

Online learning analytics on social networking sites: how to tap the potential of data mining in research of educational technology

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Robert Maribe Branch

Questions to Answer by Text Mining in Education

- What algorithm can score essays as teachers do?



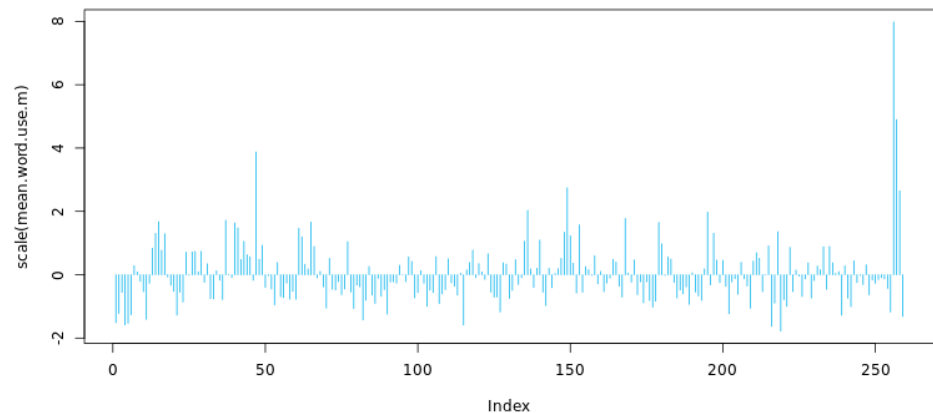
Questions to Answer by Text Mining in Education

- What courses should we recommend students' based on their course reviews and engagement levels of their enrolled courses?



Questions to Answer by Text Mining in Education

- Does the treatment improve students' lexical variety in their writing?

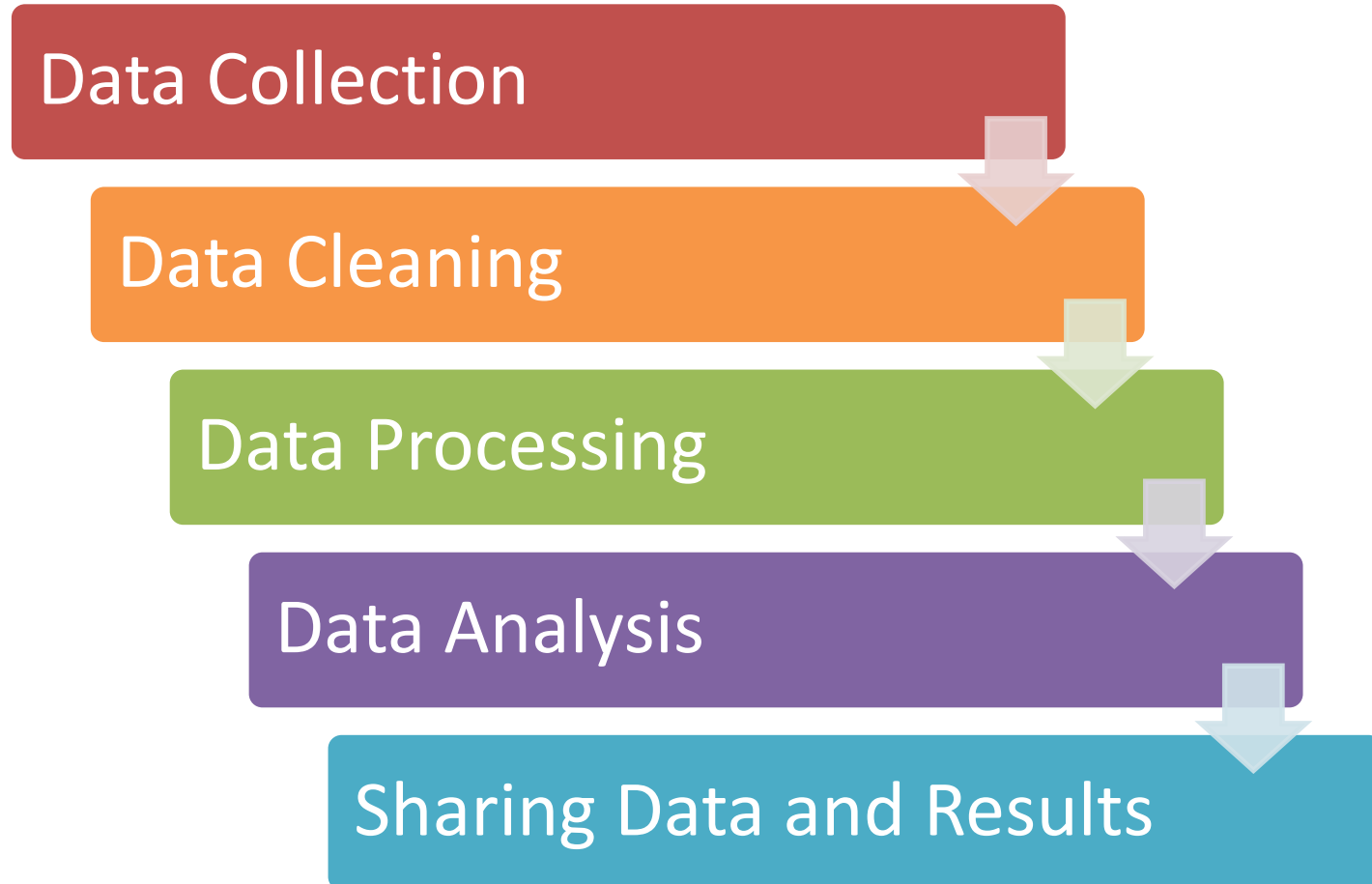


Questions to Answer by Text Mining in Education

- Are there different patterns in students' discussions; if so, are the patterns related to their academic performance?



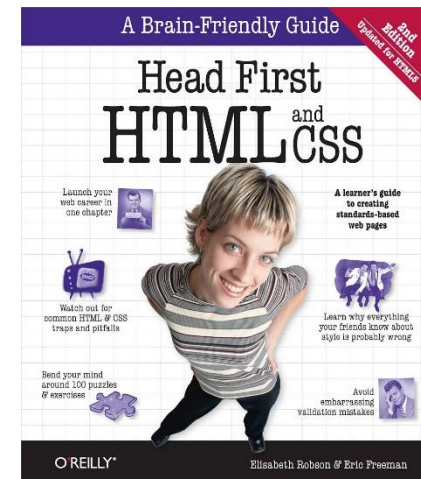
Research Pipeline



Data Collection

Scrapping data form static web pages:

- 1. A good understanding of HTML & CSS**
- 2. A good understanding of XML & JSON**



Data Collection

- XML

- `<change-log type="array">`
 - `<change-log>`
 - `<when type="datetime">2015-05-26T17:42:37Z</when>`
 - `<data>ia5m23j5hbx5ms</data>`
 - `<type>create</type>`
 - `<anon>no</anon>`
 - `<uid>gd6v7134AUa</uid>`
 - `</change-log>`
 - `</change-log>`
 - `<folders type="array"/>`
 - `<children type="array"/>`
 - `<no_answer_followup>0</no_answer_followup>`

...

Data Collection

- XML

```
- <change-log type="array">
  - <change-log>
    <when type="datetime">2015-05-26T17:42:37Z</when>
    <data>ia5m23j5hbx5ms</data>
    <type>create</type>
    <anon>no</anon>
    <uid>gd6v7134AUa</uid>
  </change-log>
</change-log>
<folders type="array"/>
<children type="array"/>
<no_answer_followup>0</no_answer_followup>
...

```

Data Collection

- JSON

```
{
  hey: "guy",
  anumber: 243,
  - anobject: {
    whoa: "nuts",
    - anarray: [
      1,
      2,
      "thr<h1>ee"
    ],
    more: "stuff"
  },
  awesome: true,
  bogus: false,
  meaning: null,
  japanese: "明日がある。",
  link: http://jsonview.com,
  notLink: "http://jsonview.com is great"
}
```

Data Collection

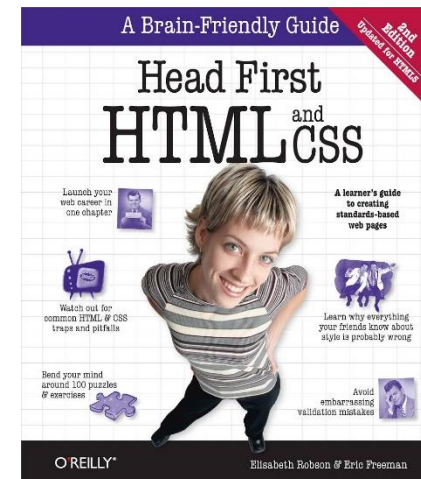
- JSON

```
{
  hey: "guy",
  anumber: 243,
  - anobject: {
    whoa: "nuts",
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      1,
      2,
      "thr<h1>ee"
    ],
    more: "stuff"
  },
  awesome: true,
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  meaning: null,
  japanese: "明日がある。",
  link: http://jsonview.com,
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}
```

Data Collection

Scrapping data form static web pages:

- 1. A good understanding of HTML & CSS**
- 2. A good understanding of XML & JSON**



Data Collection

Scrapping data form static web pages:

- 1. A good understanding of HTML & CSS**
- 2. A good understanding of XML & JSON**
- 3. Familiar with Development Tools of Browsers**



Chrome DevTools

The Chrome DevTools are a set of web authoring and debugging tools built into Google Chrome. Use the DevTools to iterate, debug and profile your site.

Chrome Canary always has the latest DevTools.

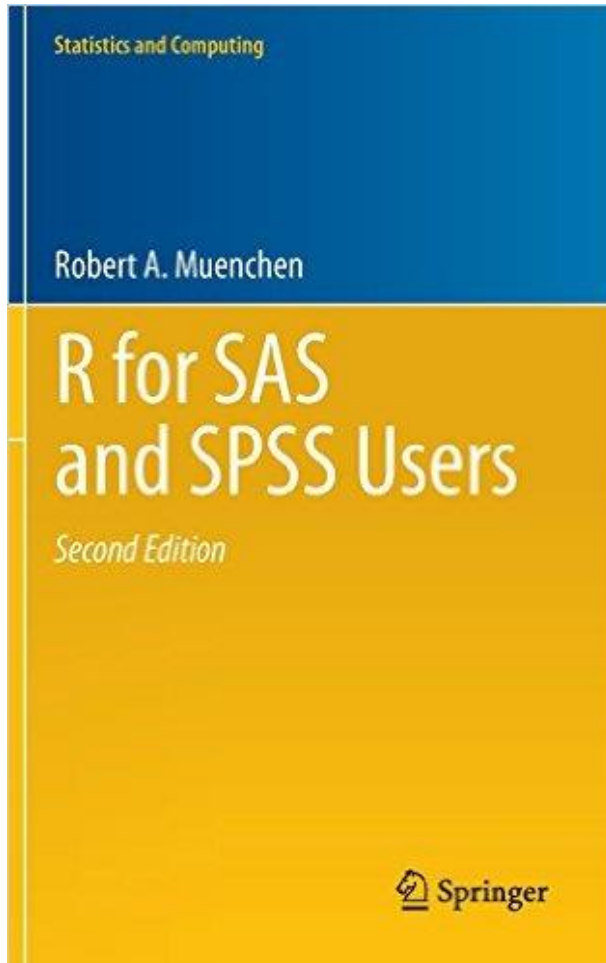
- Select **More Tools > Developer Tools** from the Chrome Menu.
- Right-click on a page element and select Inspect
- Use `Ctrl/Cmd + Shift + I` ([more shortcuts](#))

Data Collection

Scrapping data form static web pages:

- 1. A good understanding of HTML & CSS**
- 2. A good understanding of XML & JSON**
- 3. Familiar with Development Tools of Browsers**
- 4. Familiar with R and package “XML”**

Data Collection

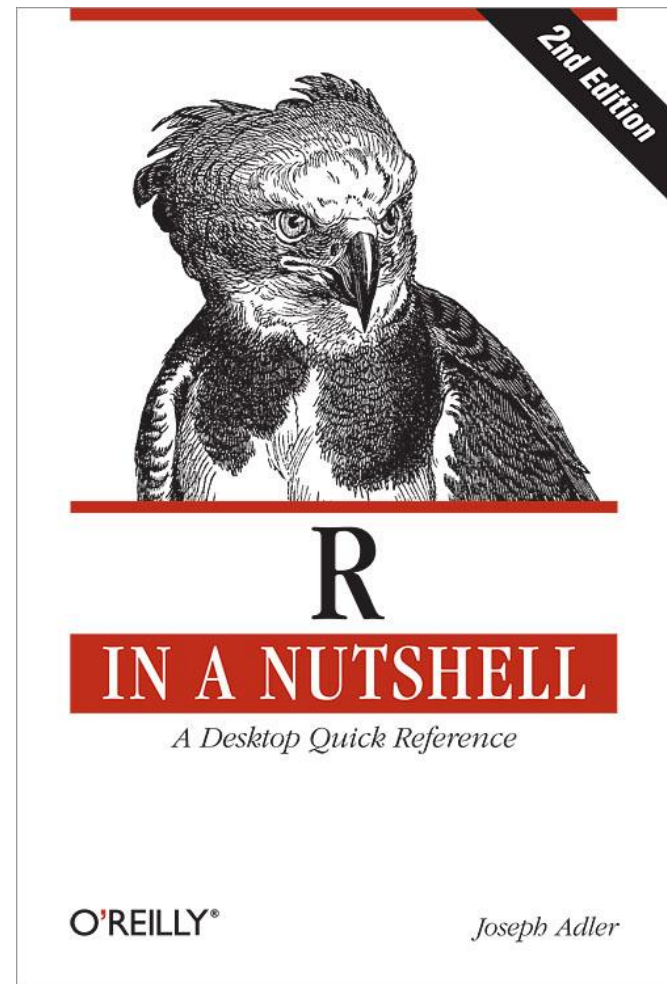


R for SAS and SPSS Users

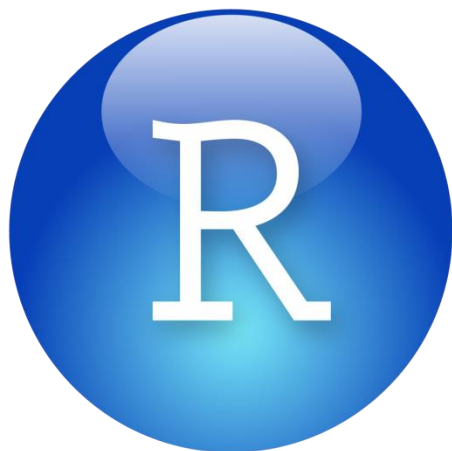
**Google *r xml package*
*filetype:pdf***

Data Collection

R in a Nutshell



Data Collection



```
getwd()
```

```
setwd("C:/Users/John/Anal  
ysis")
```

```
setwd("/home/Analysis")
```

```
setwd("XXX/TwitterHashtag  
R/data")
```

Data Collection



```
install.packages("XML")  
library(XML)
```

Data Collection



```
a <- 3
```

```
b <- c(1, 3, 7, 8)
```

```
c <- "Hello"
```

```
d <- c("piggy1", "piggy2",  
      "piggy3")
```

Data Collection



Function

```
tweetCollectByUser <-  
  function(username, numberOfTweets,  
            nameOfFile) {  
    .....  
  }
```

```
tweetCollectByUser( "aect", 300,  
  "tweetsOfAect" )
```

Data Collection



Authentication

1. Register your own app
2. Keep your consumer keys and secrets
3. Go to *Data Collection/Authentication.R*
4. Replace consumer keys and secrets with yours
5. Run lines 1-42

Data Collection



Collect User Info

1. Go to *Data Collection/collectUsers.R*
2. Run lines 1-33
3. **Practice:** Find 5 twitter accounts that you would like to collect information about, and collect their basic information in a .csv file

Data Collection



Collect tweets of particular users

1. Go to *Data Collection/getTweetsByUser.R*
2. Run lines 1-24

Data Collection



Collect tweets of particular users

1. Go to *Data Collection/getTweetsByAllUser.R*
2. Run lines 1-68
3. **Practice:** Get tweets from 2 different twitter accounts

Data Collection



Collect tweets by Hashtag

1. Go to *Data Collection/hashtagSearch.R*
2. Run lines 1-22
3. **Practice:** Get tweets with one hashtag you like

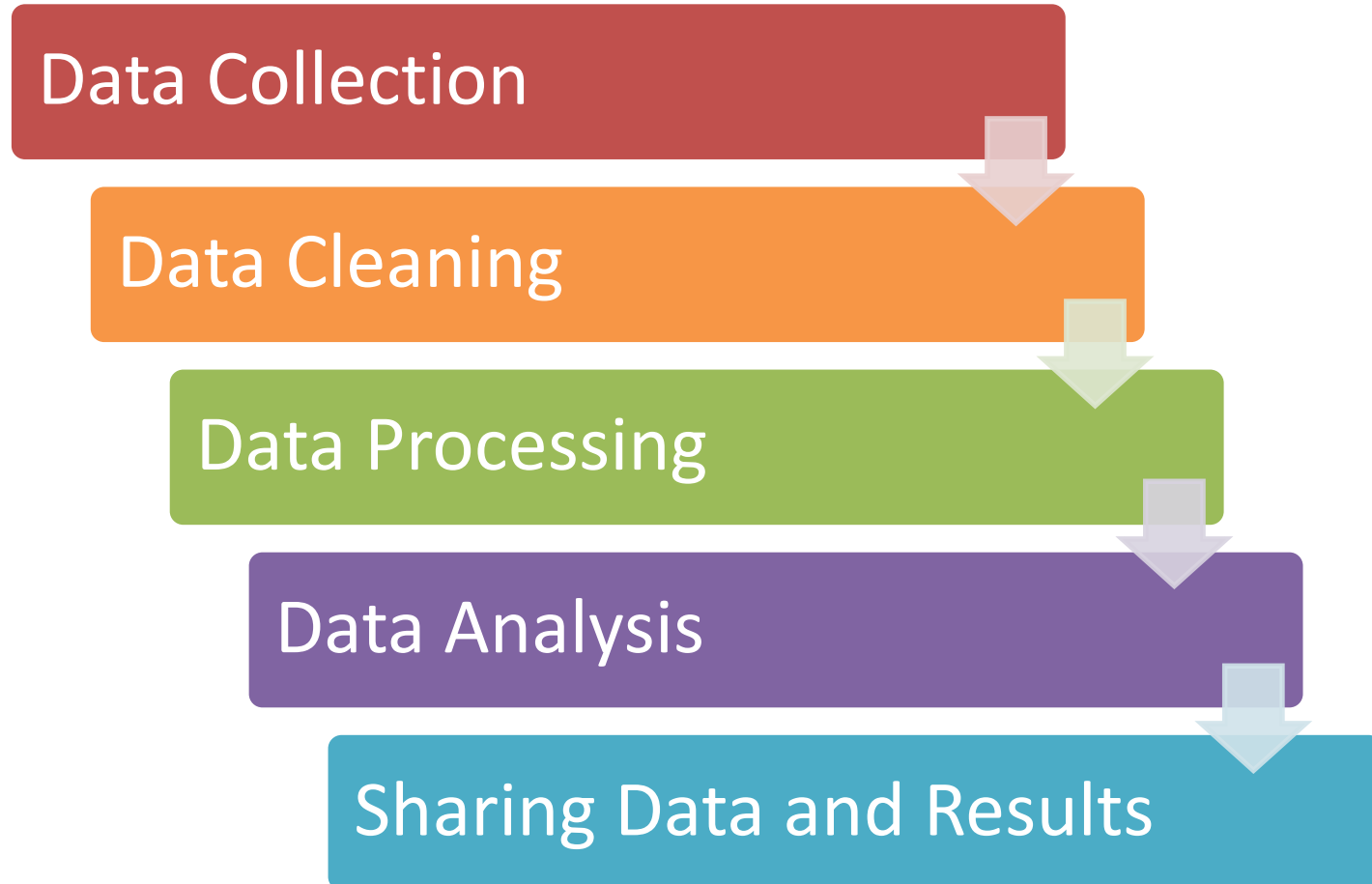
Data Collection



Collect tweets by Web Scrapping

1. Go to *Data Collection/parse_Tweets.R*
2. Run lines 1-34, 76-77
3. **Practice:** Do one web scrapping yourself
 1. *Search a hashtag using Twitter; keep scrolling down until you have all or enough number of tweets*
 2. *Download the HTML page*
 3. *.....*

Research Pipeline



Data Cleaning

	text	favorite	favoriteC	replyToSM	created	truncated	replyToSlid	replyToUI	statusSou
1	@mesterman @Ed	FALSE	0	mestermar	2015/4/15 23:52	FALSE	5.88E+17	5.88E+17	14906194 <a href=
2	#monopolistic	FALSE	0	NA	2015/4/15 23:44	FALSE	NA	5.88E+17	NA <a href=
3	RT @heosat: Ar	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
4	RT @heosat: Ar	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
5	RT @heosat: Ar	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
6	Another new re	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
7	#Teachers shou	FALSE	0	NA	2015/4/15 23:01	FALSE	NA	5.88E+17	NA <a href=
8	RT @CirrusAsse	FALSE	0	NA	2015/4/15 22:44	FALSE	NA	5.88E+17	NA <a href=
9	Teachers: get	FALSE	0	NA	2015/4/15 22:32	FALSE	NA	5.88E+17	NA <a href=
10	How 2 Put Meta	FALSE	0	NA	2015/4/15 22:02	FALSE	NA	5.88E+17	NA <a href=
11	RT @CanvasPenn	FALSE	0	NA	2015/4/15 21:11	FALSE	NA	5.88E+17	NA <a href=
12	Great tool for	FALSE	0	NA	2015/4/15 20:38	FALSE	NA	5.88E+17	NA <a href=
13	Be the change	FALSE	0	NA	2015/4/15 20:23	FALSE	NA	5.88E+17	NA <a href=
14	DYSLEXIC WHO,,	FALSE	0	NA	2015/4/15 20:02	FALSE	NA	5.88E+17	NA <a href=
15	7 Cyberlearnin	FALSE	0	NA	2015/4/15 20:01	FALSE	NA	5.88E+17	NA <a href=
16	RT @grahamlfox	FALSE	0	NA	2015/4/15 19:54	FALSE	NA	5.88E+17	NA <a href=
17	RT @Spencer_GG	FALSE	0	NA	2015/4/15 19:47	FALSE	NA	5.88E+17	NA <a href=
18	RT @bsarte: #M	FALSE	0	NA	2015/4/15 19:45	FALSE	NA	5.88E+17	NA <a href=
19	#GoogleClassro	FALSE	2	NA	2015/4/15 19:43	FALSE	NA	5.88E+17	NA <a href=
20	#MDM: Mobile c	FALSE	1	NA	2015/4/15 19:35	FALSE	NA	5.88E+17	NA <a href=
21	bsarte: #MDM:	FALSE	1	NA	2015/4/15 19:32	FALSE	NA	5.88E+17	NA <a href=
22	#MDM: Mobile c	FALSE	1	NA	2015/4/15 19:31	FALSE	NA	5.88E+17	NA <a href=
23	#MDM: Mobile c	FALSE	1	NA	2015/4/15 19:25	FALSE	NA	5.88E+17	NA <a href=
24	#MDM: Mobile c	FALSE	1	NA	2015/4/15 19:20	FALSE	NA	5.88E+17	NA <a href=
25	El impacto de	FALSE	0	NA	2015/4/15 19:13	FALSE	NA	5.88E+17	NA <a href=

Data Cleaning

	text	favorite	favorite	replyToSM	created	truncated	replyToSlid	replyToUI	statusSou
1	@mesterman @Ed	FALSE	0	mesterman	2015/4/15 23:52	FALSE	5.88E+17	5.88E+17	14906194 <a href=
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4	RT @heosat: Ar	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
5	RT @heosat: Ar	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
6	Another new re	FALSE	0	NA	2015/4/15 23:35	FALSE	NA	5.88E+17	NA <a href=
7	#Teachers shou	FALSE	0	NA	2015/4/15 23:01	FALSE	NA	5.88E+17	NA <a href=
8	RT @CirrusAsse	FALSE	0	NA	2015/4/15 22:44	FALSE	NA	5.88E+17	NA <a href=
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24	#MDM: Mobile d	FALSE	1	NA	2015/4/15 19:20	FALSE	NA	5.88E+17	NA <a href=
25	El impacto de	FALSE	0	NA	2015/4/15 19:13	FALSE	NA	5.88E+17	NA <a href=

Data Cleaning



Regular Expression

madam, baad, dad, gooffoog

Data Cleaning



Regular Expression

```
reg <- "([a-zA-Z0-9]+://)?([a-zA-Z0-9-  
9_]+:[a-zA-Z0-9_]+@)?([a-zA-Z0-9.-  
]+\\.[A-Za-z]{2,4})(:[0-9]+)?(/.*)?«
```

Data Cleaning



Regular Expression

```
reg <- "([a-zA-Z0-9]+://)?([a-zA-Z0-9-  
9_]+:[a-zA-Z0-9_]+@)?([a-zA-Z0-9.-  
]+\\.[A-Za-z]{2,4})(:[0-9]+)?(/.*)?«
```

www.regular-expressions.info

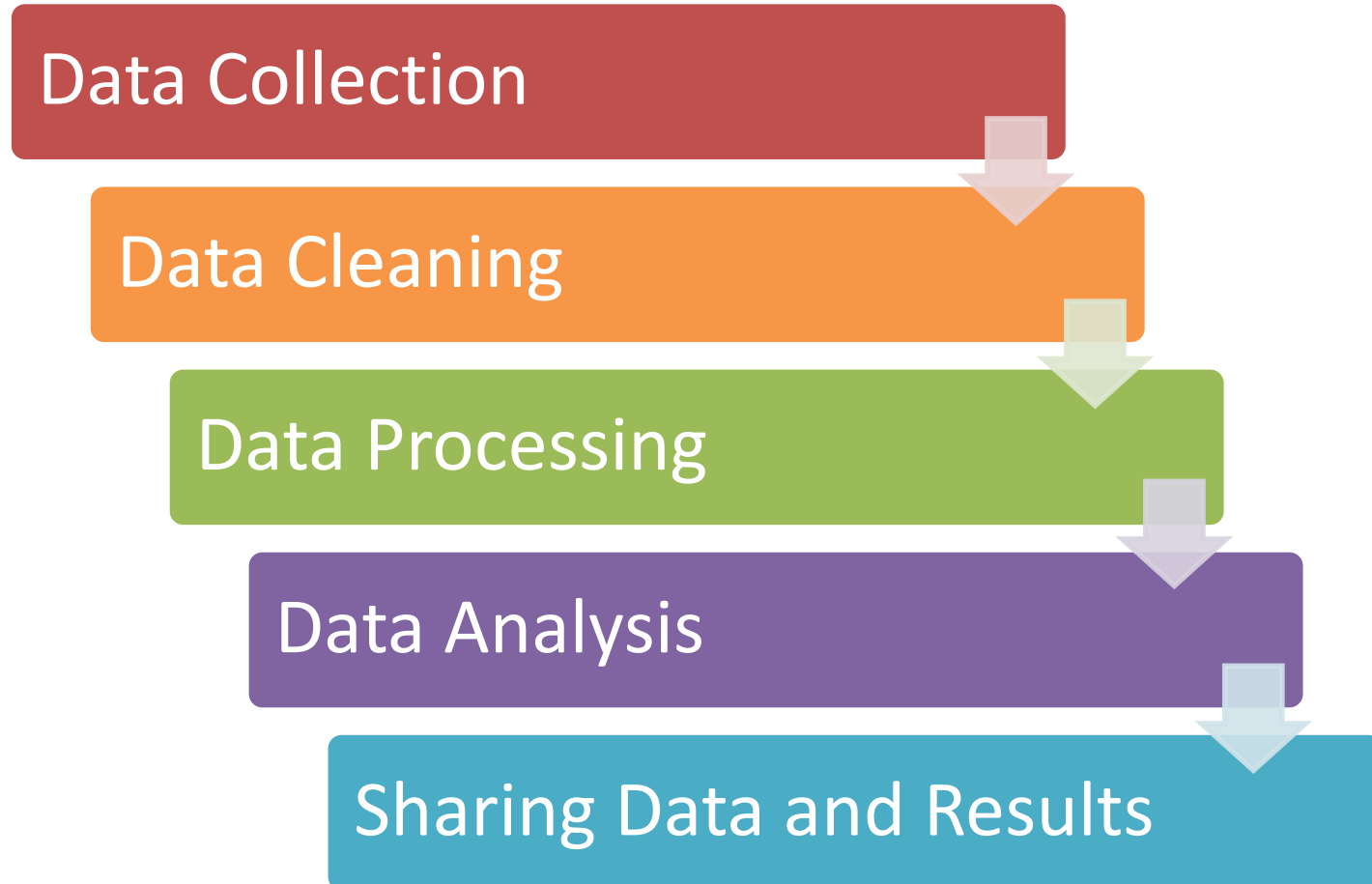
Data Cleaning



Clean tweets

1. Go to *Data Cleaning/cleanData.R*
2. Run lines 1-57
3. **Practice:** Clean the diary data yourself

Research Pipeline



Data Processing

Basic Procedures:

Data Processing

Basic Procedures:

- 1. Remove punctuation**

Data Processing

Basic Procedures:

- 1. Remove punctuation**
- 2. Remove other non-characters**

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters

!@#\$%^&*()_+-~|\\/<>

Data Processing

Basic Procedures:

- 1. Remove punctuation**
- 2. Remove other non-characters**
- 3. Remove stop words**

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words

a, an, the, he, him, I, me, ...

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words
4. Lowercases

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words
4. Lowercases
5. Stem

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words
4. Lowercases
5. Stem

do
does
did

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words
4. Lowercases
5. Stem

go
goes
went

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words
4. Lowercases
5. Stem

lie
lay
laid

Data Processing

Basic Procedures:

1. Remove punctuation
2. Remove other non-characters
3. Remove stop words
4. Lowercases
5. Stem

try
tries
tried

Data Processing

Assumption:

1. Bag of words

Data Processing

Assumption:

1. Bag of words

A dog bites a man.

A man bites a dog.

Data Processing

Assumption:

1. Bag of words

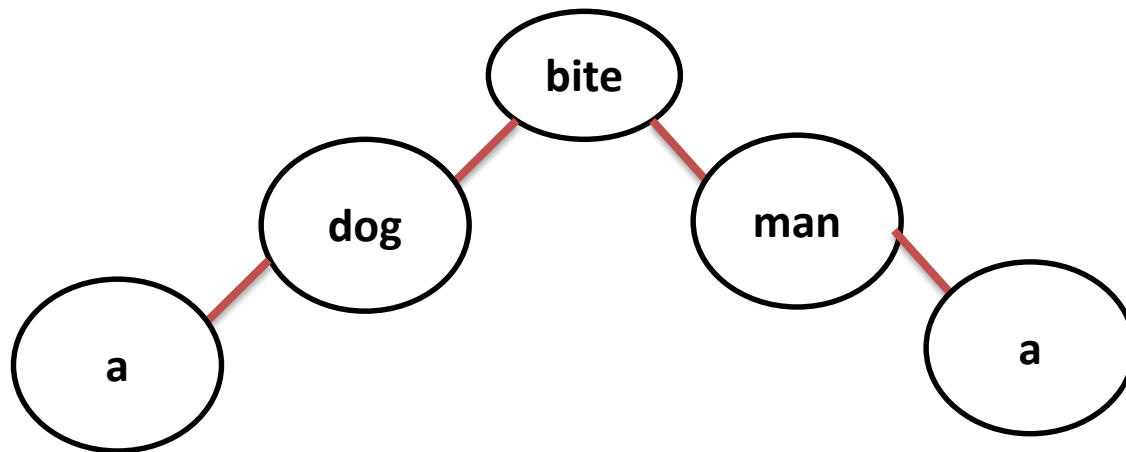
“a”, “man”, “dog”, “bites”

Data Processing

Assumption:

1. Bag of words

“a”, “man”, “dog”, “bites”



Data Processing

Assumption:

- 1. Bag of words**
- 2. Words as features**

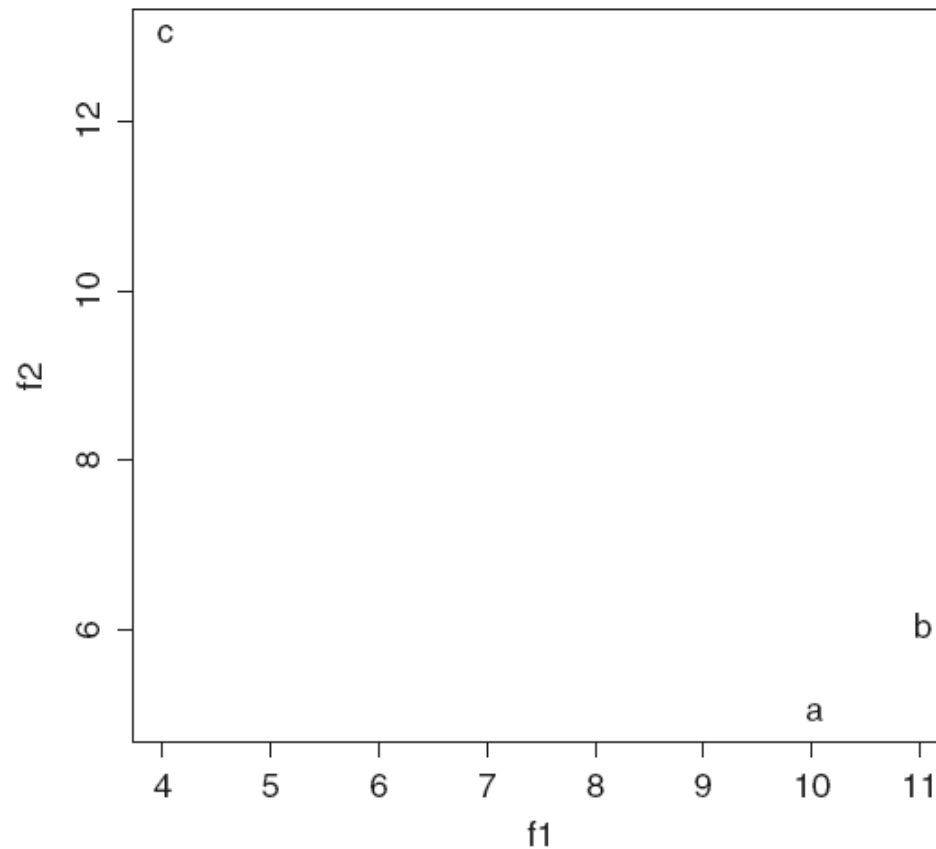
Data Processing

Explanation:

	f1	f2
a	10	5
b	11	6
c	4	13

Data Processing

Explanation:



Data Processing

Goal:

[illegible]

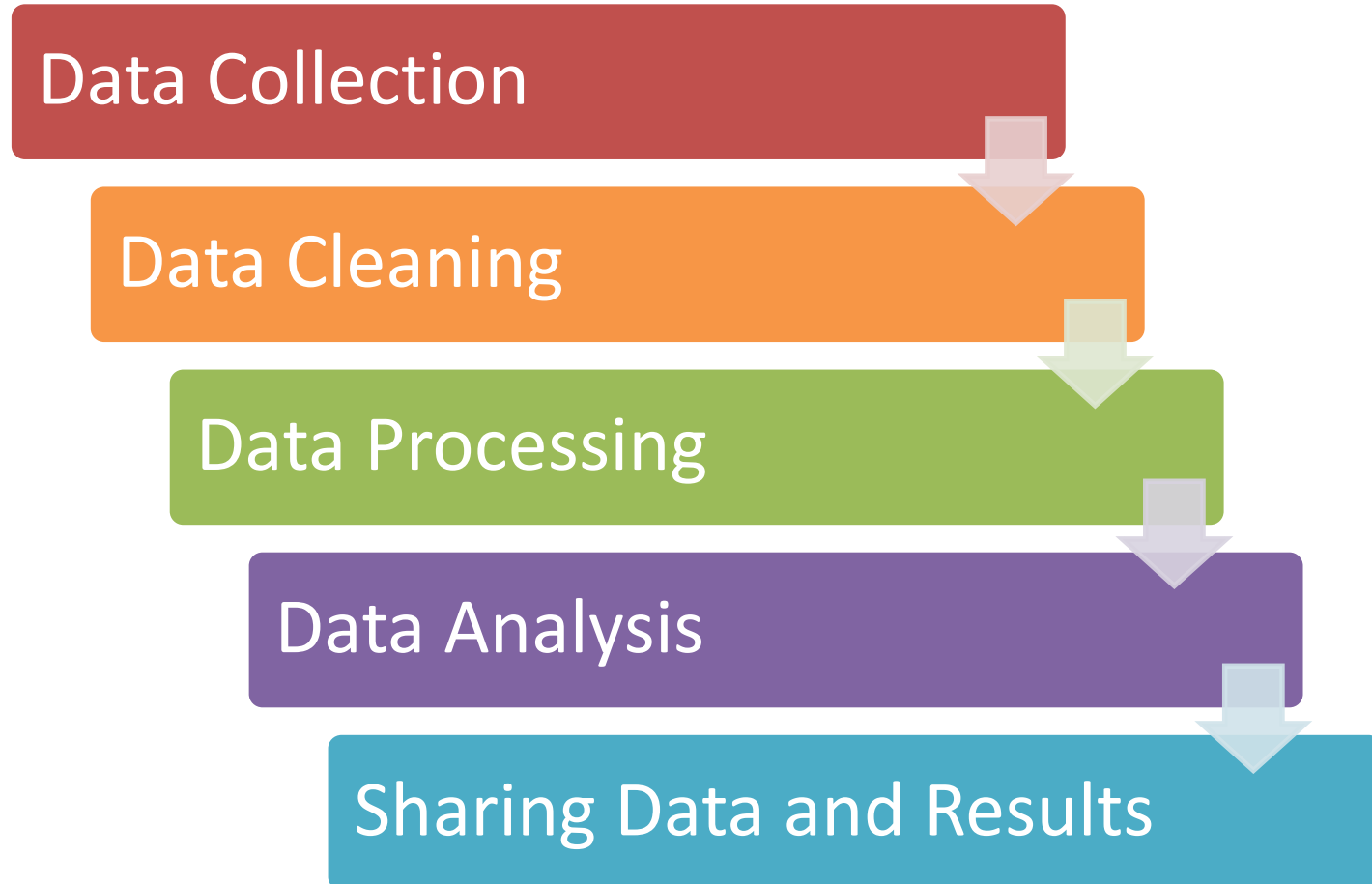
Data Processing



Data Processing

1. Go to *Data Processing/preProcess.R*
2. Run lines 1-45
3. **Practice:** Process the cleaned diary data yourself.

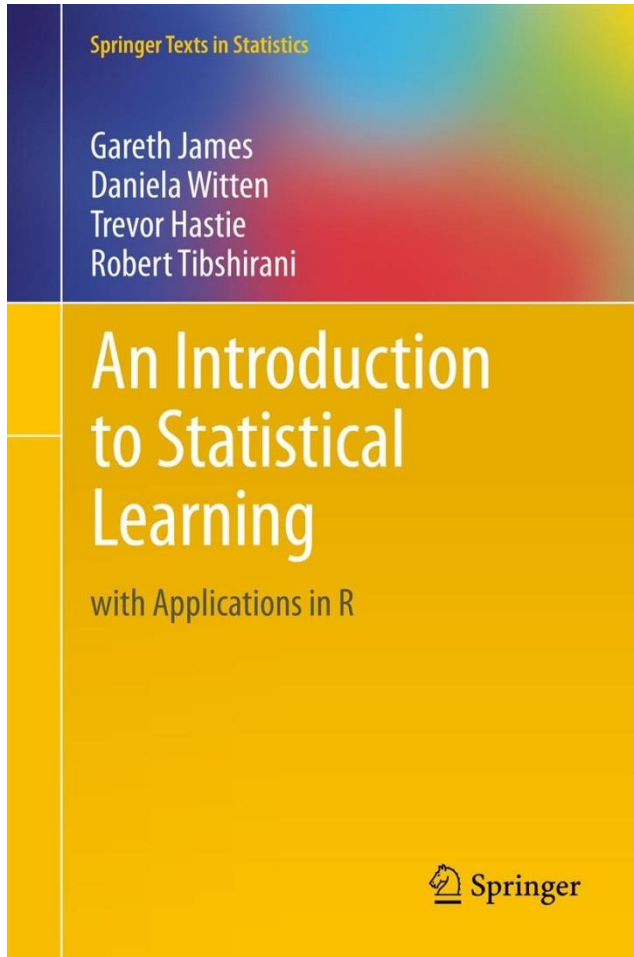
Research Pipeline



Data Analysis

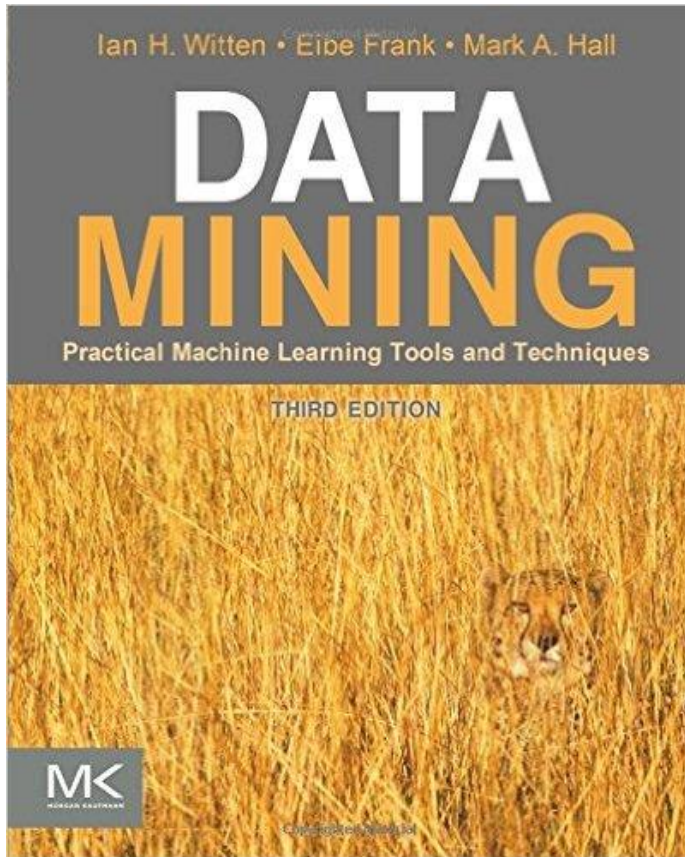
- **Unsupervised Learning**
 - **Clustering Analysis**
 - **Sentimental Analysis**
 - **Latent Semantic Analysis**
- **Supervised Learning**
 - **Support Vector Machine**
 - **Random Forests**
 - **.....**

Data Analysis



An Introduction to Statistical Learning with Application in R

Data Analysis



Data Mining: Practical Machine Learning Tools and Techniques

Data Analysis

- **Lexical Variety**
- **Sentimental Analysis**
- **Clustering Analysis**

Data Analysis

Lexical Variety

Data Analysis

Lexical Variety

- 1. Vocabulary Richness = Number of unique words / Total number of words**
- 2. Mean Word Frequency = Sum of unique Word Frequency / Total number of unique words**

Data Analysis



Lexical Variety of students' diaries

1. Find data at *Data/diary.csv*
2. *Clean the data*
3. *Process the data*
4. Go to *Data Analysis/lexicalVar.R*

Data Analysis

Sentimental Analysis

Data Analysis



Lexical Variety of students' diaries

1. Find data at *Data/diary.csv*
2. *Clean the data*
3. *Process the data*
4. Go to *Data Analysis/sentiment.R*

Data Analysis

Clustering Analysis

Renkl, A. (1997). Learning from worked-out examples: A study on individual differences. *Cognitive science*, 21(1), 1-29.

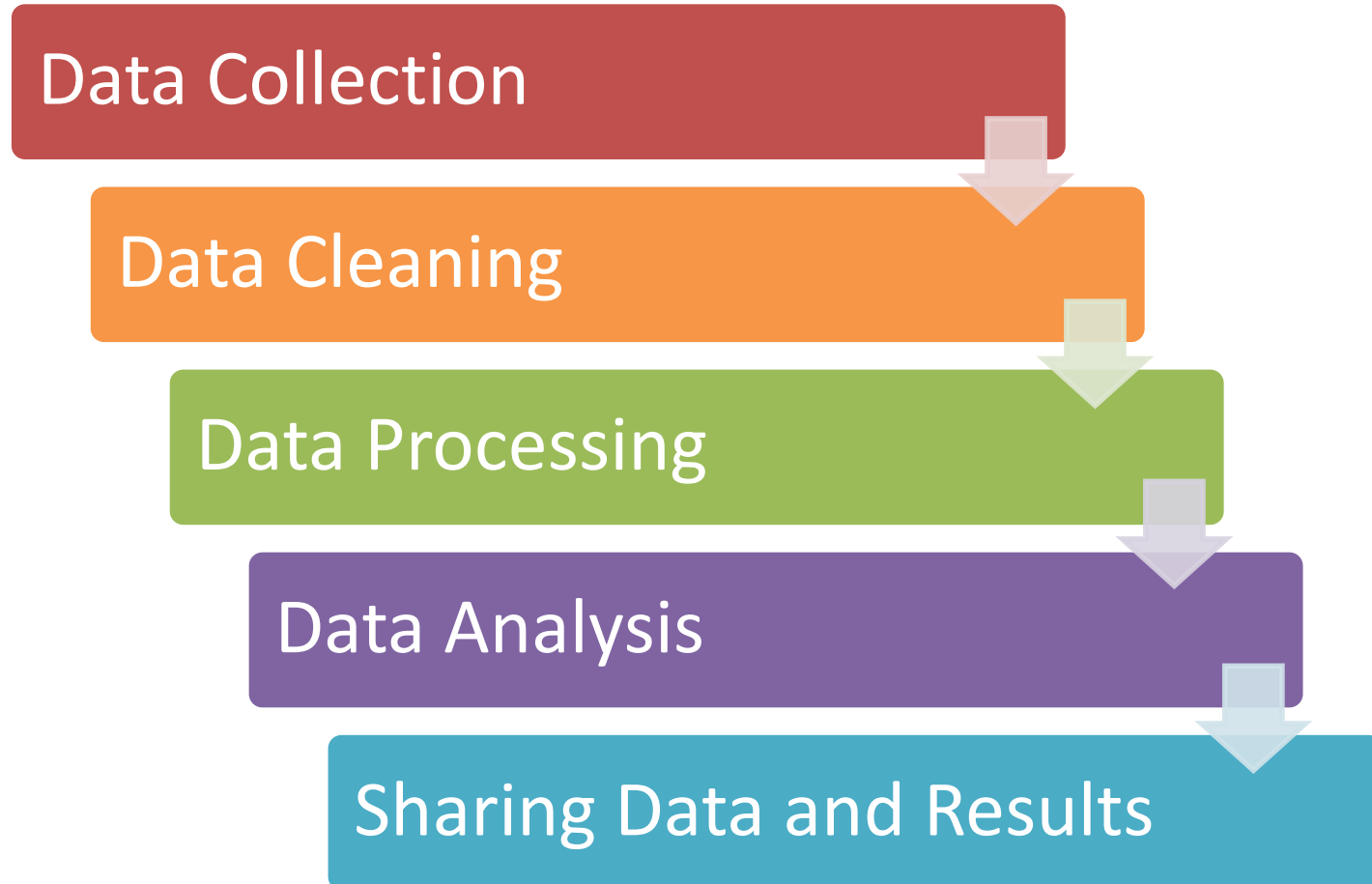
Data Analysis



Clustering Analysis

1. Go to *Data Analysis/hclusterofwords.R*

Research Pipeline



Sharing Data and Results

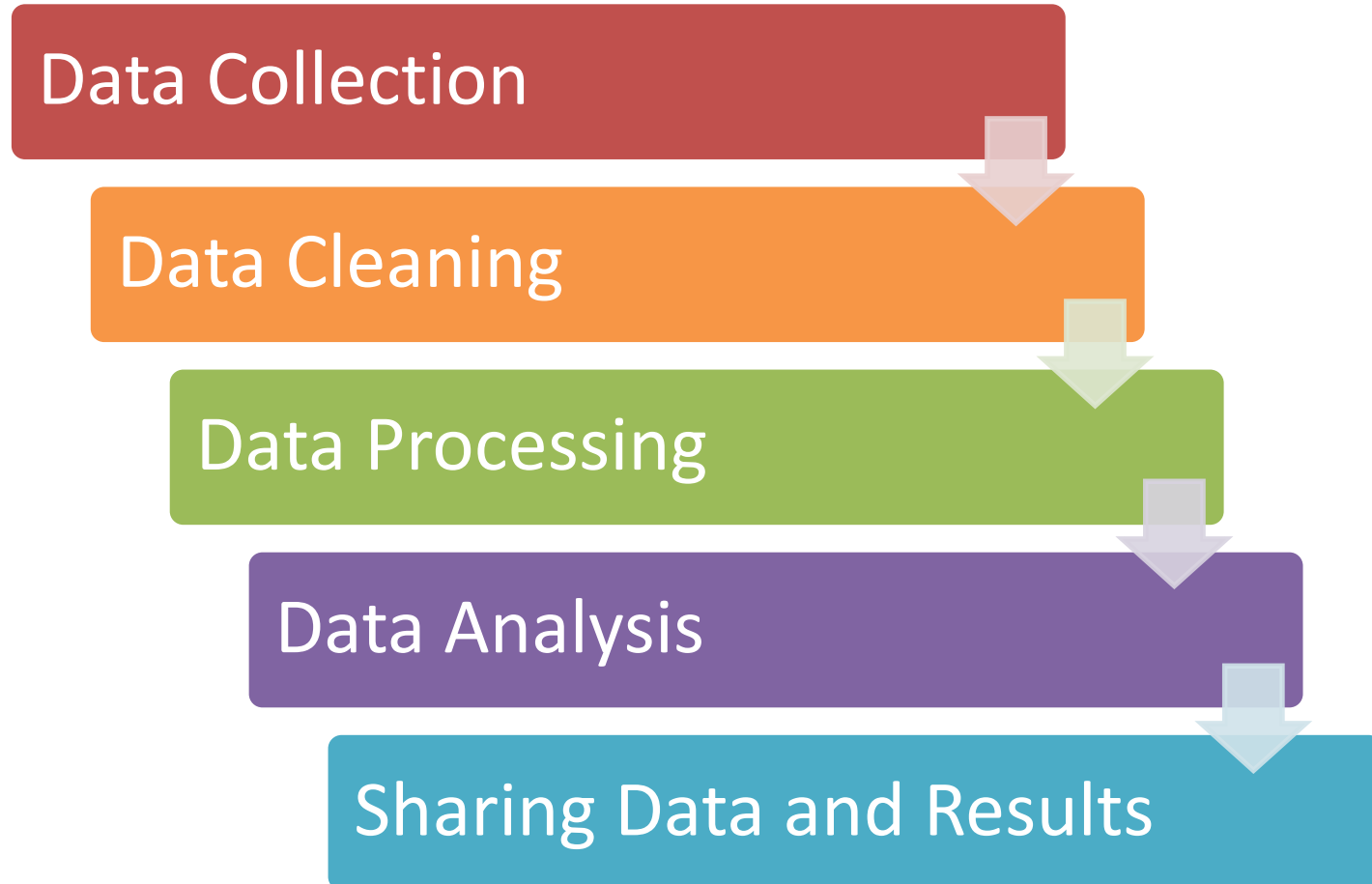
- **Git + GitHub**

- **Git:** <https://git-scm.com/downloads>
-  <https://github.com/Neo-Hao>

- **KnitR + Rpubs**

- **Example:** rpubs.com/neohao/online-help-seeking

Research Pipeline



Thanks!

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