Making statistic claims

Corpus Linguistics Kron

Outline of the session

- Lecture
 - Raw and normalised frequency
 - Descriptive statistics (mean, mode, media, measure of dispersion)
 - Inferential statistics (chi squared, LL, Fisher's Exact tests)
 - Collocation statistics

Quantitative analysis

- Corpus analysis is both qualitative and quantitative
- One of the advantages of corpora is that they can readily provide quantitative data which intuitions cannot provide reliably
- "The use of quantification in corpus linguistics typically goes well beyond simple counting" (McEnery and Wilson 2001: 81)

– What can we do with those numbers and counts?

Raw frequency

- The arithmetic count of the number of linguistic feature (a word, a structure etc)
- The most direct quantitative data provided by a corpus
- Frequency itself does **NOT** tell you much in terms of the validity of a hypothesis
 - There are 250 instances of the *f**k* swearword in the spoken BNC, so what?
 - Does this mean that people swear frequently or infrequently – when they speak?

Normalized frequency

- ...in relation to what?
 - Corpus analysis is inherently comparative
- There are 250 instances of the swearword in the spoken BNC and 500 instances in the written BNC
 - Do people swear twice as often in writing as in speech?
 - Remember the written BNC is 9 times as large as the spoken BNC
- When comparing corpora of different sizes, we need to normalize the frequencies to a common base (e.g. per million tokens)
 - Normalised freq = raw freq / token number * common base
 - The swearword is 4 times as frequent in speech as in writing
 - Swearword in spoken BNC = 250 / 10 * 1 = 25 per million tokens
 - Swearword in written BNC = 500 / 90 * 1 = 6 per million tokens
 - …but is this difference statistically significant?

Normalized frequency

- The size of a sample may affect the level of statistical significance
- Tips for normalizing frequency data
 - The common base for normalization must be comparable to the sizes of the corpora
 - Normalizing the spoken vs. written BNC to a common base of 1000 tokens?
- Warning
 - Results obtained on an irrationally enlarged or reduced common base are distorted

Descriptive statistics

- Frequencies are a type of descriptive statistics
- Descriptive statistics are used to describe a dataset
- A group of ten students took a test and their scores are as follows

- 4, 5, 6, 6, 7, 7, 7, 9, 9, 10

 How will you report the measure of *central* tendency of this group of test results using a single score?

The mean

- The mean is the arithmetic average
- The most common measure of central tendency
- Can be calculated by adding all of the scores together and then dividing the sum by the number of scores (i.e. 7)

- 4+5+6+6+7+7+7+9+9+10=70/10=7

 While the mean is a useful measure, unless we also knows how dispersed (i.e. spread out) the scores in a dataset are, the mean can be an uncertain guide

The mode and the median

- The mode is the most common score in a set of scores
 - The mode in our testing example is 7, because this score occurs more frequently than any other score
 - 4, 5, 6, 6, 7, 7, 7, 9, 9, 10
- The median is the middle score of a set of scores ordered from the lowest to the highest
 - For an odd number of scores, the median is the central score in an ordered list
 - For an even number of scores, the median is the average of the two central scores
 - In the above example the median is 7 (i.e. (7+7)/2)

Measure of dispersion: range

- The range is a simple way to measure the dispersion of a set of data
 - The difference between the highest and lowest frequencies / scores
 - In our testing example the range is 6 (i.e. highest 10 lowest 4)
- Only a poor measure of dispersion
 - An unusually high or low score in a dataset may make the range unreasonably large, thus giving a distorted picture of the dataset

Measure of dispersion: variance

- The variance measures the distance of each score in the dataset from the mean
 - In our test results, the variance of the score 4 is 3 (i.e. 7–4); and the variance of the score 9 is 2 (9–7)
- For the whole dataset, the sum of these differences is always zero
 - Some scores will be above the mean while some will be below the mean
- Meaningless to use variance to measure the dispersion of a whole dataset

Measure of dispersion: std dev

 Standard deviation is equal to the square root of the quantity of the sum of the deviation scores squared divided by the number of scores in a dataset

$$\sigma = \sqrt{\frac{\sum (F - \mu)^2}{N}}$$

- F is a score in a dataset (i.e. any of the ten scores)
- $-\mu$ is the mean score (i.e. 7)
- N is the number of scores under consideration (i.e. 10)
- Std dev in our example of test results is 1.687

Measure of dispersion: std dev

- For a normally distributed dataset (i.e. where most of the items are clustered towards the centre rather than the lower or higher end of the scale)
 - 68% of the scores lie within one standard deviation of the mean
 - 95% lie within two standard deviations of the mean
 - 99.7% lie within three standard deviations of the mean
- The standard deviation is the most reasonable measure of the dispersion of a dataset



Normal distribution (bell-shaped curve)

Computing std dev with SPSS

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1	A	4				
2	В	5				
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4	D	6				
5	E	7				
6	F	7				
7	G	7				
8	Н	8				
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10	J	10				

SPSS Menu - Analyze – Descriptive statistics - Descriptives

Descriptive Statistics

	Ν	Minimum	Maximum	Mean	Std. Deviation
score	10	4	10	6.80	1.687
Valid N (listwise)	10				

Inferential statistics

- Descriptive statistics are useful in summarizing a dataset
- Inferential statistics are typically used to formulate or test a hypothesis
 - Using statistical measures to test whether or not any differences observed are statistically significant
- Tests of statistical significance
 - chi-square test
 - log-likelihood (LL) test
 - Fisher's Exact test
- Collocation statistics
 - Mutual information (MI)
 - z score

Statistical significance

- In testing a linguistic hypothesis, it would be nice to be 100% sure that the hypothesis can be accepted
- However, one can never be 100% sure in real life cases
 - There is always the possibility that the differences observed between two corpora have been due to chance
 - In our swearword example, it is 4 times as frequent in speech as in writing
 - We need to use a statistical test to help us to decide whether this difference is statistically significant
- The level of statistical significance = the level of our confidence in accepting a given hypothesis
 - The closer the likelihood is to 100%, the more confident we can be
 - One must be more than 95% confident that the observed differences have **not** arisen by chance

Commonly used statistical tests

- Chi square test
 - ...compares the difference between the observed values (e.g. the actual frequencies extracted from corpora) and the expected values (e.g. the frequencies that one would expect if no factor other than chance was affecting the frequencies)
- Log likelihood test (LL)
 - Similar, but more reliable as LL does not assume that data is normally distributed
 - The preferred test for statistic significance

Commonly used statistical tests

- Interpreting results
 - The greater the difference (absolute value) between the observed values and the expected values, the less likely it is that the difference is due to chance; conversely, the closer the observed values are to the expected values, the more likely it is that the difference has arisen by chance
 - A probability value p close to 0 indicates that a difference is highly significant statistically; a value close to 1 indicates that a difference is almost certainly due to chance
 - By convention, the general practice is that a hypothesis can be accepted only when the level of significance is less than 0.05 (i.e. *p*<0.05, or more than 95% confident)

Online LL calculator

• <u>http://ucrel.lancs.ac.uk/llwizard.html</u>

	Corpus 1	Corpus 2
Frequency of word	250	500
Corpus size	1000000	9000000

Item	01	81	02	82	LL
Word	250	0.00	500	0.00 +	301.88

How to find the probability value p for an LL score of 301.88?

Contingency table

	right-handed	left-handed	TOTAL
male	43	9	52
female	44	4	48
TOTAL	87	13	100

degree of freedom (d.f.) = (No. of row -1) * (No. of column - 1)= (2 - 1) * (2 - 1) = 1 * 1 = 1

Critical values

d.f.	0.10	0.05	0.025	0.01	0.001
а 2		121 2423			
1	2.706	3.841	5.024	6.635	10.828
2	4.605	5.991	7.378	9.210	13.816
3	6.251	7.815	9.348	11,345	16.266
4	7.779	9.488	11.143	13.277	18.467
5	9.236	11.070	12.833	15.086	20.515
6	10.645	12.592	14.449	16.812	22.458
7	12.017	14.067	16.013	18.475	24.322
8	13.362	15.507	17.535	20.090	26.125
9	14.684	16.919	19.023	21.666	27.877
10	15.987	18.307	20.483	23.209	29.588

The chi square test or LL test score must be greater than 3.84 (1 d.f.) for a difference to be statistically significant.

Oakes, M (1998) Statistics for Corpus Linguistics, EUP, p. 266

In the example of swearword in spoken/written BNC, LL 301.88 for 1 d.f. More than 99.99% confident that the difference is statistically significant

Excel LL calculator by Xu

1	L	.og-like	lihood	Ratio Cal	culator		
2							
3	St	ep 1. Enter the 🕯	corpus sizes ir	A and B.			
4	St.	ep 2. Enter the i	trequency cou	nts in columns B a	nd C.		
5		ne white cells are c	ata cells; the gray	ones are result cells.			
6 7			А		В		
8		Corpus Size 1	52191	Corpus Size 2	52877		
9							
10	Word	Freq. in Corpus 1	Freq. in Corpus 2	Log-likelihood	Sig.		
11	will	224	138	21.77	0.000 *** +		
12	can	198	192	0.19	0.665 +		
13	would	169	125	7.20	0.007 ** +		
14	could	72	66	0.35	0.557 +		
15	must	67	30	14.96	0.000 *** +		
16	have to	132	41	51.56	0.000 *** +		
17	should	130	55	32.29	0.000 *** +		
18	may	51	35	3.21	0.073 +		
19	might	67	8	53.82	0.000 *** +		
20	ought to	10	3	4.07	0.044 * +		
21	shall	5	2	1.37	0.242 +		

www.corpus4u.org/attachment.php?attachmentid=560&d=1240826440

SPSS: Left- vs. right-handed

Define variables

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	Name	Туре	Width	Decimals		
1	gender	String	8	0		
2	tendency	String	8	0		
3	number	Numeric …	8	0		

weight case (Data – Weight cases)

Data view

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	gender	tendency	numł	per		
1	male	right-h		43		
2	male	left-h		9		
3	female	right-h		44		
4	female	left-h		4		

Weight Cases		
	 Do not weight cases Weight cases by Frequency Variable: Image: number Current Status: Do not weight cases 	OK Paste Reset Cancel Help

SPSS: Left- vs. right-handed

Cross-tab



SPSS: Left- vs. right-handed

	Value	df	Asymp. Sig. (2-sided)	Exact Sig. (2-sided)	Exact Sig. (1-sided)
Pearson Chi-Square	1.777 ^b	1	<u> </u>	Þ	
Continuity Correction	1.072	1	.300		
⊈ <u>ikelih</u> ood R <u>atio</u> ⊃	1.825	1	.177	\flat	
Fisher's Exact Test				.239	.150
N of Valid Cases	100				

a. Computed only for a 2x2 table

b.O cells (.0%) have expected count less than 5. The minimum expected count is 6.

24.

Any cells with an expected value less than 5?

Critical value (X² / LL) for 1 d.f. at p<0.05 (95%): 3.84

Is there a relationship between gender and left- or righthandedness?

- The chi-square or log-likelihood test may not be reliable with **very low frequencies**
 - When a cell in a contingency table has an expected value less than 5, Fisher's Exact test is more reliable
 - In this case, SPSS computes Fisher's exact significance level automatically when the chisquare test is selected
 - SPSS Releases 15 and 16 have removed the Fisher's Exact test module, which can be purchased separately

	men	women	TOTAL
dieting	1	8	9
no dieting	9	2	11
TOTAL	10	10	20

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16 :								
gender		tendency	number					
1 male		dieting	1					
2 male		no-diet	9					
3 female		dieting	8					
4 female		no-diet	2					
-								

Don't forget to weight cases!

<u> </u>	Crosstabs: Cell Display	×						
	Counts	ОК						
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	🗖 Row 📃 Unstandardized	Help						
	Column Standardized							
	Total Adjusted standardized							
	Noninteger Weights							
	Round cell counts O Round case weights							
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Exact Statistics Cells Format								



gender * tendency Crosstabulation

			tendency				
			d	ieting	no-diet	Total	
gender	female	Count		8	2	10	
		(Expected Count		4.5)	5.5	10.0	
	male	Count			9	10	
		Expected Count		/ 4.5	5.5	10.0	
Total		Count		/ 91	11	20	
		Expected Count		9.0	11.0	20.0	

	Value	df	Asymp. Sig. (2-sided)	Exact Sig. (2-sided)	Exact Sig. (1-sided)
Pearsøn Chi-Square	9.899 ^b	1	.002		
Conti⁄nuity Correction	7.273	1	.007		
Like/ihood Ratio	11.016	1	.001		
Fisher's Exact Test				.005	.003
N øf Valid Cases	20				

′a. Computed only f∳r a 2×2 table

¹ b. 2 cells (50.0%) have expected count less than 5. The minimum expected count is 4 50.

Force an FE test

Crosstabs		×					
Exact Tests Asymptotic only Monte Carlo Confidence level: 99 Number of samples: 10000 Exact	Continue Cancel % Help	OK Paste Reset Cancel Help					
Time limit per test: 5	h	Value	df		Asymp. Sig. (2-sided)	Exact Sig. (2-sided)	Exact Sig. (1-sided)
Exact method will be used instead of Mont computational limits allow. For nonasymptotic methods, cell counts an or truncated in computing the test statistics	Pearson Chi-Square Continuity Correction Likelihood Ratio Fisher's Exact Test	9.899 ^b 7.273 11.016		1 1 1	.002 × .007 .001 ×	.005	.003
Exact Statistics	N of Valid Cases a. Computed only f b. 2 cells (50.0%) h 50.	20 or a 2×2 tabl ave expected	e I count	less	than 5. The mir	nimum expe	xted count is <

Practice

- Use both the UCREL/Xu's LL calculator / SPSS to determine if the difference in the frequencies of passives in the CLEC and LOCNESS corpora is statistically significant
 - CLEC: 7,911 instances in 1,070,602 words
 - LOCNESS: 5,465 instances in 324,304 words



Collocation statistics

- **Collocation:** the habitual or characteristic co-occurrence patterns of words
 - Can be identified using a statistical approach in CL, e.g.
 - Mutual Information (MI), *t* test, *z* score
 - Can be computed using tools like SPSS, Wordsmith, AntConc, Xaira
 - Only a brief introduction here
 - More discussions of collocation statistics to be followed

Mutual information

 Computed by dividing the observed frequency of the co-occurring word in the defined span for the search string (socalled *node word*), e.g. a 4:4 window, by the expected frequency of the co-occurring word in that span and then taking the logarithm to the base 2 of the result

Mutual information

- A measure of collocational strength
- The higher the MI score, the stronger the link between two items
 - MI score of 3.0 or higher to be taken as evidence that two items are collocates
- The closer to 0 the MI score gets, the more likely it is that the two items co-occur by chance
- A negative MI score indicates that the two items tend to shun each other

The t test

- Computed by subtracting the expected frequency from the observed frequency and then dividing the result by the standard deviation
- A *t* score of 2 or higher is normally considered to be statistically significant
- The specific probability level can be looked up in a table of *t* distribution

The z score

- The *z* score is the number of standard deviations from the mean frequency
- The *z* test compares the observed frequency with the frequency expected if only chance is affecting the distribution
- A higher z score indicates a greater degree of collocability of an item with the node word