DramaAnalysis

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Preface

This book is in alpha stadium and not ready. It pertains to the (currently unreleased) version 3.0 of the DramaAnalysis package. Use at your own risk.

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Chapter 1

Introduction

- 1.1 Technical basics
- 1.1.1 Classes
- 1.1.2 Pipelining with margrittr

DramaAnalysis 3.0

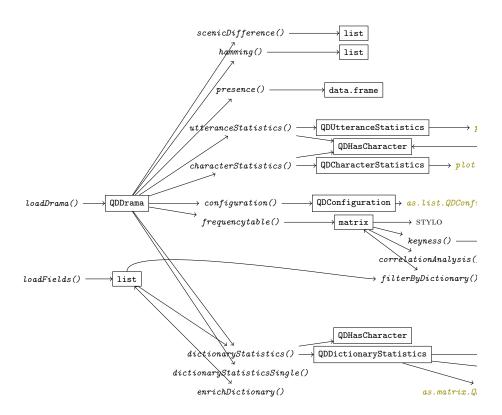


Figure 1.1: Some caption.

Chapter 2

Data

Before analysing any data, it needs to be imported, and converted into the proper structure. In QuaDramA, we process dramatic texts in multiple stages, described below.

2.1 Origin: TEI-XML

The base format that we use (and in which we put all our annotations) is an XML format known as TEI. This format is used by most researchers doing quantitative drama analysis. An excellent source for dramas in the proper format is DraCor, maintained by Frank Fischer.

While we are using GerDraCor as a basis, we have added linguistic annotations to a number of plays, and integrated more plays (e.g., translations) into the corpus. This corpus can be found here.

2.2 Preprocessed data

As a first step, we process all dramatic texts using our DramaNLP pipeline. The result of this processing is a set of CSV files for each play that contains the information in the play in a format suitable for analysis with R. This repository contains two plays in the format, which are also included in our R package.

Table 2.1: Different CSV files used in the analysis. *ID* is a place-holder for a unique identifier for the play

File	Description
ID.Metadata.csv	Meta data for the play (author, title, language,)
ID.Characters.csv	Characters of the play, with some character specific information
ID.Entities.csv	All discourse entities (including characters, but also all other coreference
	chains)
ID.Mentions.csv	Mentions associated with characters
ID. Stage Directions. csv	The stage directions of the play
ID.UtterancesWithTokens.csv	All character utterances of the play
ID. Segments. csv	Information about acts and scenes

2.3 Sample data

Load Emilia Galotti
data(rksp.0)

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```
# Load Miß Sara Sampson
data(rjmw.0)

text <- combine(rksp.0, rjmw.0)
```

Part I Structure Analysis

We're assuming here that we have loaded some texts using loadDrama(), and that this text is stored as a QDDrama-object in the variable text. For demo purposes, we will use the two plays that are included in the R-package, Lessing's *Emilia Galotti* and *Miss Sara Sampson*. Both have been preprocessed by the DramaNLP pipeline.

Chapter 3

Who's talking how much?

First, we calculate summary statistics over all characters.

```
charStats <- characterStatistics(text)
charStats</pre>
```

##		corpus	drama	character	tokens	types	utterances
##	1	test	rjmw.0	sir_william	2056	698	23
##	2	test	rjmw.0	waitwell	1826	568	41
##	3	test	rjmw.0	der_wirt	324	177	7
##	4	test	rjmw.0	mellefont	7981	1722	201
##	5	test	rjmw.0	norton	1001	407	46
##	6	test	rjmw.0	betty	482	223	20
##	7	test	rjmw.0	sara	9121	1908	166
##	8	test	rjmw.0	marwood	7225	1757	155
##	9	test	rjmw.0	hannah	258	155	16
##	10	test	rjmw.0	der_bediente	44	34	2
##	11	test	rjmw.0	arabella	398	162	13
##	12	test	rksp.0	der_prinz	5303	1257	157
##	13	test	rksp.0	der_kammerdiener	42	34	6
##	14	test	rksp.0	conti	764	325	24
##	15	test	rksp.0	marinelli	5567	1324	221
##	16	test	rksp.0	camillo_rota	106	62	6
##	17	test	rksp.0	claudia_galotti	2098	657	73

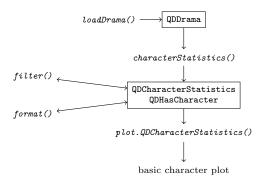


Figure 3.1: Relevant classes and functions in this chaper

##	18	test	rksp.0	pirro	343	196	25
##	19	test	rksp.0	odoardo	3248	891	108
##	20	test	rksp.0	angelo	635	300	28
##	21	test	rksp.0	emilia	2275	630	64
##	22	test	rksp.0	appiani	1112	426	48
##	23	test	rksp.0	battista	195	112	11
##	24	test	rksp.0	orsina	2832	743	64
##		uttera	${\tt nceLengthMean}$	utteranceLe	engthSd	${\tt firstBegin}$	lastEnd
##	1		89.39130	108	.948169	597	170838
##	2		44.53659	59	.768762	670	158683
##	3		46.28571	45	.492543	3571	5695
##	4		39.70647	50	.034972	5848	170188
##	5		21.76087	23	.601492	6328	170456
##	6		24.10000	36	.404019	9014	156987
##	7		54.94578	77	.785973	11766	167457
##	8		46.61290	62	.383230	28320	141005
##	9		16.12500	15	.654073	28392	58523
##	10		22.00000	12	.727922	31532	31790
##	11		30.61538	31	.967532	43987	52559
##	12		33.77707	40	.807176	426	136067
##	13		7.00000	5	.215362	1149	24954
##	14		31.83333	40	. 141778	2654	12212
##	15		25.19005	29	.844599	13147	134486
##	16		17.66667	19	. 179851	25577	26914
##	17		28.73973	29	. 131897	27006	112526
##	18		13.72000	10	.663489	27113	50425
##	19		30.07407	40	. 399571	27385	135555
##	20		22.67857		.573326	28777	64877
##	21		35.54688		. 159105	36769	134940
##	22		23.16667		.525401	44452	56735
##	23		17.72727		.948720	67465	86481
##	24		44.25000	50	.977119	88094	112182

This already gives us a lot of information about the characters. In particular, the function characterStatistics() returns a table (of the types QDCharacterStatistics, QDHasCharacter and data.frame) with information about:

- the number of tokens a character speaks (tokens),
- the number of different tokens a character speaks (types),
- the number of utterances (utterances),
- the average length of the utterances (utteranceLengthMean),
- their standard deviation (utteranceLengthSd),
- the character position of the start of the first utterance (firstBegin), and
- the character position of the end of the last utterance (lastEnd).

The function characterStatistics() provides a number of options to control its exact behaviour. Entering ?characterStatistics in the R console opens the documentation for the function with a description of all the options. We'll describe some frequently used options here as well:

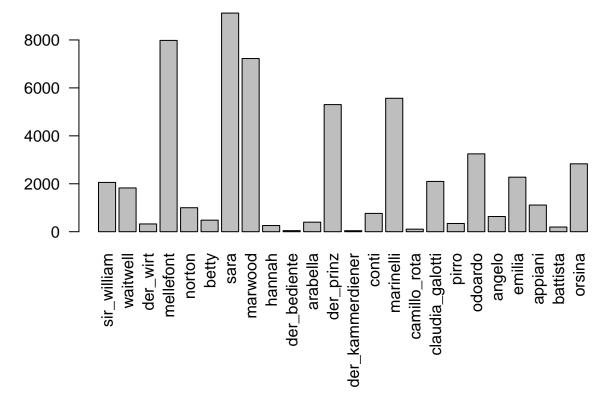
- Punctuation: By default, all punctuation marks are counted as tokens. This behaviour can be changed by setting filterPunctuation=TRUE.
- Normalization: The values in the table above are all absolute values. When comparing to other texts,
 one is often interested in normalized values. If the option normalize is set to TRUE, all values will be
 normalised (if applicable).
- Segmentation: By default, the function extracts values for the entire play. With the option segment,

it is possible to extract statistics by act or scene, as shown in the example below. Except for the additional column <code>Act</code>, the columns in the table are the same as before.

##			dwama	۸ ۵+	ah a ma a t a m	+ olrona	+	.++.~~
##	1	corpus	drama	I		2740	795	itterances 71
	2		rksp.0		<pre>der_prinz der_kammerdiener</pre>	42	34	6
	3		rksp.0	I	conti	764	325	24
	4		rksp.0	I	marinelli	1038	412	31
	5		rksp.0	I	camillo_rota	1036	62	6
	6		rksp.0	II	claudia_galotti	1252	477	50
	7		rksp.0	II	pirro	343	196	25
	8		rksp.0	II	odoardo	611	295	17
	9		rksp.0	II	angelo	385	214	19
	10		rksp.0	II	emilia	1166	417	28
	11		rksp.0	II	appiani	1112	426	48
	12		rksp.0	II	marinelli	527	255	28
	13		-	III	marinelli	1814	605	56
	14			III	der_prinz	945	398	29
##	15			III	angelo	250	147	9
##	16		rksp.0	III	battista	187	106	9
##	17		rksp.0	III	emilia	413	175	16
##	18		-	III	claudia_galotti	614	241	16
##	19		rksp.0	IV	der_prinz	803	319	30
##	20	test	rksp.0	IV	marinelli	1358	495	72
##	21	test	rksp.0	IV	battista	8	7	2
##	22	test	rksp.0	IV	orsina	2832	743	64
##	23	test	rksp.0	IV	odoardo	706	275	30
##	24	test	${\tt rksp.0}$	IV	claudia_galotti	232	125	7
##	25	test	rksp.0	V	marinelli	830	363	34
	26		rksp.0	V	der_prinz	815	321	27
	27		rksp.0	V	odoardo	1931	620	61
	28		rksp.0	V	emilia	696	251	20
##		uttera	_		n utteranceLengtl		_	
	1			.5915			426	26559
	2			.0000	0 5.2153			
	3			0000			1149	24954
	4			8333		778	2654	12212
	4		33	4838	34.2343	778 360	2654 13147	12212 23619
##	5		33 17	. 4838 . 6666	7 34.2343 7 19.1798	778 360 351	2654 13147 25577	12212 23619 26914
## ##	5 6		33 17 25	. 4838 . 6666 . 0400	7 34.2343 7 19.1798 0 28.927	778 360 351 750	2654 13147 25577 27006	12212 23619 26914 56666
## ## ##	5 6 7		33 17 25 13	.4838 .6666 .0400	34.2343 7 19.1798 0 28.927 0 10.6634	778 360 351 750 489	2654 13147 25577 27006 27113	12212 23619 26914 56666 50425
## ## ##	5 6 7 8		33 17 25 13 35	.4838 .6666 .0400 .7200	34.2345 7 19.1798 80 28.9277 90 10.6634 8 41.8008	778 360 351 750 489 323	2654 13147 25577 27006 27113 27385	12212 23619 26914 56666 50425 36244
## ## ## ##	5 6 7 8		33 17 25 13 35	.4838 .6666 .0400 .7200 .9411 .2631	34.2343 7 19.1798 10 28.9277 10 10.6634 8 41.8008 6 19.6746	778 360 351 750 489 323	2654 13147 25577 27006 27113 27385 28777	12212 23619 26914 56666 50425 36244 32436
## ## ## ## ##	5 6 7 8 9		33 . 17 . 25 . 13 . 35 . 20 .	. 4838 . 6666 . 0400 . 7200 . 9411 . 2631	34.2343 17 19.1798 10 28.927 10 10.6634 10 41.8008 10 63.826	778 360 351 750 489 323 594	2654 13147 25577 27006 27113 27385 28777 36769	12212 23619 26914 56666 50425 36244 32436 48503
## ## ## ## ## ##	5 6 7 8 9 10		33. 17. 25. 13. 35. 20. 41.	.4838 .6666 .0400 .7200 .9411 .2631 .6428	34.2343 17 19.1798 10 28.927 10 10.6634 11 18008 12 19.6746 13 18262 14 1808 15 18 18 18 18 18 18 18 18 18 18 18 18 18	778 360 351 750 489 323 594 277	2654 13147 25577 27006 27113 27385 28777 36769 44452	12212 23619 26914 56666 50425 36244 32436 48503 56735
## ## ## ## ## ##	5 6 7 8 9 10 11		33 . 17 . 25 . 13 . 35 . 20 . 41 . 23 . 18	. 4838 . 6666 . 0400 . 7200 . 9411 . 2631 . 6428 . 1666	34.2343 17 19.1798 10 28.927 10 10.6634 11 18008 12 19.6746 13 17 1862 14 18008 15 18 18 18 18 18 18 18 18 18 18 18 18 18	778 360 351 750 489 323 594 277 401	2654 13147 25577 27006 27113 27385 28777 36769 44452 50431	12212 23619 26914 56666 50425 36244 32436 48503 56735 55936
## ## ## ## ## ## ##	5 6 7 8 9 10 11 12 13		33. 17. 25. 13. 35. 20. 41. 23. 18. 32.	4838 6666 0400 7200 9411 2631 6428 1666 8214	34.2343 17 19.1798 10 28.9273 10 10.6634 10 41.8008 10 63.8263 17 25.5254 13 17.3463 16 35.2458	778 360 351 750 489 323 594 277 401 263	2654 13147 25577 27006 27113 27385 28777 36769 44452 50431 56826	12212 23619 26914 56666 50425 36244 32436 48503 56735 55936 79337
## ## ## ## ## ## ##	5 6 7 8 9 10 11		33 17 25 13 35 20 41 23 18 32 32	.4838 .6666 .0400 .7200 .9411 .2631 .6428 .1666 .8214 .3928	34.2343 17 19.1798 10 28.9273 10 10.6634 10 41.8008 10 63.8263 17 25.5254 13 17.3463 14 48.3273	778 360 351 750 489 323 594 277 401 263 982 393	2654 13147 25577 27006 27113 27385 28777 36769 44452 50431 56826 56914	12212 23619 26914 56666 50425 36244 32436 48503 56735 55936
## ### ## ## ## ## ##	5 6 7 8 9 10 11 12 13 14		33 17 25 13 35 20 41 23 18 32 32 27	4838 6666 0400 7200 9411 2631 6428 1666 8214	34.2343 17 19.1798 10 28.927 10 10.6634 8 41.8008 6 19.6746 16 63.8262 17 25.5254 13 17.3462 14 48.3273 15 19.4664	778 360 351 750 489 323 394 277 401 263 982 393 495	2654 13147 25577 27006 27113 27385 28777 36769 44452 50431 56826	12212 23619 26914 56666 50425 36244 32436 48503 56735 55936 79337 72918 64877
## ###################################	5 6 7 8 9 10 11 12 13 14 15		33. 17. 25. 13. 35. 20. 41. 23. 18. 32. 32. 27.	4838 6666 0400 7200 9411 2631 6428 1666 8214 3928	34.2343 17 19.1798 10 28.9273 10 10.6634 8 41.8008 6 19.6746 16 63.8262 17 25.5254 13 17.3462 14 48.3273 15 19.4664 18 36.0338	778 360 351 750 489 323 594 277 401 263 982 393 495	2654 13147 25577 27006 27113 27385 28777 36769 44452 50431 56826 56914 63132	12212 23619 26914 56666 50425 36244 32436 48503 56735 55936 79337 72918
######################################	5 6 7 8 9 10 11 12 13 14 15 16		33. 17. 25. 13. 35. 20. 41. 23. 18. 32. 32. 27. 20. 25.	.4838 .6666 .0400 .7200 .9411 .2631 .6428 .1666 .8214 .3928 .5862	34.2343 17 19.1798 10 28.9277 10 10.6634 8 41.8008 6 19.6746 16 63.8262 17 25.5254 13 17.3462 14 48.3273 18 19.4664 18 36.0339 10 27.3528	778 360 351 750 489 323 594 277 401 263 982 393 495 935 559	2654 13147 25577 27006 27113 27385 28777 36769 44452 50431 56826 56914 63132 67465	12212 23619 26914 56666 50425 36244 32436 48503 56735 55936 79337 72918 64877 75603

##	20	18.86111	20.750229	79972	103998
##	21	4.00000	2.828427	86363	86481
##	22	44.25000	50.977119	88094	112182
##	23	23.53333	18.528140	101533	112733
##	24	33.14286	29.952343	110037	112526
##	25	24.41176	36.802600	112824	134486
##	26	30.18519	37.235257	113752	136067
##	27	31.65574	47.382094	114998	135555
##	28	34.80000	34.284568	128910	134940

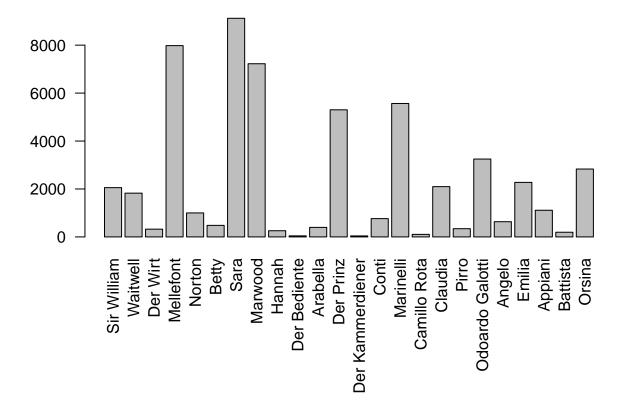
Of course, the values in the above table can be directly plotted:



3.1 Character names instead of identifiers

By default, all our functions identify characters using technical ids, which may or may not be human-readable. Even if they are, it's usually a good idea to replace them with nice to read labels before publication. We therefore provide the function format(), which can be applied to any table that contains a column with character ids (i.e., any object of type QDHasCharacter).

```
charStats <- characterStatistics(text, normalize=FALSE)</pre>
```

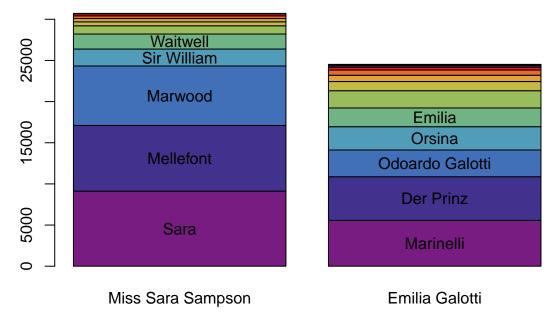


As can be seen above format() requires two arguments: The table in which we want to replace ids by characters, and the original drama object (that we got from calling the function loadDrama()).

3.2 Stacked bar plot

The plot shown above is quite wide, and some aspects (like the ranking in terms of spoken tokens), is hard to see. We often use another way of visualizing this, which can be used if you supply the QDCharacterStatistics directly into the barplot() function. In combination with the margrittr pipes, we can call it like this (enter?barplot.QDCharacterStatistics for details on the special barplot function):

```
characterStatistics(text, normalize=FALSE) %>%
  format(text) %>%
  barplot(names.arg=c("Miss Sara Sampson", "Emilia Galotti"))
```



By default, the barplot.QDCharacterStatistics() function visualises the number of tokens spoken by the characters (and ranks the characters accordingly). This can be changed by supplying the option column, and specifying another value.

```
characterStatistics(text, normalize=FALSE) %>%
  format(text) %>%
  barplot(names.arg=c("Miss Sara Sampson", "Emilia Galotti"),
          column = "utteranceLengthMean") # show mean utterance length
400
300
                   Waitwell
                   Der Wirt
200
                  Marwood
                                                    Odoardo Galotti
                                                          Conti
100
                     Sara
                                                       Der Prinz
                                                         Emilia
                 Sir William
                                                         Orsina
            Miss Sara Sampson
                                                     Emilia Galotti
```

This picture looks quite different! Other interesting columns to experiment with are types and utteranceLengthSd.

Chapter 4

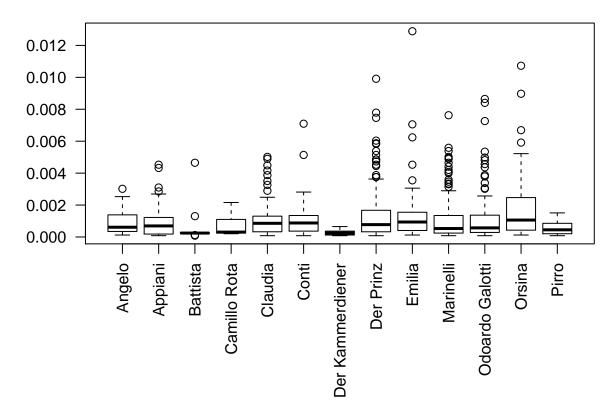
Who's talking how often?

So far, we have counted words for characters. Now we will turn to utterances, and their properties.

First, we will use the function utteranceStatistics() to extract quantitative information about utterances: utteranceStatistics(rksp.0)

```
##
      corpus drama
                            character utteranceBegin utteranceLength
## 1
                                                         3.874388e-03
        test rksp.0
                            der_prinz
                                                 426
## 2
        test rksp.0 der_kammerdiener
                                                1149
                                                         8.156607e-05
## 3
        test rksp.0
                            der_prinz
                                                1178
                                                         2.895595e-03
        test rksp.0 der_kammerdiener
## 4
                                                1526
                                                         6.525285e-04
## 5
        test rksp.0
                                                         2.854812e-04
                           der_prinz
                                                1655
        test rksp.0 der_kammerdiener
                                                1697
                                                         1.631321e-04
## 7
        test rksp.0
                            der_prinz
                                                1739
                                                         1.019576e-03
## 8
        test rksp.0 der_kammerdiener
                                                1857
                                                         3.262643e-04
## 9
        test rksp.0
                                                         2.528548e-03
                            der_prinz
                                                1919
                                                         3.670473e-04
## 10
        test rksp.0 der_kammerdiener
                                                2299
```

This creates a table that is very long, which is why we only show the first 10 rows here. The table contains one row for each utterance, and information about the speaker of the utterance, its length (measured in tokens) and its starting position (character position). We can now inspect the variance in utterance length:



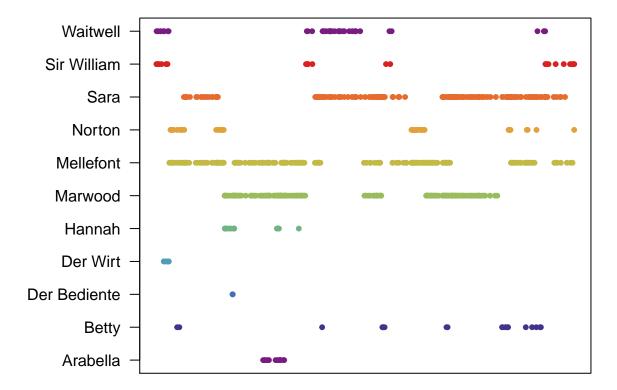
This uses the regular boxplot() function, enter ?boxplot for documentation. utteranceLength ~ character is called a formula in R and (in this case) expresses that we want to look at the column utteranceLength, grouped by the column character. The boxplot is a useful way to grasp the dispersion of a set of values (in this case: the lengths of all utterances by a character).

4.1 When are characters talking?

While the above displays the *length* of utterances, we can also display the position of utterances (remember the column utteranceBegin?). The following snippet visualizes when characters are talking, this time for Lessings Miss Sara Sampson:

```
par(mar=c(2,7,2,2))

utteranceStatistics(rjmw.0) %>%
  format(rjmw.0) %>%  # character names instead of ids
  plot()  # calling plot.QDUtteranceStatistics()
```

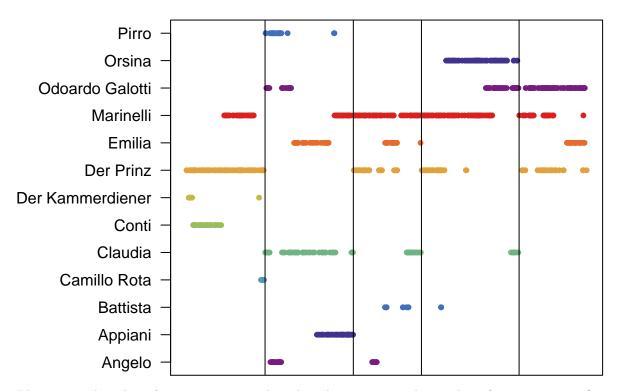


Each dot in this plot represents one utterance, the x-axis is measured in character positions. This is not really intuitive, but the flow from left to right represents the flow of the text. More technically, we again apply the function format() to display character names instead of character ids. This is the same function as above, just applied to a different table. It can be applied to any table of the type QDHasCharacter. Secondly, the call to the function plot() gets rerouted to the function plot.QDUtteranceStatistics(), because the object we supply as argument is of the type QDUtteranceStatistics. Information about this function can be retrieved by entering ?plot.QDUtteranceStatistics.

4.2 Adding act boundaries

Now it would be useful to include information on act/scene boundaries in this plot. This can be done by supplying the original drama object as a second argument to the plot() function:

```
par(mar=c(2,8,2,2))
utteranceStatistics(rksp.0) %>%
format(rksp.0) %>%
plot(rksp.0) # adding the `QDDrama` object here creates the act boundaries.
```



Please note that the information contained in this plot is very similar to the information in configuration matrices.

Chapter 5

Configuration

This section does not refer to the configuration of the R-package, but to the literary analysis concept configuration matrix (Pfister, 1988).

5.1 Matrices

Configuration matrices can be extracted with the function configuration(). As usual, entering configuration provies more detailed information about the options the function provides.

```
configuration(rksp.0)
```

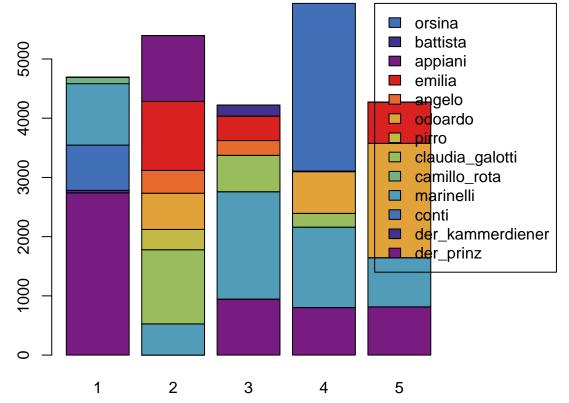
##		corpus	drama	character	1	2	3	4	5	
##	1	test	rksp.0	der_prinz	2740	0	945	803	815	
##	2	test	rksp.0	${\tt der_kammerdiener}$	42	0	0	0	0	
##	3	test	rksp.0	conti	764	0	0	0	0	
##	4	test	rksp.0	marinelli	1038	527	1814	1358	830	
##	5	test	rksp.0	camillo_rota	106	0	0	0	0	
##	6	test	rksp.0	claudia_galotti	0	1252	614	232	0	
##	7	test	rksp.0	pirro	0	343	0	0	0	
##	8	test	rksp.0	odoardo	0	611	0	706	1931	
##	9	test	rksp.0	angelo	0	385	250	0	0	
##	10	test	rksp.0	emilia	0	1166	413	0	696	
##	11	test	rksp.0	appiani	0	1112	0	0	0	
##	16	test	rksp.0	battista	0	0	187	8	0	
##	22	test	rksp.0	orsina	0	0	0	2832	0	

This creates a basic configuration matrix, but instead of just containing the presence or absence of a figure, it contains the number of spoken tokens for each act for each character. This information is in fact similar to what we can extract with characterStatistics(rksp.0, segment="Act"), but in a different form and structure. The above table contains a lot of information that can be visualised.

We first need to extract the numeric content from the above table. This can easily be done with the function as.matrix() (this will, in detail, be rerouted to the function as.matrix.QDConfiguration(), which knows what part of the table needs to be in the matrix). A matrix containing only numbers can easily be plotted using the regular barplot() function, as shown below.

```
c <- configuration(rksp.0)
mat <- as.matrix(c)</pre>
```

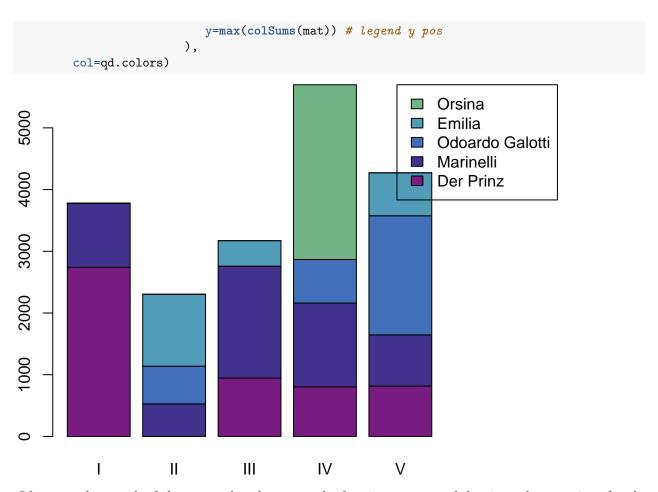




5.1.1 Filtering unimportant characters

This is informative, but doesn't look very nice and some colors are difficult to associate with characters because colors are repeating. We will therefore use the function filter(), which is very similar to format(): It can be applied to any object of the type QDHasCharacter and removes rows according to certain criteria. In this case, we filter every character except the five most talkative ones. As usual, see ?filter.QDHasCharacter to see other options.

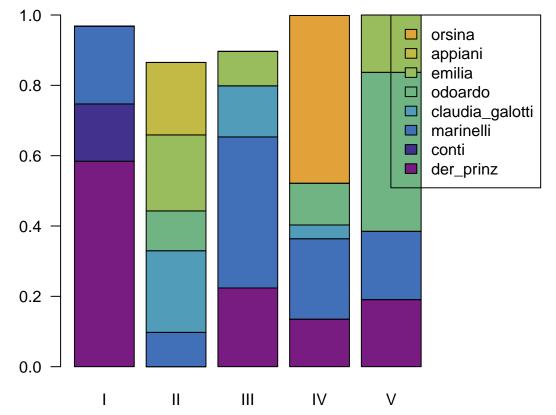
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Of course, the speech of characters that do not speak often is now removed, leaving only a portion of each act covered in the plot.

5.1.2 Normalization

Since each act has a different length, it is often useful to normalize each block, according to the total number of spoken tokens. This way, we can display the relative active presence of each character in each act. In combination with the filtering we did before, however, we need to be careful: If the scale with the filtered matrix, a certain portion of the character speech gets lost. The following snippet shows how to scale using the original matrix, but still only include the top eight characters into the matrix.



5.2 Copresence

Configuration matrices are also often used to get an overview of who is copresent on stage. This can also be achieved using the function configuration(). First, we create a configuration matrix that only represents presence or absence of a figure (and we switch to scenes). Obviously, the resulting matrix has many more columns, we include only the first 10 below.

```
configuration(rksp.0, onlyPresence = TRUE, segment="Scene")
```

```
##
     corpus drama
                         character
                                             2
                                                  3
                                                              5
                                                                    6
## 1
       test rksp.0
                         der_prinz
                                    TRUE
                                         TRUE
                                              TRUE
                                                    TRUE
                                                          TRUE
                                                                TRUE
## 2
       test rksp.0 der_kammerdiener
                                    TRUE FALSE FALSE FALSE FALSE
## 4
       test rksp.0
                             conti FALSE TRUE FALSE TRUE FALSE FALSE
## 9
       test rksp.0
                         marinelli FALSE FALSE FALSE FALSE TRUE
## 14
       test rksp.0
                       camillo_rota FALSE FALSE FALSE FALSE FALSE
## 15
       test rksp.0 claudia_galotti FALSE FALSE FALSE FALSE FALSE FALSE
                             pirro FALSE FALSE FALSE FALSE FALSE
## 16
       test rksp.0
```

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```
## 17
       test rksp.0
                           odoardo FALSE FALSE FALSE FALSE FALSE
## 20
       test rksp.0
                          angelo FALSE FALSE FALSE FALSE FALSE
                          emilia FALSE FALSE FALSE FALSE FALSE
## 25
       test rksp.0
## 27
       test rksp.0
                          appiani FALSE FALSE FALSE FALSE FALSE
## 47
       test rksp.0
                         battista FALSE FALSE FALSE FALSE FALSE
## 65
      test rksp.0
                           orsina FALSE FALSE FALSE FALSE FALSE
##
         7
      TRUE
## 1
## 2
      TRUE
## 4 FALSE
## 9 FALSE
## 14 FALSE
## 15 FALSE
## 16 FALSE
## 17 FALSE
## 20 FALSE
## 25 FALSE
## 27 FALSE
## 47 FALSE
## 65 FALSE
```

Creating a co-occurrence matrix is a simple matter of matrix multiplication, and we already know how to create a matrix.

```
# extract the configuration
c <- configuration(rksp.0, onlyPresence = TRUE, segment="Scene")

# extract a matrix
mat <- as.matrix(c)

# multiply the matrix with its inverse
# this creates the copresence matrix
copresence <- mat %*% t(mat)

# add character names
rownames(copresence) <- c$character
colnames(copresence) <- c$character</pre>
copresence
```

```
##
                     der_prinz der_kammerdiener conti marinelli camillo_rota
## der_prinz
                            17
                                               2
                                                      2
                                                                 9
## der_kammerdiener
                             2
                                                2
                                                      0
                                                                 0
                                                                               0
                             2
                                                      2
                                                                 0
## conti
                                                0
                                                                               0
## marinelli
                             9
                                                0
                                                      0
                                                                19
                                                                               0
## camillo_rota
                                                0
                                                      0
                                                                 0
                                                                               1
                             0
                                                0
                                                      0
                                                                 3
                                                                               0
## claudia_galotti
## pirro
                             0
                                                0
                                                      0
                                                                 1
                                                                               0
## odoardo
                             2
                                                      0
                                                0
                                                                 4
                                                                               0
## angelo
                                                0
                                                      0
                                                                               0
                                                                 1
## emilia
                             2
                                                0
                                                      0
                                                                 4
                                                                               Λ
## appiani
                             0
                                                0
                                                      0
                                                                 2
                                                                               0
                                                                 3
## battista
                             1
                                                Ω
                                                      0
                                                                               0
## orsina
                                                0
                                                      0
                                                                 3
##
                     claudia_galotti pirro odoardo angelo emilia appiani
```

```
2
## der_prinz
                                      0
                                             0
                                                              0
                                                                                0
## der_kammerdiener
                                      0
                                             0
                                                      0
                                                              0
                                                                       0
                                                                                0
## conti
                                      0
                                             0
                                                      0
                                                              0
                                                                       0
                                                                                0
                                      3
                                                                       4
                                                                                2
## marinelli
                                             1
                                                      4
                                                              1
## camillo_rota
                                      0
                                             0
                                                      0
                                                              0
                                                                       0
                                                                                0
## claudia_galotti
                                             3
                                                      3
                                                              0
                                                                       3
                                                                                4
                                     13
## pirro
                                             4
                                                                       0
                                      3
                                                      1
                                                              1
                                                                                1
## odoardo
                                      3
                                                                       2
                                             1
                                                     12
                                                              0
                                                                                0
## angelo
                                      0
                                             1
                                                      0
                                                              2
                                                                       0
                                                                                0
                                      3
                                             0
                                                      2
                                                                      7
## emilia
                                                              0
                                                                                1
## appiani
                                      4
                                             1
                                                      0
                                                              0
                                                                      1
                                                                                5
                                      2
                                             0
                                                      0
                                                                                0
## battista
                                                              0
                                                                      1
                                                                       0
## orsina
                                                      3
                                                              0
                                                                                0
##
                      battista orsina
## der_prinz
                              1
                                      1
## der_kammerdiener
                              0
                                      0
                              0
                                      0
## conti
## marinelli
                              3
                                      3
                              0
                                      0
## camillo_rota
## claudia_galotti
                              2
                                      1
## pirro
                              0
                                      0
## odoardo
                              0
                                      3
## angelo
                                      0
                              0
## emilia
                                      0
                              1
                                      0
## appiani
                              0
## battista
                              4
                                      0
## orsina
                              0
                                      6
```

The resulting copresence matrix shows the number of scenes in which two characters are both present. The diagonal shows the number of scenes in which a character is present in total (because each character is always copresent with itself, so to speak).

There are multiple ways to visualise this copresence. One option is a heat map, as shown below.

5.2.1 As Heatmap

The copresence can be visualised in a simple heat map. We first focus on the lower triangle and also remove the diagonal values. The actual plotting is a bit more complicated in this case, because we are just using the values in the copresence matrix as pixel intensities in the plot. Also, the axes need to be suppressed first, and can be added later with the proper names of the characters. If needed the code can also be used to include labels into the heat map.

```
c <- configuration(rksp.0, onlyPresence = TRUE, segment="Scene") %>%
  filter(rksp.0, threshold = 7) %>%
  format(rksp.0)

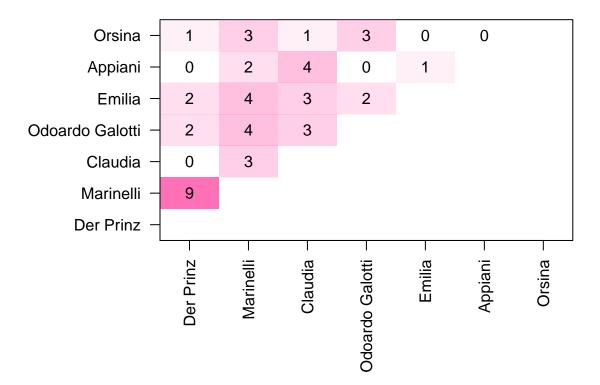
# extract a matrix
mat <- as.matrix(c)

# multiply the matrix with its inverse
# this creates the copresence matrix
copresence <- mat %*% t(mat)

# add character names
rownames(copresence) <- c$character</pre>
```

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```
colnames(copresence) <- c$character</pre>
# since it's a square matrix, we don't need the bottom right triangle
# and diagonales.
copresence[lower.tri(copresence,diag=TRUE)] <- NA</pre>
par(mar=c(10,10,1,1)) # plot margins
image(copresence,
      col = rgb(256,111,184, alpha=(seq(0,255)),
               maxColorValue = 256),
      xaxt= "n", # no x axis
      yaxt= "n", # no y axis
      frame=TRUE # print a frame around the heatmap
# include values as labels
text(y=(rep(1:ncol(copresence), each=nrow(copresence))-1)/(nrow(copresence)-1),
     x=(1:nrow(copresence)-1)/(nrow(copresence)-1),
     labels=as.vector(copresence))
# add the x axis
axis(1, at = seq(0,1,length.out = length(c$character)), labels = c$character, las=3)
# add the y axis
axis(2, at = seq(0,1,length.out = length(c$character)), labels = c$character, las=1)
```



Apparently, Marinelli and the prince have the most shared scenes. Marinelli also shares a scene with most other figures (sum of the vertical bar: 16).

5.2.2 As Network

The same information can also be visualized as a copresence network. For this, we employ the R-package igraph. A nice introduction to igraph can be found in (Arnold and Tilton, 2015), particularly for literary networks.

Technically, the matrix we created before is an adjacency matrix. It is therefore simple to convert it to a graph, and igraph offers the function graph_from_adjacency_matrix() for this.

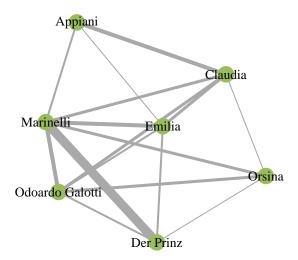
```
c <- configuration(rksp.0, onlyPresence = TRUE, segment="Scene") %>%
  filter(rksp.0, threshold = 7) %>%
  format(rksp.0)
# extract a matrix
mat <- as.matrix(c)</pre>
# multiply the matrix with its inverse
# this creates the copresence matrix
copresence <- mat %*% t(mat)</pre>
# add character names
rownames(copresence) <- c$character</pre>
colnames(copresence) <- c$character</pre>
# convert the adjacency matrix to a graph object
g <- igraph::graph_from_adjacency_matrix(copresence,
                                  weighted=TRUE,
                                                      # weighted graph
                                  mode="undirected", # no direction
                                   diag=FALSE
                                                      # no looping edges
```

Now the variable g holds the graph object. There are different things we can do with the graph. First, we can visualise it. igraph uses the same mechanism of R that we have used before for plotting, specifying a plot() function that can plot graph objects.

```
# Now we plot
plot(g,
    layout=layout_with_gem,  # how to lay out the graph
    main="Copresence Network: Emilia Galotti", # title
    vertex.label.cex=0.8,  # label size
    vertex.label.color="black", # font color
    vertex.color=qd.colors[6], # vertex color
    vertex.frame.color=NA,  # no vertex border
    edge.width=E(g)$weight # scale edges according to their weight
)
```

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Copresence Network: Emilia Galotti



5.2.3 Network properties

igraph offers a number of function to extract graph properties. We will show them below without explanation. The documentation of each function can be found here.

5.2.3.1 Density

Function: graph.density()

```
graph.density(g)
```

[1] 0.7619048

5.2.3.2 Average nearest neighbor degree

Function: knn()

```
knn(g)
```

```
## $knn
                          Marinelli
                                            Claudia Odoardo Galotti
##
         Der Prinz
          4.214286
                           2.760000
                                            4.714286
                                                            5.214286
##
##
            Emilia
                            Appiani
                                              Orsina
##
          6.166667
                           7.285714
                                            8.375000
##
## $knnk
## [1]
            NaN
                     NaN 7.285714 6.294643 5.365079 2.760000
```

5.2.3.3 Edge connectivity

Function: edge_connectivity()

```
edge_connectivity(g)
```

[1] 3

5.2.4 Graph Export

As a final step, one might want to further work on the graph using Gephi, or other tools. In order to do so, one can export the graph into an appropriate file:

This results in a file called rksp.O.graphml, that starts similarly as this:

This file can be opened with Gephi.

Chapter 6

Character Exchange

Character exchange – entering and leaving the stage – often takes place at scene boundaries. We offer multiple ways to measure and visualize this exchange over the course of the play.

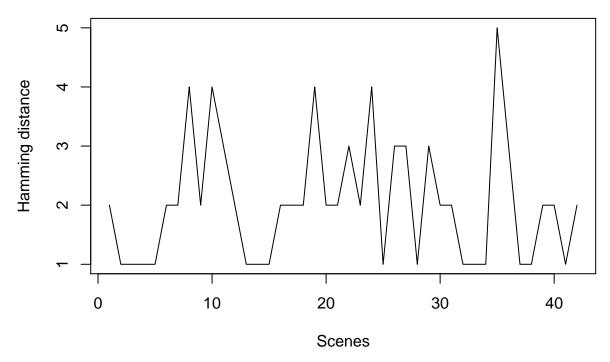
6.1 Hamming Distance

The hamming distance has been introduced in information theory by Richard Hamming (Hamming, 1950). Intuitively, the hamming distance expresses the number of characters that either enter or leave the stage. Its application to a dramatic text is straightforward, using the function hamming(). By default, the hamming distance is calculated over scene boundaries.

```
hamming(rksp.0, variant="Hamming")
```

```
## [1] 2 1 1 1 1 2 2 4 2 4 3 2 1 1 1 2 2 2 4 2 2 3 2 4 1 3 3 1 3 2 2 1 1 1 5 ## [36] 3 1 1 2 2 1 2
```

The return value of the function is just a vector of numbers, one less than the number of scenes in the play. Of course, this can be plotted:



Hamming distance, in its original definition as well as the Hamming variant in our function, returns absolute values. As usual, these values should be normalized, and the function hamming() provides two ways of doing that.

6.1.1 NormalizedHamming

Straightforwardly, one can normalise by the overall number of characters in the play. This value is only maximal, if at once scene boundary, one subset of characters enters the stage, and the entire other subset leaves the stage. As we can see below, this is not the case in our demo text.

```
hamming(rksp.0, variant="NormalizedHamming")
```

```
## [1] 0.15384615 0.07692308 0.07692308 0.07692308 0.07692308 0.15384615

## [7] 0.15384615 0.30769231 0.15384615 0.30769231 0.23076923 0.15384615

## [13] 0.07692308 0.07692308 0.07692308 0.15384615 0.15384615 0.15384615

## [19] 0.30769231 0.15384615 0.15384615 0.23076923 0.15384615 0.30769231

## [25] 0.07692308 0.23076923 0.23076923 0.07692308 0.23076923 0.15384615

## [31] 0.15384615 0.07692308 0.07692308 0.07692308 0.38461538 0.23076923

## [37] 0.07692308 0.07692308 0.15384615 0.07692308 0.15384615
```

6.1.2 Trilcke

One more possible normalization has been proposed by Trilcke et al. (2017). Instead of normalizing with all characters in the play, the Trilcke variant only normalizes with the characters in the two adjacent scenes. Thus, if all characters leave the stage, and a new set of characters enter it, the distance is maximal. This however does not have to include all characters in the play.

```
hamming(rksp.0, variant="Trilcke")
```

```
## [1] 0.6666667 0.5000000 0.5000000 0.5000000 0.5000000 0.6666667 0.66666667 
## [8] 1.0000000 0.6666667 1.0000000 0.7500000 0.6666667 0.5000000 0.3333333 
## [15] 0.3333333 0.5000000 0.5000000 0.6666667 1.0000000 0.6666667 0.6666667 
## [22] 0.7500000 0.5000000 0.8000000 0.3333333 0.7500000 0.7500000 0.3333333 
## [29] 0.7500000 0.6666667 0.6666667 0.3333333 0.3333333 0.3333333 1.0000000
```

[36] 1.0000000 0.5000000 0.5000000 0.6666667 0.6666667 0.5000000 0.5000000 The Trilcke variant is the default setting of the hamming()-function.

Text Collections

So far, we have mostly processed single texts. In this chapter, we will cover the important things to know when processing collections of texts – be it genres or the works of an author.

The following code can only be reproduced if the GerDraCor data set and the collection repository have been installed. This can be done with the following code (assuming an internet connection).

```
installData("gdc")

## Pulling new data from https://github.com/quadrama/data_gdc.git.

## Already up-to-date
installCollectionData()

## Warning in utils::unzip(tf, exdir = dataDirectory): Fehler 1 während des

## Extrahierens aus Zipfile

## Warning in file.rename(from = file.path(repoDirectory, "collections"), to =

## getOption("qd.collectionDirectory")): kann Datei '/Users/reiterns/QuaDramA/

## Data2/metadata-master/collections' nicht in '/Users/reiterns/QuaDramA/

## Data2/collections' umbenennen. Grund 'No such file or directory'
```

7.1 Defining collections

Before processing, it's necessary to define a collection of texts, by assembling their ids in a list. These are considered to be sets of plays without internal structure (e.g., no play is marked as prototypical).

```
# load ids that are comedies and tragedies
ids.tragedy <- unique(loadSet(c("trauerspiel", "tragödie", "tragoedie")))
ids.comedy <- unique(loadSet(c("lustspiel", "komödie", "komoedie")))

# make sure we're using the GerDraCor corpus
ids.tragedy <- sub("tg:", "gdc:", ids.tragedy, fixed = TRUE)
ids.comedy <- sub("tg:", "gdc:", ids.comedy, fixed = TRUE)</pre>
```

7.2 Loading multiple plays

Loading multiple plays at once is straightforward, and not different from loading a single one.

Part II Text Analysis

Word Field Analysis

What we consider a word field here may differ from specific uses in linguistics. In this context, a word field is a list of words that belong to a common theme / topic / semantic group. Multiple word fields can be assembled to create a dictionary. On a technical level, what we describe in the following works for arbitrary lists of words. A semantic relation between the words is technically not required. Thus, the following pieces of code can be used with arbitrary word lists.

For **demo purposes** (this is really a toy example), we will define the word field of Love as containing the words Liebe (love) and Herz (heart). In R, we can put them in a character vector, and we use lemmas:

```
wf_love <- c("liebe", "herz")</pre>
```

We will test this word field on Emilia Galotti, which should be about love.

```
data("rksp.0")
```

8.1 Single word field

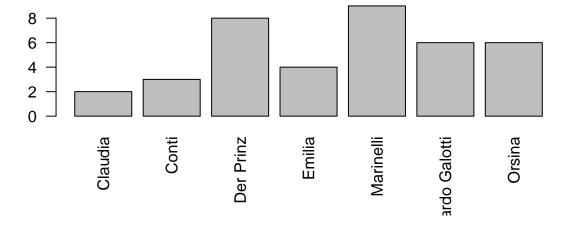
The core of the word field analysis is collecting statistics about a dictionary. Therefore, we use the function called dictionaryStatisticsSingle() (single, because we only want to analyse a single word field):

```
dictionaryStatisticsSingle(
  rksp.0,  # the text we want to process
  wordfield=wf_love  # the word field
)
```

```
##
      corpus drama
                            character x
## 1
        test rksp.0
                               angelo 0
## 2
        test rksp.0
                              appiani 0
## 3
        test rksp.0
                             battista 0
## 4
        test rksp.0
                         camillo_rota 0
## 5
        test rksp.0
                      claudia_galotti 2
##
        test rksp.0
                                 conti 3
## 7
        test rksp.0 der_kammerdiener 0
## 8
        test rksp.0
                            der_prinz 8
## 9
        test rksp.0
                               emilia 4
        test rksp.0
                            marinelli 9
## 11
        test rksp.0
                              odoardo 6
        test rksp.0
                               orsina 6
## 12
## 13
        test rksp.0
                                pirro 0
```

What this table shows us the number of times the characters in the play use words that appear in this list. By default, these are absolute numbers.

We can visualise these counts in a simple bar chart:



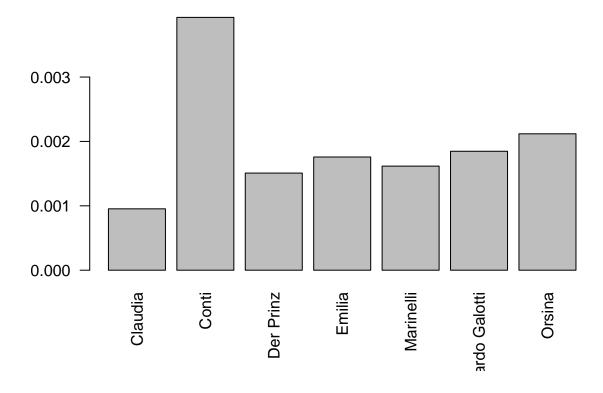
Apparently, the prince and Marinelli are mentioning these words a lot more than the other characters.

When comparing characters, it often (but not always) makes sense to normalize the counts according to the total number of spoken words by a character. This can be enabled by setting the argument normalizeByFigure=TRUE. This will divide the number of words in this field by the total number of words a character speaks.

```
dstat <- dictionaryStatisticsSingle(rksp.0, wordfield=wf_love,
    normalizeByFigure = TRUE  # apply normalization
) %>% format(rksp.0)  # reformat character names

# remove figures not using these words at all
dstat <- dstat[dstat$x>0,]

barplot(dstat$x,
    names.arg = dstat$character,
    las=2
)
```



8.2 Multiple Word Fields

The function dictionaryStatistics() can be used to analyse multiple dictionaries at once. To this end, dictionaries are represented as lists of character vectors. The (named) outer list contains the keywords, the vectors are just words associated with the keyword.

New dictionaries can be easily created like this:

```
wf <- list(Liebe=c("liebe", "herz", "schmerz"), Hass=c("hass", "hassen"))</pre>
```

This dictionary contains the two entries Liebe (love) and Hass (hate), with 3 respective 2 entries. Dictionaries can be created in code, like shown above. In addition, the function loadFields() can be used to download dictionary content from a URL or a directory. By default, the function loads this dictionary from GitHub (that we used in publications), for the keywords Liebe and Familie (family). Since version 2.3.0, this dictionary is included in the package as base_dictionary and can be used right away (without internet connection). It is also the default dictionary used by dictionaryStatistics().

The function loadFields() offers parameters to load from different URLs via http or to load from plain text files that are stored locally. The latter can be achieved by specifying the directory as baseurl. Entries for each keyword should then be stored in a file named like the keyword, and ending with txt (by default, can be overwritten). See ?loadFields for details. Some of the options can also be specified through dictionaryStatistics(), as exemplified below.

The following examples use the base_dictionary, i.e., a specific version of the fields we have been using in QuaDramA.

```
dstat <- dictionaryStatistics(
  rksp.0, # the text
  fieldnames = # fields we want to measure (from base_dictionary)
    c("Liebe", "Familie", "Ratio", "Krieg", "Religion"),
  normalizeByFigure = TRUE, # normalization by figure
  normalizeByField = TRUE # normalization by field</pre>
```

```
)
dstat
```

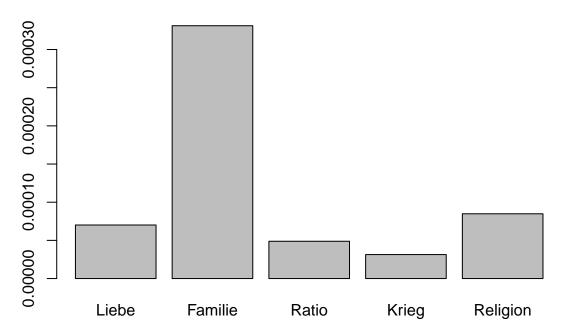
```
Liebe
                                                        Familie
##
      corpus
             drama
                           character
                                                                        Ratio
##
        test rksp.0
                              angelo 5.025967e-05 4.314529e-05 7.290755e-05
  1
##
  2
        test rksp.0
                              appiani 8.610133e-05 1.355080e-04 6.661338e-05
## 3
        test rksp.0
                            battista 0.000000e+00 1.404988e-04 0.000000e+00
## 4
        test rksp.0
                        camillo_rota 1.003613e-04 1.292324e-04 4.367575e-04
                     claudia_galotti 6.591891e-05 2.350575e-04 3.972037e-05
## 5
        test rksp.0
## 6
        test rksp.0
                                conti 1.531692e-04 5.379043e-05 3.635835e-05
## 7
        test rksp.0 der_kammerdiener 0.000000e+00 0.000000e+00 0.000000e+00
## 8
        test rksp.0
                           der_prinz 7.021317e-05 6.974600e-05 4.714313e-05
## 9
        test rksp.0
                               emilia 7.014262e-05 3.311757e-04 4.884005e-05
## 10
        test rksp.0
                           marinelli 8.026020e-05 1.451804e-04 3.991777e-05
## 11
        test rksp.0
                              odoardo 5.895608e-05 2.319657e-04 6.271666e-05
                              orsina 5.259046e-05 8.223048e-05 4.250366e-05
## 12
        test rksp.0
## 13
        test rksp.0
                               pirro 0.000000e+00 1.597508e-04 2.699492e-05
##
             Krieg
                       Religion
  1
      5.624297e-05 0.000000e+00
      4.817575e-05 0.000000e+00
      0.000000e+00 0.000000e+00
     8.423181e-05 0.000000e+00
## 4
      4.681329e-05 5.853528e-05
## 6
      2.337322e-05 0.000000e+00
## 7
      0.000000e+00 0.000000e+00
      2.862261e-05 3.308290e-05
      3.139717e-05 8.482745e-05
## 10 3.528446e-05 1.890842e-05
## 11 4.123417e-05 4.861291e-05
## 12 2.522195e-05 1.238973e-05
## 13 0.000000e+00 1.022966e-04
```

The variable dstat now contains multiple columns, one for each word field. We have been using the option normalizeByFigure before. When comparing multiple fields, it often happens that they have a different size (i.e., different number of words). In this case, it makes sense to also normalize with the number of words in the word field. This is achieved by normalizeByField=TRUE. This makes the numbers very small, but they should be used in comparison anyway.

8.2.1 Bar plot by character

It is now straightforward to show the distribution of fields for a single character:

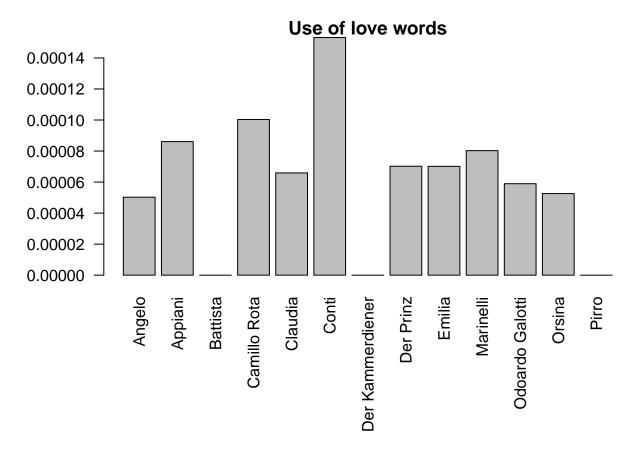
Emilia's speech



8.2.2 Bar plot by field

Conversely, we can also show who uses words of a certain field how often:

```
dstat <- dictionaryStatistics(</pre>
  rksp.0, # the text
  fieldnames = # fields we want to measure (from base_dictionary)
    c("Liebe", "Familie", "Ratio", "Krieg", "Religion"),
  normalizeByFigure = TRUE, # normalization by figure
  normalizeByField = TRUE
                             # normalization by field
) %>%
  format(rksp.0)
mat <- as.matrix(dstat)</pre>
par(mar=c(9,4,1,1))
barplot(mat[,1],
                                  # Select the row for 'love'
        main="Use of love words", # title for plot
        beside = TRUE,
                                  # not stacked
        names.arg = dstat$character, # x axis labels
                                  # rotation for labels
```



8.3 Dictionary Based Distance

camillo_rota

conti

der_prinz
emilia

marinelli

claudia_galotti

Technically, the output of dictionaryStatistics() is a data.frame. This is suitable for most uses. In some cases, however, it's more suited to work with a matrix that only contains the raw numbers (i.e., number of family words). Calculating character distance based on dictionaries, for instance. For these cases, the package provides an S3 method that extracts the numeric part of the data.frame and creates a matrix. We have used this function as.matrix() already above.

The matrix doesn't have row names by default. The snippet below can be used to add row names.

```
ds <- dictionaryStatistics(rksp.0,</pre>
                             fieldnames=c("Liebe", "Familie", "Ratio", "Krieg", "Religion"),
                             normalizeByFigure=TRUE)
m <- as.matrix(ds)
rownames(m) <- ds$character</pre>
##
                           Liebe
                                      Familie
                                                    Ratio
                                                                 Krieg
## angelo
                     0.004724409 0.003149606 0.007874016 0.006299213
                     0.008093525 0.009892086 0.007194245 0.005395683
## appiani
## battista
                     0.00000000 0.010256410 0.00000000 0.000000000
```

0.009433962 0.009433962 0.047169811 0.009433962

0.006196378 0.017159199 0.004289800 0.005243089

0.014397906 0.003926702 0.003926702 0.002617801

0.006600038 0.005091458 0.005091458 0.003205733

0.006593407 0.024175824 0.005274725 0.003516484 0.007544458 0.010598168 0.004311119 0.003951859

```
## odoardo
                   0.005541872 0.016933498 0.006773399 0.004618227
                   0.004943503 0.006002825 0.004590395 0.002824859
## orsina
## pirro
                   0.00000000 0.011661808 0.002915452 0.000000000
##
                       Religion
## angelo
                   0.000000000
                   0.000000000
## appiani
                   0.000000000
## battista
## camillo_rota
                   0.000000000
## claudia_galotti 0.0033365110
## conti
                   0.000000000
## der_kammerdiener 0.0000000000
## der_prinz
                  0.0018857251
## emilia
                   0.0048351648
                   0.0010777798
## marinelli
## odoardo
                   0.0027709360
## orsina
                   0.0007062147
                   0.0058309038
## pirro
```

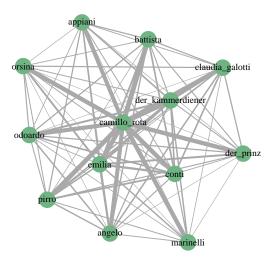
Every character is now represented with five numbers, which can be interpreted as a vector in five-dimensional space. This means, we can easily apply distance metrics supplied by the function dist() (from the default package stats). By default, dist() calculates Euclidean distance.

```
cdist <- dist(m)
# output not shown</pre>
```

The resulting data structure is similar to the one in the weighted configuration matrix, which means everything from Section 5.2.2 can be applied here. In particular, we can convert these distances into a network:

This network can of course be visualised again.

Dictionary distance network



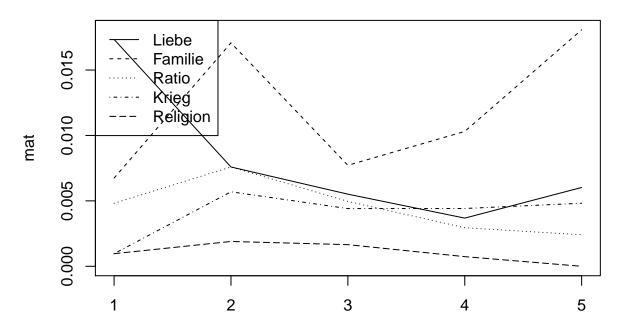
Although this network is similar to the one shown in Section 5.2.2 (both undirected and weighted), it displays a totally different kind of information. The networks based on copresence connect characters that appear together on stage, while this network connects characters with similar thematic profile in their speech (within the limits of being able to detect thematic profiles via word fields).

8.4 Development over the course of the play

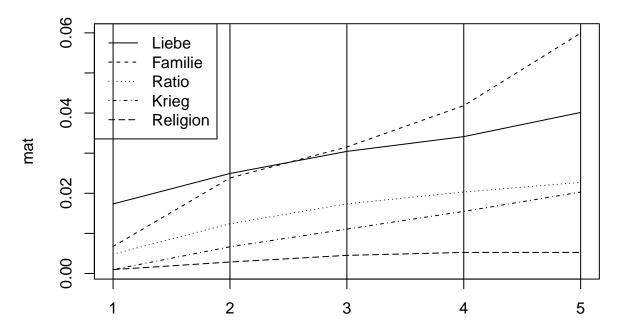
Finally, the function dictionaryStatistics() can be used to track word field for segments of the play. To do that, we use the parameter segment, and set it to either "Act" or "Scene".

8.4.1 Individual characters

We can now plot the theme progress over the course of the play. This can be done for specific characters, as shown below.



Depending on the use case, it might be easier to interpret if the numbers are cumulatively added up. The snippet below shows how this works.

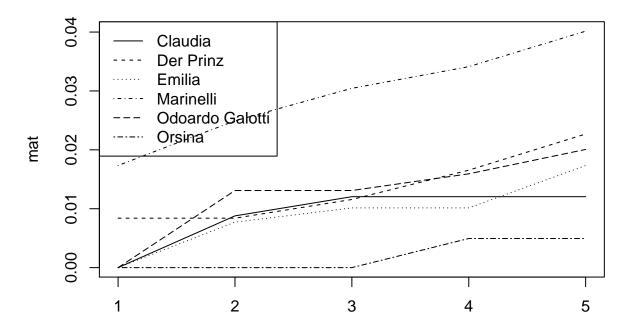


8.4.2 Comparing characters

Simultaneously to the setting above, we now want to compare the development of several characters for a single word field. This unfortunately requires some reshuffling of the data, using the function **reshape** (from the **stats** package).

```
dsl <- dictionaryStatistics(rksp.0,</pre>
                             fieldnames=c("Liebe"),
                             normalizeByFigure=TRUE,
                             segment="Act") %>%
  filter(rksp.0,
         threshold = 6) %>%
  format(rksp.0)
dsl <- reshape(dsl, direction = "wide", # the table becomes wider
               timevar = "Number.Act", # the column that specifies multiple readings
               times = "Liebe",
                                          # the number to distribute
               idvar=c("corpus","drama","character") # what identifies a character
)
mat <- as.matrix.data.frame(dsl[,4:ncol(dsl)])</pre>
rownames(mat) <- dsl$character</pre>
mat <- apply(mat,1,cumsum)</pre>
matplot(mat, type="1", lty = 1:ncol(mat), col="black", main="Liebe")
legend(x="topleft", legend=colnames(mat), lty=1:ncol(mat))
```

Liebe



Advanced Text Analysis

9.1 When are characters mentioned?

When characters are speaking on stage, they are actively present. But they can also be passively present, if other characters refer to them. Both levels of presence can be extracted with the presence() function:

```
# Load Emilia Galotti
data(rksp.0)

pres <- presence(rksp.0)
pres</pre>
```

##		corpus	drama	character	scenes	actives	passives	presence
##	1	test	rksp.0	angelo	43	2	1	0.02325581
##	2	test	rksp.0	appiani	43	5	14	-0.20930233
##	3	test	rksp.0	battista	43	4	6	-0.04651163
##	4	test	rksp.0	camillo_rota	43	1	2	-0.02325581
##	5	test	rksp.0	claudia_galotti	43	13	14	-0.02325581
##	6	test	rksp.0	conti	43	1	3	-0.04651163
##	7	test	rksp.0	${\tt der_kammerdiener}$	43	1	2	-0.02325581
##	8	test	rksp.0	der_prinz	43	14	21	-0.16279070
##	9	test	rksp.0	emilia	43	7	21	-0.32558140
##	10	test	rksp.0	marinelli	43	19	21	-0.04651163
##	11	test	rksp.0	odoardo	43	12	11	0.02325581
##	12	test	rksp.0	orsina	43	6	10	-0.09302326
##	13	test	rksp.0	pirro	43	4	5	-0.02325581

As we can see, each character has a few numbers associated: The column actives shows the number of scenes in which the character is actively present. This is equivalent to the information in the configuration matrix. The column passives shows the number of scenes in which a character is mentioned. By default, this excludes the scenes in which they are present themselves (this behaviour can be changed by adding the parameter passiveOnlyWhenNotActive = TRUE to the call of the presence function).

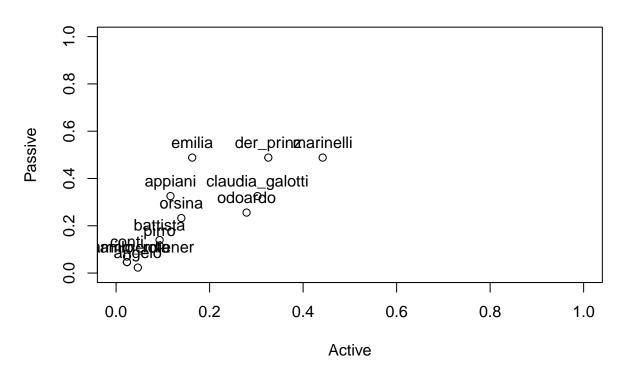
A simple visualisation that shows the characters active and passive presence in one plot can be generated like this: The first line (plot()) is responsible for the plotting of the symbols, the second line (text()) adds the character names or ids numbers.

```
plot(x=pres$active/pres$scenes,
    y=pres$passive/pres$scenes,
    xlim=c(0,1),
    ylim=c(0,1),
```

```
xlab="Active",
  ylab="Passive",
  main="Character Presence")

text(x=pres$actives/pres$scenes,
  y=pres$passives/pres$scenes,
  labels=substr(pres$character,0,20),
  pos=3)
```

Character Presence



Resterampe

10.0.1 Character meta data

We will now combine this information with additional meta data about characters, i.e., gender.



10.0.2 Character groups

Next, we want to make the same analysis not for individual characters, but for character groups, based on categories such as gender.

```
by.y = c("corpus","drama", "figure_id"))
par(mar=c(2,2,2,2))
boxplot(utteranceLength ~ Gender, # what do we want to correlate
        data=ustat,
        las = 1 # rotate axis labels
      )
                          0
                                                              0
.00
                                                              0
                         8
                          0
                                                              00 00 00
00
                          :00 -
00
 0
                     FEMALE
                                                            MALE
```

According to this picture, female characters speak slightly longer utterances in this play.

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