

# MODELING INTENSIFICATION IN NEW ZEALAND ENGLISH DATA

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## Habilitation project

### *Acquisition, Variation, and Diachronic Development of Intensification in English*

- ▶ synchronic quantitative corpus-based study
- ▶ adjectival intensification in New Zealand English (NZE)
- ▶ based on the New Zealand component of the *International Corpus of English (ICE-NZ)*

Setting the Stage

Data and Methodology

Results

Summary

Problems & Outlook

References

# Intensification

- (1) yeah... just it would make it **so** awkward eh you know (ICE-NZ S1A-001:1\$M)
- (2) um... sara's got a **really** nice sleeveless green... you know coat jacket (ICE-NZ S1A-002:1\$Q)
- (3) she was a **very** nervous sort of a woman (ICE-NZ S1A-018:1\$A)

# Intensification

Intensification is related to the semantic category of *degree* (degree adverbs) and ranges between very low intensity (downtoning) and very high (amplifiers) (Quirk et al. 1985: 589–590).

- ▶ Amplifiers (Tagliamonte 2008)
  - ▶ Maximizers (e.g. *completely*)
  - ▶ Boosters (e.g. *very much*)
- ▶ Downloners
  - ▶ Approximators (e.g. *almost*)
  - ▶ Compromisers (e.g. *more or less*)
  - ▶ Diminishers (e.g. *partly*)
  - ▶ Minimizers (e.g. *hardly*)

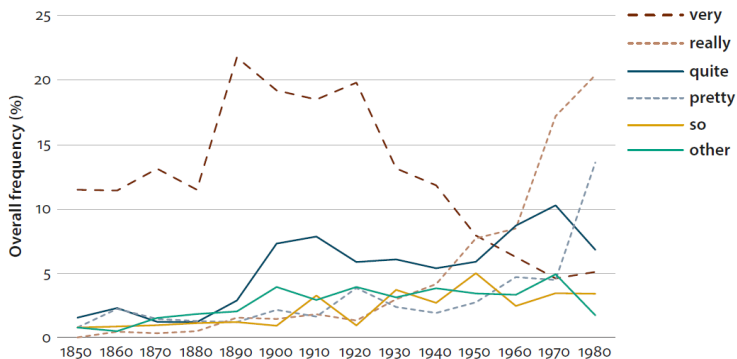
# Previous Research

- ▶ Intensification...
  - ▶ major area of grammatical change  
(cf. Brinton and Arnovick 2006: 441)
  - ▶ crucial for the “social and emotional expression of speakers” (Ito and Tagliamonte 2003: 258)
  - ▶ teenage talk and young(ish) speakers  
(Bauer and Bauer 2002; Macaulay 2006)
  - ▶ female speakers (Tagliamonte 2006, 2008; D’Arcy 2015)

# Previous Research

- ▶ Ongoing changes are accompanied by ...
  - ▶ gender and age differences (apparent time construct)
  - ▶ differences in the syntactic function (predicative vs attributive)
  - ▶ the semantic type of the modified adjective
  - ▶ emotional value of the modified adjective (emotional vs non-emotional)
  
- ▶ Intensifying *really* replaces *very* (lexical replacement)  
(cf. D'Arcy 2015; Ito and Tagliamonte 2003; Tagliamonte 2005, 2008)

## Previous study of intensification in New Zealand English (D'Arcy 2015)



(D'Arcy 2015: 468)



# Research Question

Q<sub>1</sub>

Is the NZE Intensifier system currently undergoing change?

# ICE NEW ZEALAND

# ICE New Zealand

New Zealand component of the *International Corpus of English* (Bauer et al. 1999)

- ▶ released in 1999 (*The Victoria University of Wellington*)
- ▶ consists of one million words (600,000 spoken and 400,000 written)
- ▶ representing diverse spoken and written text types
- ▶ here only private dialogues (200,000 words)

# DATA PROCESSING

# Data Processing

- ▶ Split spoken data into utterances
- ▶ Removal of meta information
- ▶ Part-of-speech tagging
- ▶ Retrieving adjectives (PoS-tag JJ)
- ▶ Determining whether adjective is preceded by an intensifying adverb (PoS-tag RB)

# Data Processing

- ▶ Determining the syntactic type of adjective (predicative vs attributive (if followed by NN\* tag))
- ▶ Removal of
  - ▶ negated adjectives
  - ▶ comparative and superlative forms
  - ▶ non-intensifiable forms  
(categorical, e.g. nationalities | locations: *asian*, *Asia*)
- ▶ Sentiment Analysis  
determines the emotional value of adjectives based on the *Word-Emotion Association Lexicon* (Mohammad and Turney 2013)
- ▶ Manual cross-evaluation of automated classification
- ▶ Adding speaker information (age, sex, etc.).

## DATA SUMMARY

# Data Summary: ICE-NZ data

Age	Sex	Speakers (N)	Adj. (N)	Int. (N)	Int. (%)
16-24	female	39	1102	140	12.7
16-24	male	29	811	81	10.0
25-39	female	23	629	65	10.3
25-39	male	16	481	35	7.3
40-49	female	16	509	60	11.8
40-49	male	9	172	7	4.1
50+	female	7	259	27	10.4
50+	male	6	236	25	10.6
<b>Total</b>		145	4199	440	10.5

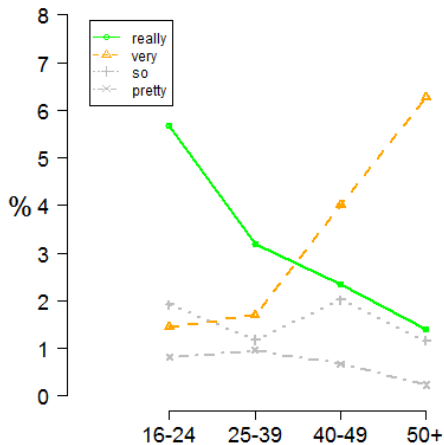


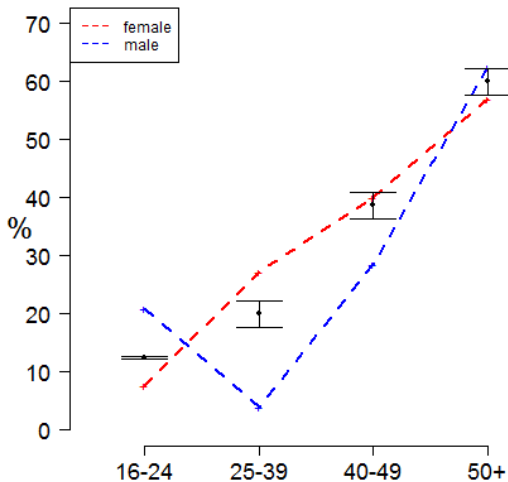
# Data Summary: Intensifiers ICE-NZ

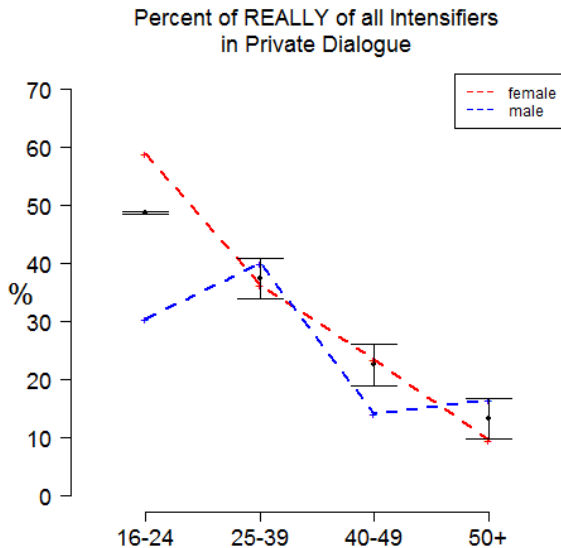
<b>Intensifier</b>	<b>N</b>	<b>%</b>	<b>Int. (%)</b>
∅ Intensification	3759	89.52	
really	150	3.57	34.09
very	96	2.29	21.82
so	66	1.57	15.00
too	34	0.81	7.73
pretty	29	0.69	6.59
real	18	0.43	4.09
well	7	0.17	1.59
absolutely, right, totally	5	0.36	3.42
bloody	4	0.10	0.91
crazy, particularly	2	0.10	0.90
actually, badly, completely, definitely, dread- fully, enormously, entirely, excruciatingly, fuck- ing, fully, horrendously, incredibly, obviously, purely, shocking, true, wicked	1	0.34	3.91
<b>Total</b>	<b>4199</b>	<b>10.48</b>	<b>100</b>

RESULTS  
—  
VISUALIZATION

# Intensifiers across Age Cohorts



Percent of VERY of all Intensifiers  
in Private Dialogue



# STATISTICAL ANALYSIS

# Research question

Q<sub>2</sub>

Which factors correlate with the use of *really* (innovation)?  
(age, sex, syntactic function, ...)

# Statistical Analysis

- ▶ Mixed-effects binomial logistic regression models
  - ▶ intensified adjectives only, step-wise step-up model fitting

## Dependent Variable(s)

<b>really</b>	nominal	occurrence of <i>really</i> vs other intensifier
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## Independent Variable(s)

<b>age</b>	categorical	age groups in ascending order		
<b>sex</b>	nominal	male   female	extra	linguistic
<b>eth</b>	nominal	pakeha   maori		
<b>occ</b>	nominal	acmp   sml		
<b>emo</b>	nominal	emotional   nonemotional	intra	linguistic
<b>fun</b>	nominal	attributive   predicative		
<b>sem</b>	categorical	semantic type of adjective		
<b>grad</b>	nominal	gradable   nongradable		



## REGRESSION RESULTS

# Regression Results

	Group(s)	Variance	SDev.	L.R. $\chi^2$	Sig.
<b>Random Effect(s)</b>	flid	0.83	0.91	24.01 (DF 1)	<.001***
<b>Fixed Effect(s)</b>	Estimate	VIF	OddsRatio	z value	Sig.
(Intercept)	-0.69		0.5	-1.62	n.s.
age:25-39	-0.59	1.12	0.55	-1.49	n.s.
age:40-49	-1.46	1.14	0.23	-3.04	<.01 **
age:50+	-2.18	1.08	0.11	-3.55	<.001***
grad:nograd	1.06	1.01	2.88	2.7	<.01 **
sex::male	-0.84	1.04	0.43	-2.4	<.05 *
<b>Model statistics</b>					<b>Value</b>
Number of Groups					115
Number of cases in model					388
Observed misses					240
Observed successes					148
R2 (Nagelkerke)					0.408
C					0.842
Somers' Dxy					0.684
Prediction accuracy					79.12%
<b>Model Likelihood Ratio Test</b>			L.R. $\chi^2$	59.43 (DF 6)	<.001***

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## Summary

- ▶ The NZE intensifier system is undergoing change
- ▶ The observed change is accompanied by extra-linguistic stratification and pragmatic constraints

But why is *really* taking over???

# Collocates

Q<sub>3</sub>

Why and how is *really* replacing *very*?  
Why not another variant?

H<sub>1</sub>

Successful variants (*really*) associate with many and particularly high frequency adjectives.



Adjective	Intensifier	Age			
		16-24	25-39	40-49	50+
difficult	really	0	2	0	0
good	really	27	9	2	3
hard	really	5	3	0	1
important	really	0	1	1	2
interesting	really	1	2	1	2
little	really	0	0	0	0
nice	really	7	1	1	1
strong	really	0	0	0	0
difficult	very	0	4	5	12
good	very	5	15	20	23
hard	very	0	3	4	4
important	very	2	6	4	8
interesting	very	1	0	2	8
little	very	0	5	4	11
nice	very	3	2	5	3
strong	very	0	2	4	8
difficult	other	2	3	3	2
good	other	8	6	6	6
hard	other	5	2	2	0
important	other	1	0	2	2
interesting	other	1	1	0	1
little	other	0	0	1	1
nice	other	0	0	0	0
strong	other	0	3	0	0

 increasing trend  
 receding trend

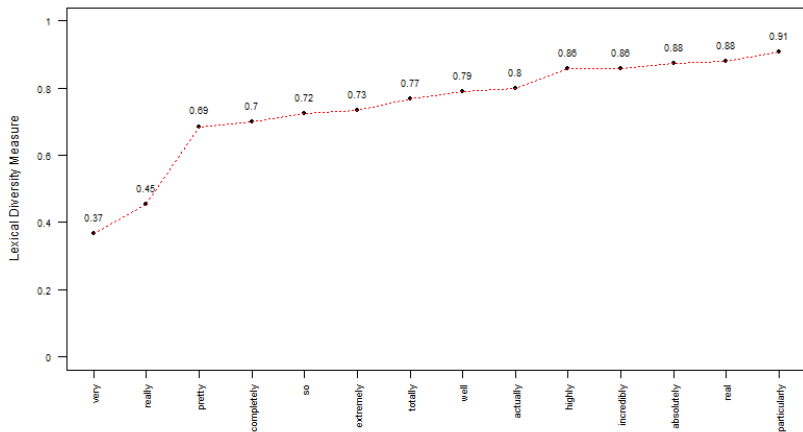


Figure: adjective types/intensifier tokens (ICE NZ)

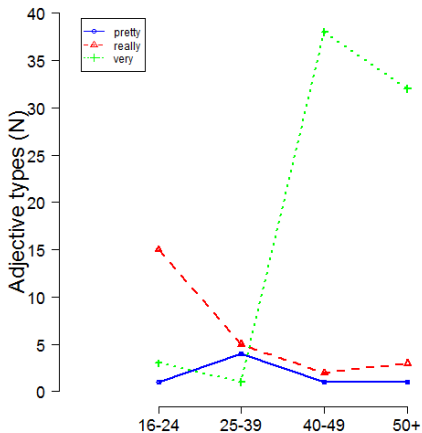


Figure: intensifier types : adjective types : age (ICE NZ)

## Covarying Collexeme Analysis (CCA)

- ▶ Member of the *collostructional analysis* family that aims at measuring the degree of attraction of lemmas in one slot of a construction to lemmas in another slot of the same construction. (Stefanowitsch and Gries 2005)
- ▶ CCA differs from most collocation statistics such that it takes syntactic structure into account and it applies Fisher-Yates exact test and is thus not based on any distributional assumptions.

### Collocations by Age

Age	Adjective	Intensifier	OddsRatio	Bonf. Corr.	Sig
16-24	really	good	5.44		p<.001
50+	very	difficult	20.07		p<.001
50+	very	good	4.72		p<.001
50+	very	strong	21.33		p<.01

## Correspondence Analysis (CA)

- ▶ Exploratory method for correspondence between rows and column values/counts in distributional data (similar to factor analysis) (Baayen 2008: 128–136)
- ▶ Restrictions on intensifiers  
Minimum of 7 intensifier tokens and intensifier type has to collocate with more than 1 adjective type
- ▶ Restrictions on adjectives  
Minimum of 7 adjective tokens

	pretty	real	really	so	very
bad	1	2	3	1	1
clear	0	1	0	0	7
close	0	0	2	2	3
different	0	0	1	2	4
difficult	2	0	2	2	21
first	0	0	0	0	9

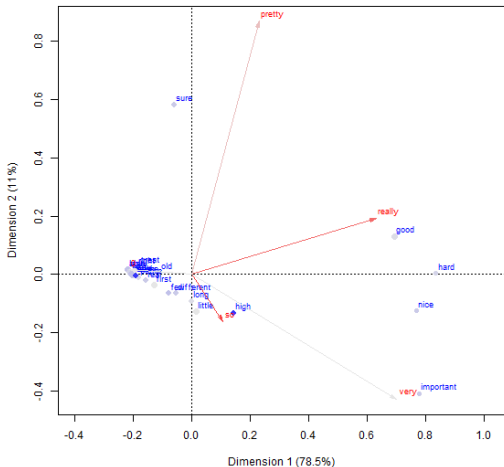


Figure: Correspondence of intensifier and adjective tokens (NZE)

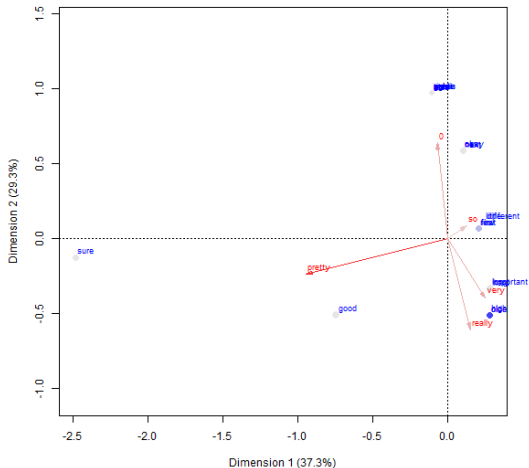


Figure: Correspondence of intensifier and adjective types (NZE)

# SUMMARY



## Intensifying *really*

- ▶ use declines almost linearly with age (incoming innovation)
- ▶ is dis-preferred by male speakers (female dominated change)
- ▶ is preferred by non-gradual adjectives

*Really* is socially stratified and correlates with extra-linguistic factors (age, sex) and is subject to pragmatic constraints (gradability).

## Successful intensifiers (*really*)

- ▶ increase the number of adjectives that they co-occur with
- ▶ collocate significantly with high frequency adjectives (*good* among 16-24 year olds)
- ▶ semantically similar to dominant variant (have similar collocation profiles)

Successful intensifier variants collocate with many adjectives (semantically bleached), particularly high frequency adjectives, and show collocation profiles similar to those of (previously) dominant variants.

## PROBLEMS & OUTLOOK

## Problems

- ▶ apply multinominal regression instead of binomial (dependent variable is categorical, really, very, other, null)
- ▶ use Wellington corpus instead of ICE (1,000,000 instead of 200,000 words) → increase of data base
- ▶ apply SVSM to adjectives to find distinct types of adjectives
- ▶ define variable context and variants better (not all adjectives take all intensifiers)!
- ▶ clean data: find true variants (? *completely* good : *very* good)

THANK YOU SO, REALLY, VERY MUCH!

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# MODELING INTENSIFICATION IN NEW ZEALAND ENGLISH DATA

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## APPENDIX

# Regression Results

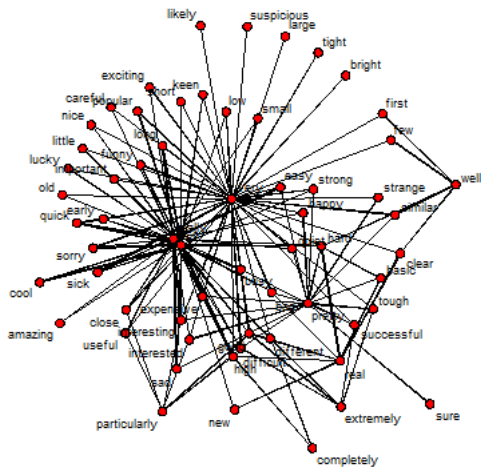
	Group(s)	Variance	Std. Dev.	L.R. $\chi^2$ (df1)	Sig.
<b>Random Effect(s)</b>	flid	0.44	0.66	29	p<.001***
<b>Fixed Effect(s)</b>	Estimate	VIF	OddsRatio	z value	Sig.
(Intercept)	-5.04		0.01	-14.55	p<.001***
age:25-39	-0.57	1.07	0.57	-2.09	p<.05*
age:40-49	-0.94	1.08	0.39	-2.7	p<.01**
age:50+	-1.48	1.03	0.23	-2.98	p<.01**
sex:male	-0.85	1.01	0.43	-3.46	p<.001***
fun:predicative	0.74	1	2.09	4.09	p<.001***
grad:nograd	1.88	1.01	6.52	6.31	p<.001***
emo:emotional	0.79	1.01	2.21	4.49	p<.001***
<b>Model statistics</b>					Value
Number of Groups					145
Cases in model					4199
Observed successes					150
R <sup>2</sup> (Nagelkerke)					0.155
C					0.844
Somers' D <sub>xy</sub>					0.688
Prediction accuracy					96.43%
<b>Model LL Ratio Test</b>			L.R. $\chi^2$ (df8)	176.67	p<.001***

## Distributional Analysis)

- ▶ Network Analysis
- ▶ Words that share collocates are semantically similar (Stefanowitsch 2010: 368–370).
- ▶ Semantic similarity can thus be measured in collocate frequency.

	extremely	particularly	pretty	real	really	so	very
amazing	0	0	0	0	4	1	0
bad	0	0	1	2	3	1	1
basic	0	0	1	1	0	0	3
bright	0	0	0	0	0	0	4
busy	0	0	1	0	2	2	1
careful	0	0	0	0	1	0	6





# Semantic Vector Space Models (SVSM)

- ▶ Distributional approach to semantics (cf. Levshina 2015: 323–331).
- ▶ Semantic similarity is measured as similarity of vectors of collocate frequency.
- ▶ The similarity measure is a cosine measure (the more similar the cosine value the greater the similarity in the vectors and thus the distributional collocate profile of two intensifiers).

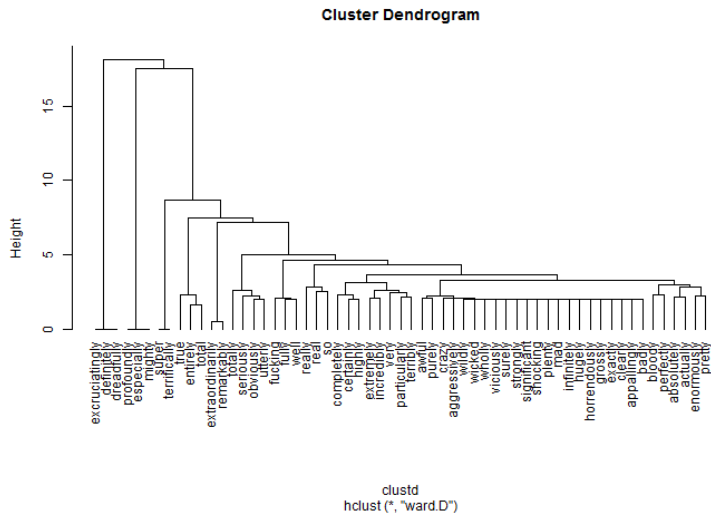


Figure: Clustering of intensifiers by adjective collocates (NZE)