



Seminar in Methodology and Statistics

Fisher's Exact Test

as a measure of association strength

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Fisher's Exact Test

- › Used to test for associations between two variables
 - Identifying dependent bigrams
- › Computes a p-value
- › Calculates significance exactly (unlike χ^2 test)
- › Based on **hypergeometric** distribution
 - Drawing from a finite population without replacement



Using Fisher's Exact Test

- › Natural language data is skewed
 - Fisher's test does not require a normal distribution of data
- › Sparse data problem
 - Fisher's test can be used with small sample sizes

However, Fisher's Exact Test is more computationally intensive



Using Fisher's Exact Test

- › Bigram association
 - Identifying dependent word pairs
white, snow

- › Standard test for collocation analysis
 - Computing the association between words and syntactic frames (Stefanowitsch and Gries, 2003)
[N *waiting to happen*]



Problem: Determining animacy of nouns

- › Human: doctor, player, photographer, Englishman
- › Inanimate: banana, Netherlands, feeling, crime

- › Automatically determine this based on co-occurrence with verbs

- › The doctor thought John was right
- › The banana thought John was right



Animacy: Data

- A dictionary of words and their animacy property
- A list of verbs and their subject arguments, extracted from a corpus, with frequency counts

<noun animacy="nonanimate">gevoel</noun>	85#blijf intransitive su#gevoel
<noun animacy="nonanimate">IJsselmeer</noun>	298#ontsta intransitive su#gevoel
<noun animacy="nonanimate">noord</noun>	1#schrijf transitive su#gevoel
<noun animacy="nonanimate">paasei</noun>	8#rest intransitive su#gevoel
<noun animacy="human">doctor</noun>	
<noun animacy="human">Engelsman</noun>	7#ontdek transitive su#Engelsman
<noun animacy="human">roker</noun>	4#ontwerp transitive su#Engelsman
<noun animacy="human">symfonieorkest</noun>	3#overschat transitive su#Engelsman
<noun animacy="nonhuman">fuchsia</noun>	
<noun animacy="nonhuman">pony</noun>	1#besta intransitive su#Yeti
<noun animacy="nonhuman">yeti</noun>	2#duik_op part_intransitive(op) su#Yeti
	1#lijk pc_pp(op) su#Yeti



Fisher's Exact Test on animacy data

- › Hypothesis: Animate nouns are associated with different verbs than inanimate nouns

- › Variables:
 1. Verb is “ontstaan” (*to start, to arise*)
 2. Subject is “gevoel” (*feeling*)

- › Binary variables

- › 4 classifications



Contingency table

- The Fisher's exact test is calculated using tables
- Totals are fixed

The noun “gevoel” (*feeling*) as a subject of the verb “ontstaan” (*to start, to arise*)

	gevoel	¬gevoel	Row totals
ontstaan	298	5927	6225
¬ontstaan	405	111952	112357
Column totals	703	117879	118582

$p < 0.00001$



Dependence and independence

- The p-value can go both ways: Association strength

The noun “gevoel” (*feeling*) as a subject of the verb “schrijven” (*to write*)

	gevoel	¬gevoel	Row totals
schrijven	1	299	300
¬schrijven	702	117578	118282
Column totals	703	117879	118582

$p > 0.99999$



Association strength

- › This p-value can be used as a measure of association strength
- › A low value indicates a strong association, a high value indicates none
- › Because the totals are fixed, you cannot compare p-values from samples of different sizes



Hypothesis

- › H_0 : The noun x and the verb y are independent in subject relations
- › H_1 : The noun x occurs as a subject of the verb y more often than would be expected by chance



Calculating the value

- › The p-value expresses the total probability of the observed distribution (table) and all the more extreme ones

	gevoel	¬gevoel
ontstaan	298	5927
¬ontstaan	405	111952

	gevoel	¬gevoel
ontstaan	300	5925
¬ontstaan	403	111950

	gevoel	¬gevoel
ontstaan	299	5926
¬ontstaan	404	111951

	gevoel	¬gevoel
ontstaan	301	5924
¬ontstaan	402	111949



Calculating the value

	gevoel	¬gevoel	totals
ontstaan	298	5927	6225
¬ontstaan	405	111952	112357
totals	703	117879	118582

$$\text{> } P(n) = \frac{6225! * 112357! * 703! * 117879!}{298! * 5927! * 405! * 111952! * 118582!}$$

$$\text{> } P(n + 1) = \frac{6225! * 112357! * 703! * 117879!}{299! * 5926! * 404! * 111951! * 118582!}$$

> etc

$$\text{> } p = P(n) + P(n + 1) + P(n + 2) + \dots$$

> A and B are associated more strongly than would be expected by chance ($\alpha = 0.001$)



Association strength

“gevoel” subject relations (inanimate)

0.0000000000000000	ontsta	<i>arise</i>
0.0000000000000830	heb	<i>have</i>
0.0000000000002380	speel	<i>play</i>
0.0000000000501125	ben	<i>be</i>
0.0000000003404273	zeg	<i>say</i>
0.731409478841741	krijg	<i>get</i>
0.823487761949459	spreek	<i>speak</i>
0.853510038160385	neem	<i>take</i>
0.902189553992116	ken	<i>know</i>
1.0000000000002866	schrijf	<i>write</i>



Association strength

“hippie” subject relations (human)

0.001468162077883	ga	<i>go</i>
0.019216198962412	kom	<i>come</i>
0.048523337414639	noem	<i>call, name</i>
0.053750193619017	zeg	<i>say</i>
0.101731760645688	vind	<i>think, find</i>
0.847872307894773	heb	<i>have</i>
1.0000000000000009	maak	<i>make</i>



Association strength

“haai” (*shark*) subject relations (nonhuman)

0.000512195152676	heb (<i>have</i>)
0.049576264380802	zit (<i>sit</i>)
0.067926380355665	verschijn (<i>appear</i>)
0.184926869408314	besta (<i>exist</i>)
1.0000000000000016	sta (<i>stand</i>)
1.0000000000000070	doe (<i>do</i>)
1.0000000000000196	maak (<i>make</i>)
1.0000000000000472	lig (<i>lie, be situated</i>)

‘Zit’ is also a position verb



Classification task

- › For a list of nouns, decide whether they refer to something human, nonhuman animate or inanimate, using these subject-verb associations
- › Train a classifier on the weighted associations
- › Test it on novel data (unseen nouns)



Evaluation

Measure of association	Correctly classified
Pointwise Mutual Information	90.64%
Fisher's Exact Test	88.85%
Frequency	83.09%
None (Baseline)	81.81%



Fisher's Exact Test for association strength

- › Fisher's Exact Test is a very robust measure
- › It is computationally intensive
- › Cannot compare data from samples of different sizes

- › For this animacy classification task, PMI seems to perform better



References

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Questions?



Thank you for your attention