(> z)
(Intercept)	-0.0314509	1.04118	-0.03	0.9759
type: DIN	2.13327	0.342115	6.24	< 1e-9
conc _{μ}	-0.24192	0.172902	-1.40	0.1618
conc _{σ}	0.217801	0.497279	0.44	0.6614
sent ⁺	1.62563	0.990182	1.64	0.1006
sent ⁻	0.0794144	0.975404	0.08	0.9351
pos: <i>adjective</i>	-1.56217	0.49569	-3.15	0.0016
pos: <i>verb</i>	-1.38621	0.447837	-3.10	0.0020
lexform: <i>multi-word</i>	-0.550717	0.304911	-1.81	0.0709
sent ⁺ * sent ⁻	1.48935	5.80966	0.26	0.7977