Statistics Seminar, Spring 2009

Binomial (or Binary) Logistic Regression

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Linear regression: Univariate

One independent variable, one (continuous) dependent variable.

 $Outcome_i = Model_i + Error_i$ $Y_i = b_0 + b_1 X_1 + \mathcal{E}i$

- b_0 : interception at y-axis
- b_1 : line gradient
- X_1 : predictor variable
- *E*: Error
- X_1 predicts Y.

Linear regression: Multivariate

Several independent variables, one (continuous) dependent variable.

$$Y_i = b_0 + b_1 X_1 + b_2 X_2 + \dots + b_n X_n + \epsilon_i$$

 b_0 : interception at y-axis

 b_1 : line gradient

- b_n : regression coefficient of X_n
- X_1 : predictor variable
- $\boldsymbol{\varepsilon}$: Error

 X_1 predicts Y.

Assumption

- Linear regression assumes linear relationships between variables.
- This assumption is usually violated when the dependent variable is categorical.
- The logistic regression equation expresses the multiple linear regression equation in logarithmic terms and thereby overcomes the problem of violating the linearity assumption.

Assumption cont.

$$log_{base}[number] = 0 = 4 \qquad \implies \qquad 2^4 = 2 \times 2 \times 2 \times 2 = 16$$

$$ln = log_e[number] \qquad \qquad | e = Eulers constant \approx 2,7182818284...$$

$$ln[odds] \Rightarrow 'logit' \qquad \qquad logit(p) = ln \frac{p}{(1-p)}$$

$$e^{\operatorname{logit}(p)} = \frac{p}{1-p}$$

$$e^{\operatorname{logit}(p)}(1-p) = p = e^{\operatorname{logit}(p)} - pe^{\operatorname{logit}(p)}$$

$$p + pe^{\operatorname{logit}(p)} = e^{\operatorname{logit}(p)}$$

$$p(1+e^{\operatorname{logit}(p)}) = e^{\operatorname{logit}(p)}$$

$$p = \frac{1}{1+e^{-\operatorname{logit}(p)}}$$

Binary logistic regression: Univariate

One independent variable, one *categorical* dependent variable.

$$P(Y) = \frac{1}{1 + e^{-(b_0 + b_1 x_1)}}$$

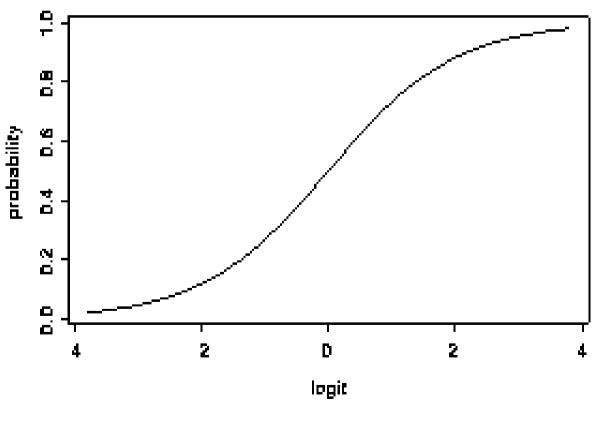
P: probability of *Y* occuring

- *e*: natural logarithm base (= 2,7182818284...)
- b_0 : interception at y-axis
- b_1 : line gradient

 X_1 predicts the probability of *Y*.

Binary logistic regression: Univariate cont.

As P(Y) ranges from 0 to 1, the logit ranges from $-\infty$ to $+\infty$.



http://data.princeton.edu/wws509/notes/c3s1.html

Binary logistic regression: Multivariate

Several independent variables, one *categorical* dependent variable.

$$P(Y) = \frac{e^{b_0 + b_1 x_1 + b_2 x_2 + \dots + b_n x_n}}{1 + e^{b_0 + b_1 x_1 + b_2 x_2 + \dots + b_n x_n}}$$

- *P*: probability of *Y* occuring
- *e*: natural logarithm base
- b_0 : interception at y-axis
- b_1 : line gradient
- b_n : regression coefficient of X_n
- X_1 : predictor variable

 X_1 predicts the probability of *Y*.

Binary logistic regression: Multivariate cont.

=> Linear regression predicts the *value* that Y takes.

Instead, in logistic regression, the *frequencies* of values 0 and 1 are used to predict a value:

=> Logistic regression predicts the *probability* of Y taking a specific value.

Research question

- Broad: How intelligible is Danish to Swedish listeners without previous exposure? Here: Which factors predict whether a Danish word is easily decoded by Swedish pre-schoolers or not?
- Dependent variable: Word intelligibility Every word can be
 - Decoded (1), or
 - Not decoded (0)
- Independent variables:
 - Phonetic distance
 - Toneme
 - Number of 'difficult' sounds for the listener

Binary variable

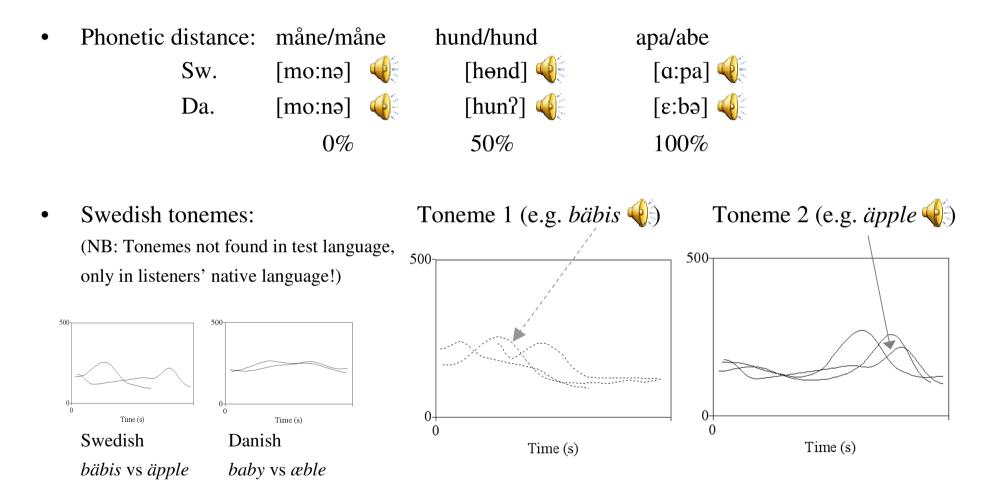
Continuous variable (0.00 - 1.00) Binary variable (0,1) Categorical variable (0,1,2,3,...)

Experiment

- 50 Danish word were auditorily presented to 12 Swedish children via headphones
- Similarly, 200 pictures (i.e. 4 pictures per sound) were presented visually on a touch screen
- The children were instructed to point to the corresponding picture
- Resulting data: Intelligibility scores per word per subject

 $50 \ge 12 = 600 \text{ scores}$

Independent variables: Examples

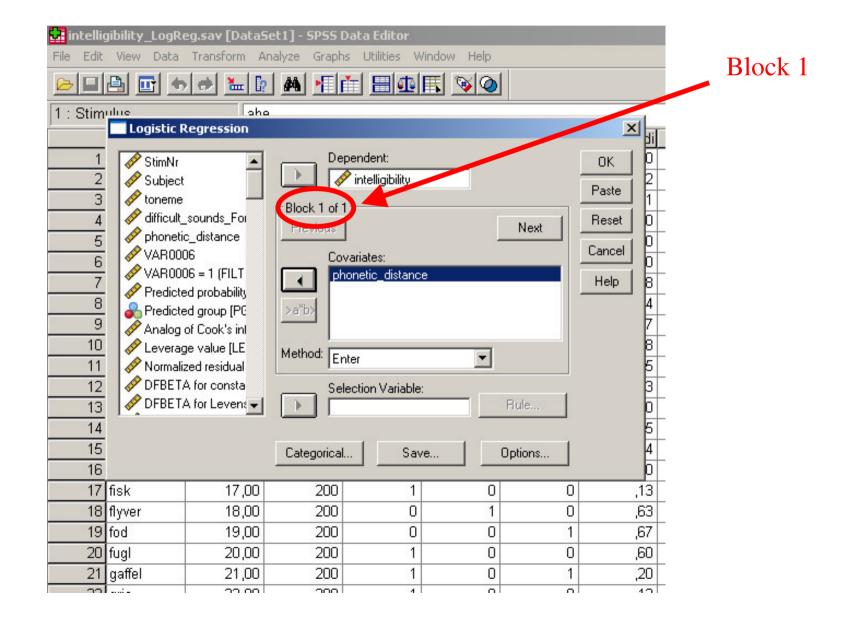


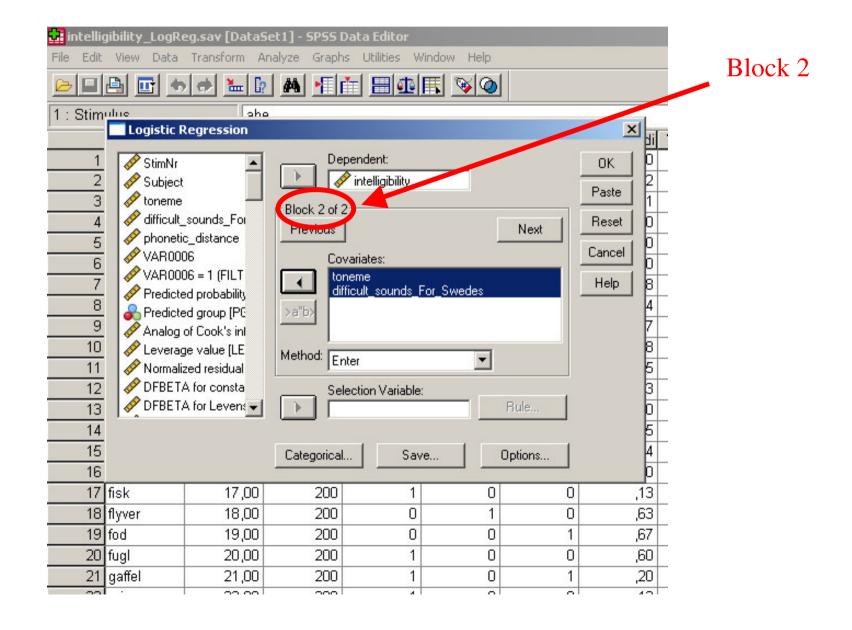
'Difficult sounds': Danish sounds that have been shown to be significantly more difficult to decode for Swedes (Schüppert & Gooskens, in prep.): [ε], [ð], [j], [Œ]

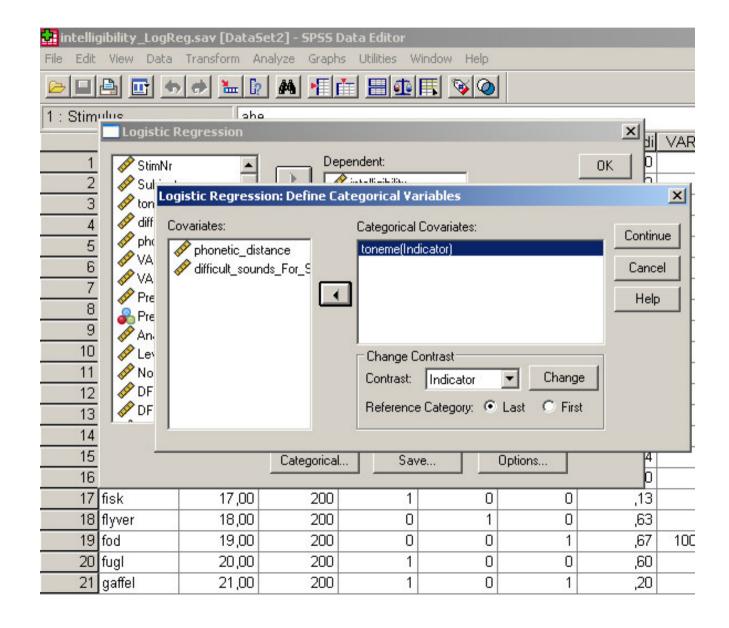
Data

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2	baby	2,00	200	1	0	1	,42					
3	badekar	3,00	200	1	1	2	,71					
4	ballon	4,00	200	1	0	0	,50					
5	banan	5,00	200	0	0	1	,50					
6	bil	6,00	200	1	0	0	,00,					
7	bjørn	7,00	200	1	0	1	,38					
8	blomst	8,00	200	0	1	0	,64					
9	bog	9,00	200	1	0	0	,67					
10	bold	10,00	200	1	0	0	,38					
11	bord	11,00	200	0	0	0	,75					
12	båd	12,00	200	1	0	1	,33					
13	cykel	13,00	200	1	0	0	,50					
14	dør	14,00	200	1	0	1	,75					
15	elefant	15,00	200	1	0	1	,14					
16	finger	16,00	200	1	0	0	,70					
17	fisk	17,00	200	1	0	0	,13					
18	flyver	18,00	200	0	1	0	,63					
19	fod	19,00	200	0	0	1	,67					
20	fugl	20,00	200	1	0	0	,60					
21	gaffel	21,00	200	1	0	1	,20					
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17	fisk	17,00	200		1	0	0	,13		
18	flyver	18,00	200		0	1	0	,63		
19	fod	19,00	200		0	0	1	,67		
20	fugl	20,00	200		1	0	0	,60		
21	gaffel	21,00	200		1	0	1	,20		

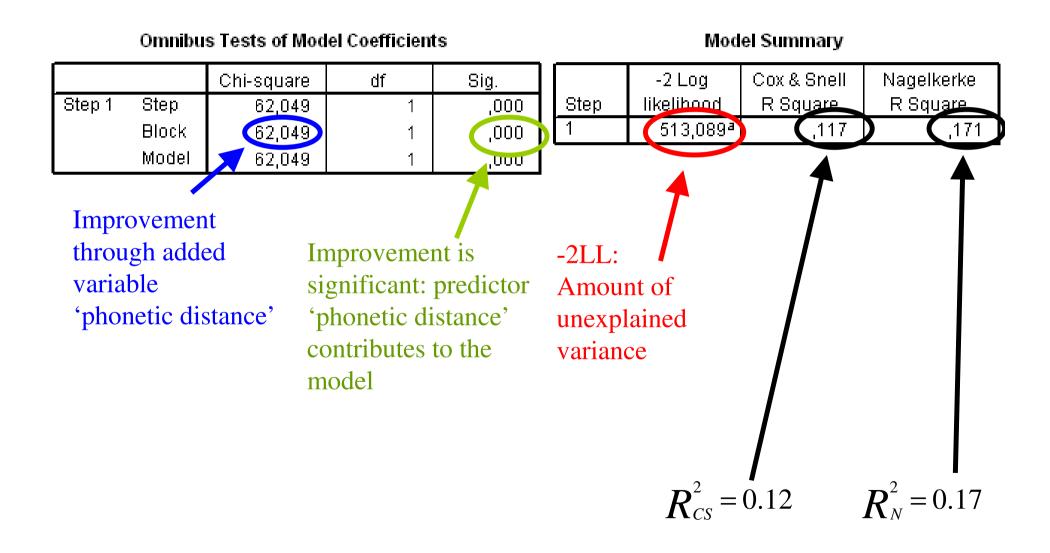






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20 ft		20,00	200	1	0	0		50
21 g	aπel	21,00	200	1	0	1	.4	20

Output: Block 1 (Phonetic distance)



Output: Block 1 (Phonetic distance)

Step

Constant

3,084

							Predicted	dicted	
			intelligibility					Percentage	
		Observed					1	Co	rrect
	Step 1	intelligibility	0			18	113		13,7
			1			12	357		96,7
		Overall Perc	entage						75,0
Exp(B) < 1 Indicates that pho distance correlate with intelligibilit	es negat		Turneti		•	oredicts 75% o			
		Va	riables in the	e Equation					
							95	,0% C.I.:	for EXP(B)
	В	S.E.	Wald	df	Sig.	Exp(E		ower	Upper
ep phonetic_distance	-3,704	,526	49,658	1	,000	- C.C	025	,009	,069

21,849

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1

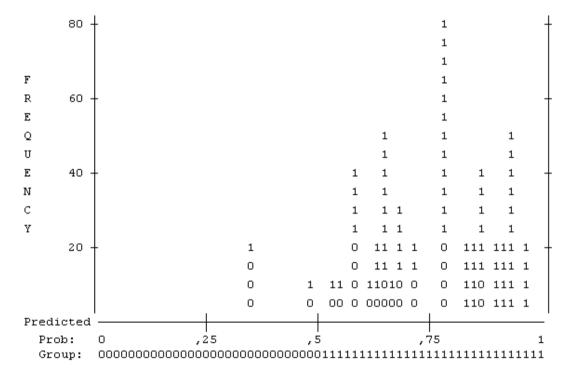
86,654

,331

Classification Table^a

Output: Block 1 (Phonetic distance)

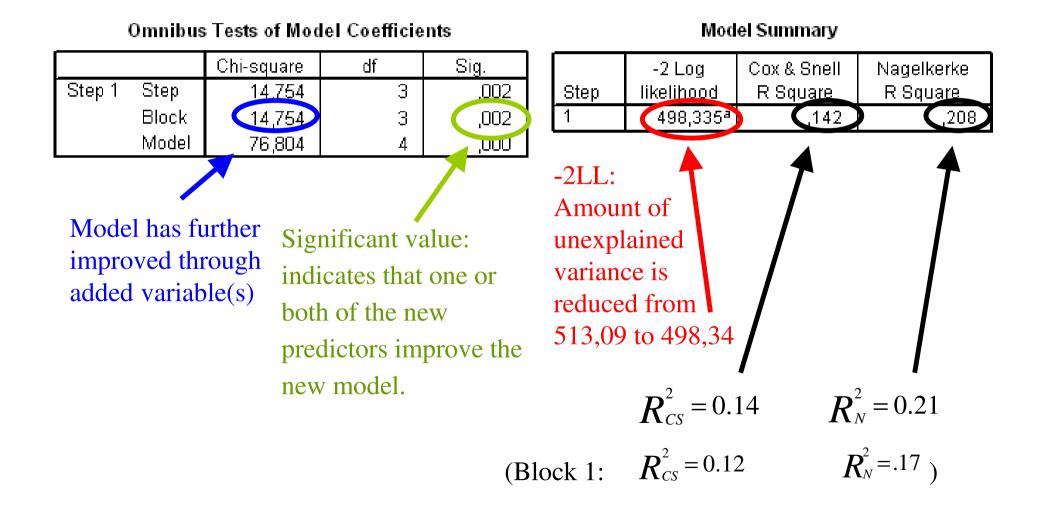
Observed Groups and Predicted Probabilities



Nondecoded stimuli seem to be difficult to predict (the zeroes should be concentrated further to left).

Decoded stimuli are more correctly predicted by the model (note the 1columns on the right hand side of the plot).

Output: Block 2 (Phonetic distance, Toneme, Difficult Sounds)



Output: Block 2 (Phonetic distance, Toneme, Difficult Sounds)

					· ·				
								95,0% C.I.:	for EXP(B)
		B	S.E.	Wald	df	Sig.	Exp(B)	Lower	Upper
Step	phonetic_distance	-2,950	,572	26,555	1	,000	,052	,017	,161
1	toneme(1)	,272	,286	,905	1	,341	1,313	,749	2,302
	difficult_sounds_ For_Swedes	-,363	,171	4,506	1	.034	,695	,497	,973
	Constant	2,746	,456	36,209	Γ	,000	5,573		

Variables in the Equation

Significant value indicates that variable 'difficult sounds' *improves* the model. Exp(B) < 1 indicates a negative correlation.

Non-significant value indicates that variable 'toneme' *does not* improve the model.

Results and Conclusion

Phonetic distance correlates negatively with intelligibility and contributes significantly to the model.

- Tonemes seem not to be contributing to the model. This phenomenon, that listeners are familiar with from their native language but that is missing in the test language, does not seem to puzzle the listeners.
- The number of difficult sounds correlate negatively with intelligibility and contribute significantly to the model.

Together, phonetic distance and number of strange sounds account for 14% to 21% of the variance.

References

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