

Behaviour Profiles and hierarchical cluster analysis

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Outline

1. Introduction to Behaviour Profiles and hierarchical cluster analysis

2. Univariate BP: verbs of communication

3. Multivariate BP: polysemy of SPEAK

Behavioural Profiles

- BP are a popular method of comparing the corpusbased distributional properties of several near synonyms or word senses.
- Based on ideas of Atkins (1987), Hanks (1996)
- Developed by Divjak (2003) and Gries (2006)

Steps of BP analysis

1. Create the distributional profiles of your units.

2. Compute distances between them.

3. Represent them visually, e.g. with the help of a cluster analysis.

Distributional profiles

Verb	Transitive	Intransitive	Clause
walk	1	99	0
think	5	50	45
believe	30	20	50



Verb	Transitive	Intransitive	Clause
walk	0.01	0.99	0
think	0.05	0.5	0.45
believe	0.3	0.2	0.5

Steps of BP analysis

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Manhattan distances

- The simplest method: take the absolute differences between *a* and *b* and sum them up.
- E.g. think and walk:

```
> abs(0.01 - 0.05) + abs(0.99 - 0.5) + abs(0 -
0.45)
[1] 0.98
```

Manhattan distances with R

Euclidean distances

- In real life, represents the distance between objects 'as the crow flies'.
- Euclidean distance is computed as the square root of the sum of the squared differences between *a* and *b* in each column.
- Cf. Pythagoras' law!
- E.g. think and walk:

```
> sqrt((0.01 - 0.05)^2 + (0.99 - 0.50)^2 + (0 -
0.45)^2)
[1] 0.6664833
```

Euclidean distances with R

```
> dist(test) #the default option
```

walk think

think 0.6664833

believe 0.9788769 0.3937004

Canberra distances

- sum of all |a b|/|a + b| for each column (i.e. absolute differences and absolute sums)
- E.g. *think* and *walk*:
 - Transitive: |0.05 0.01|/|0.05 + 0.01| = 0.67
 - Intransitive: |0.5 0.99|/|0.5 + 0.99| = 0.33
 - Clause: |0.45 0|/|0.45 + 0| = 1
 - Total: 0.67 + 0.33 + 1 = 1.99

Canberra distances with R

Steps of BP analysis

1. Create the distributional profiles of your units.

2. Compute distances between them.

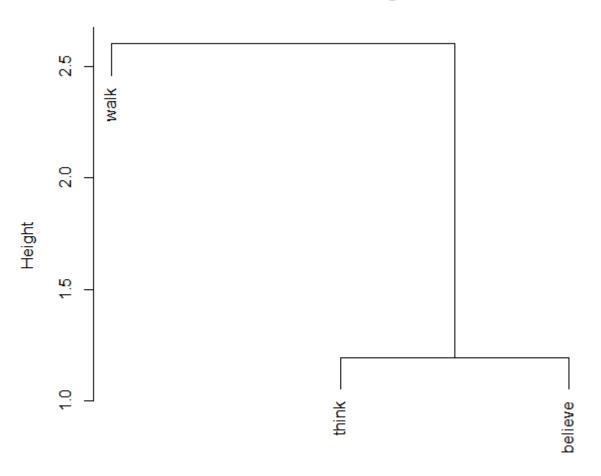
3. Represent them visually, e.g. with the help of a cluster analysis.

Hierarchical cluster analysis

- Take a distance matrix.
- Pick the smallest distance between two objects and merge them in one cluster.
- Then pick the next smallest distance between two objects and/or clusters and merge them.
- Stops when all objects are merged in one cluster tree.

```
> plot(hclust(test.dist))
```

Cluster Dendrogram



test.dist hclust (*, "complete")

Clustering methods

- OK, but how to compute distances between clusters?
 - Single (the minimally possible distances between the clusters are compared, and the smallest is taken)
 - Complete (the maximally possible distances between clusters are compared, and the smallest is taken)
 - Average (the average distances between the clusters are computed, and the smallest is taken)
 - Ward (based on variance minimization)

A simple analogy

- Imagine you have a choice between two clubs. How do you decide which one to join?
- Single: you choose the club your best friend has joined.
- Complete: you find out which club your biggest enemy is a member of. You choose the other one.
- Average: you choose the club which has on average more likable members.

What is your own social strategy?

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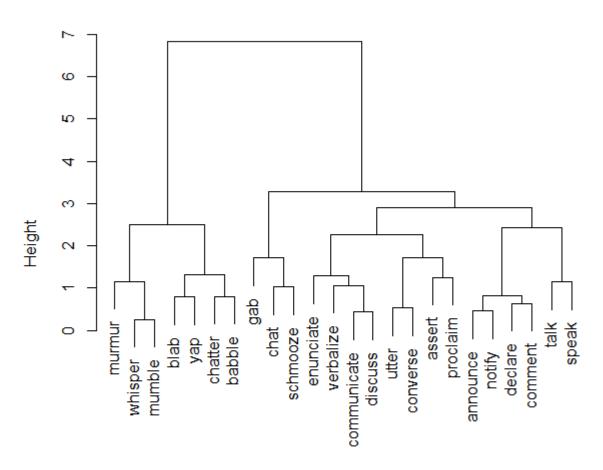
2. Univariate BP: verbs of communication

3. Multivariate BP: polysemy of SPEAK

Verbs of communication revisited

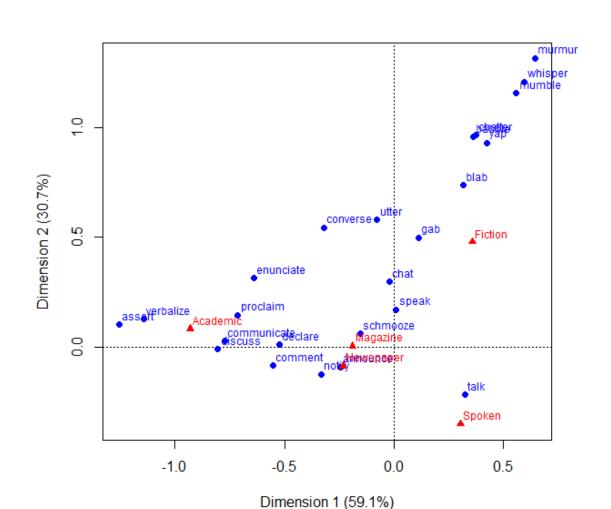
```
1. Transform the frequencies into proportions (row sums = 1):
> speak.bp <- prop.table(as.matrix(speak), 1)</pre>
2. Compute the distances between the verbs:
> speak.dist <- dist(speak.bp, method =</pre>
"canberra")
3. Perform a hierarchical cluster analysis:
> speak.clust <- hclust(speak.dist, method =</pre>
"ward.D2")
> plot(speak.clust)
```

Cluster Dendrogram



speak.dist hclust (*, "ward.D2")

Compare: simple CA (Dimensions 1 & 2)



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2. Univariate BP: verbs of communication

3. Multivariate BP: polysemy of SPEAK

Data: SPEAK from COCA

> str(speak_poly)

12 senses of SPEAK

1. Produce words

Cats and dogs can't speak.

2. Express thoughts or feelings about something

He spoke at one point about whether or not this crisis could be resolved between Arabs or between Iraq and the rest of the world.

- 3. Touch upon a topic, mention, bring up Have I spoken of her looks?
- 4. Regarding, as for

Speaking of lust, let's hit the trail.

12 senses of SPEAK (cont.)

5. Characterize, define smth/smb as smth/smb

He frequently spoke of the Roman communion as "the only true Church..."

6. Act as a spokesperson

Ladies and gentlemen, I, your captain, speak for the crew of the S.S. Ariel, bidding you farewell.

7. Engage in a conversation with smb.

I spoke with you this morning.

8. Speak publicly

Speaking in Colombia, the president added new priorities.

12 senses of SPEAK (cont.)

9. Tell

They'll see me as a man who speaks the truth and has suffered on.

10. Pronounce

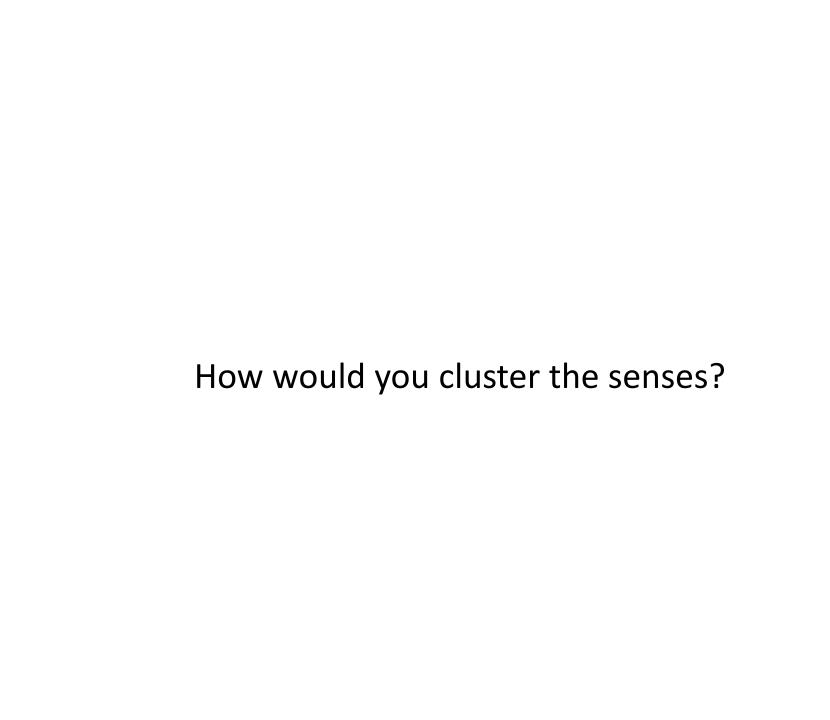
I'm not sure he spoke my name.

11. Have command of a language

Jakov had lived in America for thirty years but he did not speak English.

12. Express non-verbally

...the pseudosmile around the corners of his mouth spoke volumes.



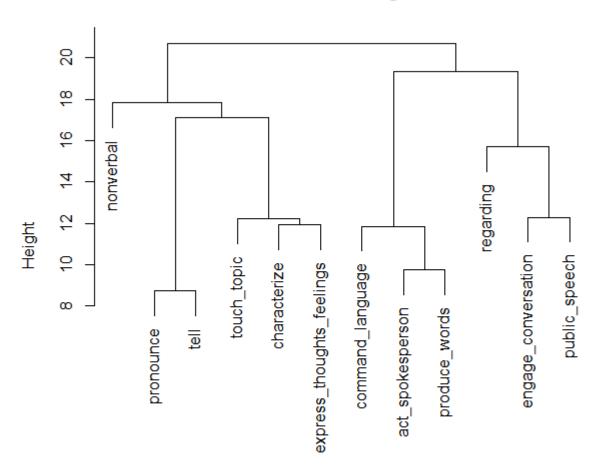
Creating BP for many variables

```
First, split the data frame into a list with data for individual senses:
> speak.split <- split(speak poly,</pre>
speak poly$Sense)
Remove the first column with senses (no longer necessary):
> speak.split <- lapply(speak.split, function(x) x
= x[, -1])</pre>
Create the BP vectors for each of the senses:
> speak.split.bp <- lapply(speak.split, bp)</pre>
Put the vectors in one matrix:
> speak.bp <- do.call(rbind, speak.split.bp)</pre>
```

Clustering the senses: Canberra distances, Ward clustering

```
> speak.dist <- dist(speak.bp, method =
"canberra")
> speak.clust <- hclust(speak.dist, method =
"ward.D2")
> plot(speak.clust)
```

Cluster Dendrogram

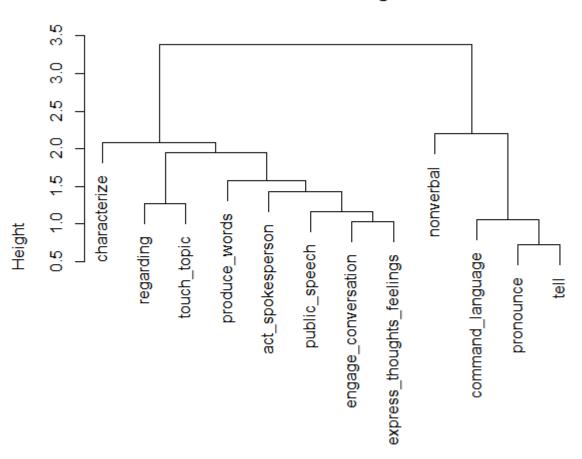


speak.dist hclust (*, "ward.D2")

Trying other options: Euclidean distance, Ward clustering

```
> speak.dist <- dist(speak.bp, method =
"euclidean")
> speak.clust <- hclust(speak.dist, method =
"ward.D2")
> plot(speak.clust)
```

Cluster Dendrogram

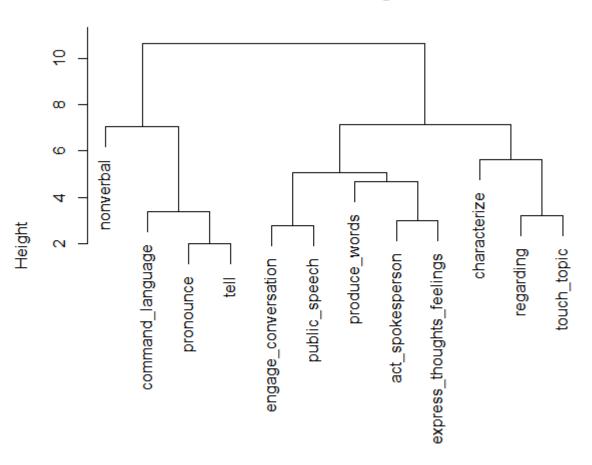


speak.dist hclust (*, "ward.D2")

Trying other options: Manhattan distance, Ward clustering

```
> speak.dist <- dist(speak.bp, method =
"manhattan")
> speak.clust <- hclust(speak.dist, method =
"ward.D2")
> plot(speak.clust)
```

Cluster Dendrogram

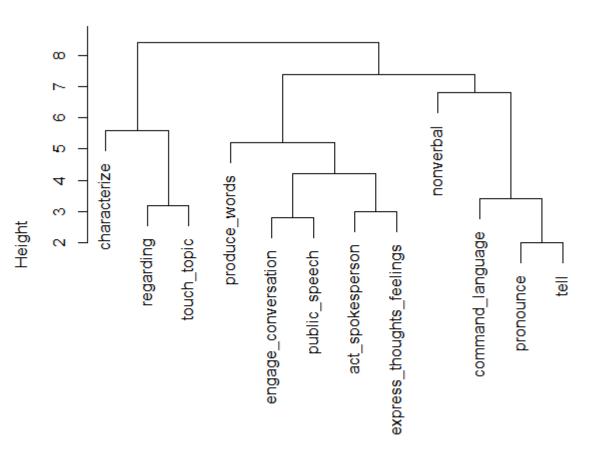


speak.dist hclust (*, "ward.D2")

Trying other options: Manhattan distances, complete clustering

```
> speak.dist <- dist(speak.bp, method =
"manhattan")
> speak.clust <- hclust(speak.dist, method =
"complete")
> plot(speak.clust)
```

Cluster Dendrogram

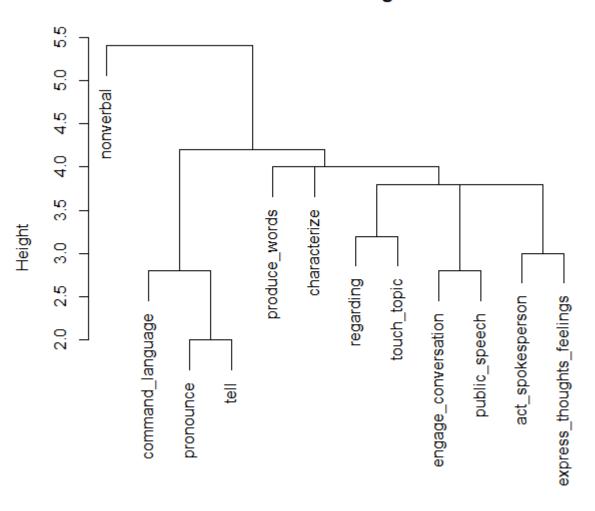


speak.dist hclust (*, "complete")

Trying other options: Manhattan distance, single clustering

```
> speak.dist <- dist(speak.bp, method =
"manhattan")
> speak.clust <- hclust(speak.dist, method =
"complete")
> plot(speak.clust)
```

Cluster Dendrogram

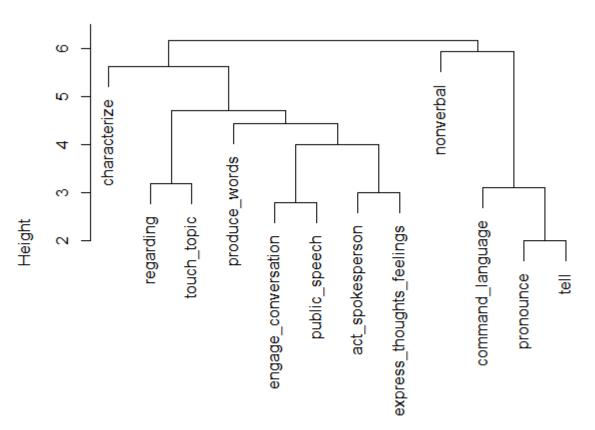


speak.dist hclust (*, "single")

Trying other options: Manhattan distance, average clustering

```
> speak.dist <- dist(speak.bp, method =
"manhattan")
> speak.clust <- hclust(speak.dist, method =
"average")
> plot(speak.clust)
```

Cluster Dendrogram



speak.dist hclust (*, "average")

Interpretation

• Which of the solutions speaks to your intuitions the most? The least?

References

- Atkins, B.T.S. (1987). Semantic ID tags: Corpus evidence for dictionary senses. The uses of large text databases. Proceedings of the Third Annual Conference of the UW Centre for the New Oxford English Dictionary, 17–36. Waterloo, Canada.
- Divjak, D. (2003). On trying in Russian: A tentative network model for near(er) synonyms. In *Belgian Contributions to the 13th International Congress of Slavicists*, Ljubljana, 15–21 August 2003. Special issue of *Slavica Gandensia*, 25–58.
- Gries, S. Th. (2006). Corpus-based methods and Cognitive Semantics: The many senses of to run. In S. Th. Gries & A. Stefanowitsch (Eds.), Corpora in Cognitive Linguistics. Corpusbased Approaches to Syntax and Lexis, 57–99. Berlin/New York: Mouton de Gruyter.
- Hanks, P. (1996). Contextual dependency and lexical sets. *International Journal of Corpus Linguistics* 1(1). 75–98.