

# week13 exercise

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```
library(brolgar)
library(ptmixed)
library(reshape2)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(magrittr)
```

## Exercise 1

```
df_long = as.data.frame(heights)
```

1

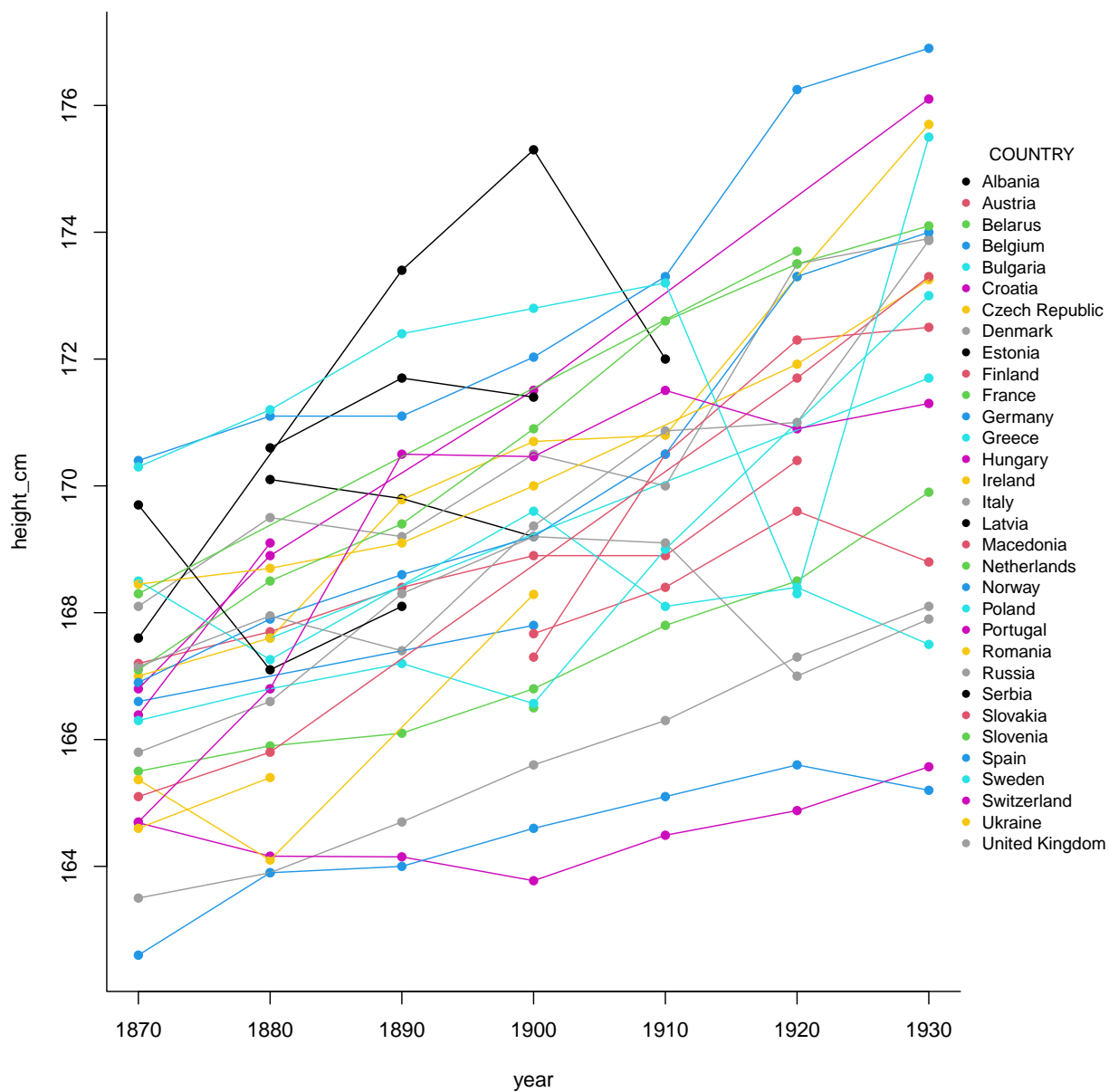
```
df_long1 = df_long[(df_long$continent == "Europe") & (df_long$year >=
  1870) & (df_long$year <= 1930), ]
head(df_long1, 5)
```

```
##   country continent year height_cm
## 6  Albania      Europe 1880      170.1
## 7  Albania      Europe 1890      169.8
## 8  Albania      Europe 1900      169.2
## 74 Austria      Europe 1870      167.2
## 75 Austria      Europe 1880      167.7
```

```

coun_n = length(unique(df_long1$country))
make.spaghetti(year, height_cm, id = country, group = country,
  data = df_long1, legend.title = "COUNTRY", col = 1:coun_n,
  cex.legend = 0.85, legend.inset = -0.18)

```



## Exercise 2

1

```
df_long_wd = dcast(df_long, country + continent ~ year,
  value.var = "height_cm")
head(df_long_wd, 5)
```

```
##      country continent 1550 1650 1660 1670 1680 1690 1700 1710 1720 1730 1740
## 1 Afghanistan      Asia   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 2  Albania      Europe   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 3  Algeria      Africa   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 4  Angola      Africa   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 5  Argentina Americas   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
##    1750 1760 1770 1780 1790 1800 1810 1820 1830 1840 1850 1860 1870
## 1   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA 168.4
## 2   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 3   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 4   NA   NA   NA   NA 160.4 158.6 160.5   NA   NA   NA   NA   NA   NA
## 5   NA   NA 170.3 168.2 168.0 168.0 168.8 169.9 170.9 169.6 168.2 167.4 167.6
##    1880 1890 1900 1910 1920 1930 1940 1950 1960 1970 1980 1990
## 1 165.690   NA   NA   NA   NA 166.8   NA   NA   NA   NA   NA 167.1
## 2 170.100 169.800 169.200   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 3    NA    NA    NA 168.8 166.241 169.0   NA   NA   NA   NA   NA 171.3
## 4 168.800 169.100 168.100 168.0 165.700 166.7   NA   NA   NA   NA   NA   NA
## 5 167.565 167.792 167.868 168.2 169.000 169.8 170.6 170.8   NA   NA   NA 174.4
##    2000
## 1 161.4
## 2 167.9
## 3 169.5
## 4   NA
## 5   NA
```

2

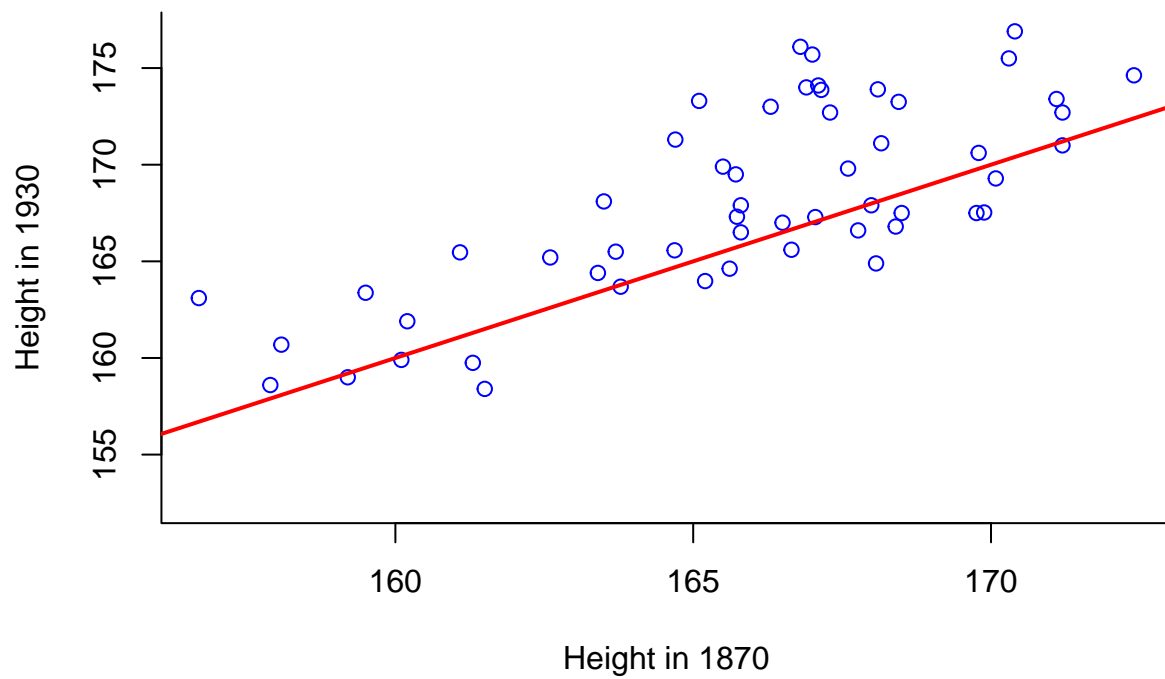
```
rename_f = function(x) {
  paste0("height_", as.character(x))
}
colnames(df_long_wd)[3:ncol(df_long_wd)] = sapply(colnames(df_long_wd)[3:ncol(df_long_wd)],
  rename_f)
head(df_long_wd, 5)
```

```
##      country continent height_1550 height_1650 height_1660 height_1670
## 1 Afghanistan      Asia          NA          NA          NA          NA
## 2  Albania      Europe          NA          NA          NA          NA
## 3  Algeria      Africa          NA          NA          NA          NA
## 4  Angola      Africa          NA          NA          NA          NA
## 5  Argentina Americas          NA          NA          NA          NA
## height_1680 height_1690 height_1700 height_1710 height_1720 height_1730
```

```
## 1      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      NA
## 5      NA      NA      NA      NA      NA      NA
## height_1740 height_1750 height_1760 height_1770 height_1780 height_1790
## 1      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      160.4
## 5      NA      NA      NA      170.3      168.2      168.0
## height_1800 height_1810 height_1820 height_1830 height_1840 height_1850
## 1      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA
## 4      158.6      160.5      NA      NA      NA      NA
## 5      168.0      168.8      169.9      170.9      169.6      168.2
## height_1860 height_1870 height_1880 height_1890 height_1900 height_1910
## 1      NA      168.4      165.690      NA      NA      NA
## 2      NA      NA      170.100      169.800      169.200      NA
## 3      NA      NA      NA      NA      NA      168.8
## 4      NA      NA      168.800      169.100      168.100      168.0
## 5      167.4      167.6      167.565      167.792      167.868      168.2
## height_1920 height_1930 height_1940 height_1950 height_1960 height_1970
## 1      NA      166.8      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA
## 3      166.241      169.0      NA      NA      NA      NA
## 4      165.700      166.7      NA      NA      NA      NA
## 5      169.000      169.8      170.6      170.8      NA      NA
## height_1980 height_1990 height_2000
## 1      NA      167.1      161.4
## 2      NA      NA      167.9
## 3      NA      171.3      169.5
## 4      NA      NA      NA
## 5      NA      174.4      NA
```

3

```
par(bty = "l")
plot(x = df_long_wd$height_1870, y = df_long_wd$height_1930,
     type = "p", col = "blue", xlab = "Height in 1870", ylab = "Height in 1930")
abline(a = 0, b = 1, col = "red", lwd = 2)
```



4

For most countries, height increase during 1870 to 1930. But there are some exceptions.

### Exercise 3

1

```
load("data/data_metadata_country.RData")
```

2

Horizontal merge.

3

Country.Code can be key for the merge.

```

population = select(population, !X)
metadata = select(metadata, !X)
data_df = merge(population, metadata, by = "Country.Code",
               all = T)
head(data_df, 5)

```

##	Country.Code	Country.Name	Indicator.Name	Indicator.Code
## 1	ABW	Aruba	Population, total	SP.POP.TOTL
## 2	AFE	Africa Eastern and Southern	Population, total	SP.POP.TOTL
## 3	AFG	Afghanistan	Population, total	SP.POP.TOTL
## 4	AFW	Africa Western and Central	Population, total	SP.POP.TOTL
## 5	AGO	Angola	Population, total	SP.POP.TOTL
##	X1960	X1961	X1962	X1963
## 1	54608	55811	56682	57475
## 2	130692579	134169237	137835590	141630546
## 3	8622466	8790140	8969047	9157465
## 4	97256290	99314028	101445032	103667517
## 5	5357195	5441333	5521400	5599827
##	X1967	X1968	X1969	X1970
## 1	59522	59471	59330	59106
## 2	158313235	162875171	167596160	172475766
## 3	10010030	10247780	10494489	10752971
## 4	113319950	115921723	118615741	121424797
## 5	5827503	5868203	5928386	6029700
##	X1974	X1975	X1976	X1977
## 1	60028	60715	61193	61465
## 2	193512956	199284304	205202669	211120911
## 3	11869879	12157386	12425267	12687301
## 4	133953892	137548613	141258400	145122851
## 5	6802494	7032713	7266780	7511895
##	X1981	X1982	X1983	X1984
## 1	62614	63116	63683	64174
## 2	237937461	245386717	252779730	260209149
## 3	11155195	10088289	9951449	10243686
## 4	162323313	167023385	171566640	176054495
## 5	8631457	8947152	9276707	9617702
##	X1988	X1989	X1990	X1991
## 1	64332	64596	65712	67864
## 2	292795186	301124880	309890664	318544083
## 3	10383460	10673168	10694796	10745167
## 4	195969722	201392200	206739024	212172888
## 5	11060261	11439498	11828638	12228691
##	X1995	X1996	X1997	X1998
## 1	77050	79417	81858	84355
## 2	353466601	362985802	372352230	381715600
## 3	16418912	17106595	17788819	18493132
## 4	235861484	242200260	248713095	255482918
## 5	13912253	14383350	14871146	15366864
##	X2002	X2003	X2004	X2005
## 1	91781	92701	93540	94483
## 2	422741118	433807484	445281555	457153837

```

## 3 21000256 22645130 23553551 24411191 25442944 25903301 26427199
## 4 284952322 292977949 301265247 309824829 318601484 327612838 336893835
## 5 17516139 18124342 18771125 19450959 20162340 20909684 21691522
##      X2009      X2010      X2011      X2012      X2013      X2014      X2015
## 1      99212      100341      101288      102112      102880      103594      104257
## 2 509410477 523459657 537792950 552530654 567892149 583651101 600008424
## 3 27385307 28189672 29249157 30466479 31541209 32716210 33753499
## 4 346475221 356337762 366489204 376797999 387204553 397855507 408690375
## 5 22507674 23364185 24259111 25188292 26147002 27128337 28127721
##      X2016      X2017      X2018      X2019      X2020      X2021      X2022
## 1      104874      105439      105962      106442      106585      106537      106445
## 2 616377605 632746570 649757148 667242986 685112979 702977106 720839314
## 3 34636207 35643418 36686784 37769499 38972230 40099462 41128771
## 4 419778384 431138704 442646825 454306063 466189102 478185907 490330870
## 5 29154746 30208628 31273533 32353588 33428486 34503774 35588987
##      Region      IncomeGroup
## 1 Latin America & Caribbean      High income
## 2
## 3      South Asia      Low income
## 4
## 5      Sub-Saharan Africa Lower middle income
##
## 1
## 2
## 3 The reporting period for national accounts data is designated as either calendar year basis (CY) o
## 4
## 5
##      TableName
## 1      Aruba
## 2 Africa Eastern and Southern
## 3      Afghanistan
## 4 Africa Western and Central
## 5      Angola

```

5

```
data_df = filter(data_df, !Region == "")
```

6

a)

```
summarize(group_by(data_df, Region), large2020 = max(X2020))
```

```

## # A tibble: 7 x 2
##   Region      large2020
##   <chr>      <dbl>
## 1 East Asia & Pacific 1411100000
## 2 Europe & Central Asia 144073139

```

```
## 3 Latin America & Caribbean 213196304
## 4 Middle East & North Africa 107465134
## 5 North America 331511512
## 6 South Asia 1396387127
## 7 Sub-Saharan Africa 208327405
```

b)

```
summarize(group_by(data_df, Region), large2020_country = Country.Name[which.max(X2020)])
```

```
## # A tibble: 7 x 2
##   Region                large2020_country
##   <chr>                <chr>
## 1 East Asia & Pacific    China
## 2 Europe & Central Asia  Russian Federation
## 3 Latin America & Caribbean Brazil
## 4 Middle East & North Africa Egypt, Arab Rep.
## 5 North America         United States
## 6 South Asia            India
## 7 Sub-Saharan Africa    Nigeria
```

## Exercise 4

```
data(iris)
df.list = split(iris, iris$Species)
lapply(df.list, summary)
```

```
## $setosa
##   Sepal.Length   Sepal.Width   Petal.Length   Petal.Width
##   Min.    :4.300   Min.    :2.300   Min.    :1.000   Min.    :0.100
##   1st Qu.:4.800   1st Qu.:3.200   1st Qu.:1.400   1st Qu.:0.200
##   Median :5.000   Median :3.400   Median :1.500   Median :0.200
##   Mean    :5.006   Mean    :3.428   Mean    :1.462   Mean    :0.246
##   3rd Qu.:5.200   3rd Qu.:3.675   3rd Qu.:1.575   3rd Qu.:0.300
##   Max.    :5.800   Max.    :4.400   Max.    :1.900   Max.    :0.600
##   Species
##   setosa      :50
##   versicolor : 0
##   virginica  : 0
##
##
##
## $versicolor
##   Sepal.Length   Sepal.Width   Petal.Length   Petal.Width   Species
##   Min.    :4.900   Min.    :2.000   Min.    :3.000   Min.    :1.000   setosa      : 0
##   1st Qu.:5.600   1st Qu.:2.525   1st Qu.:4.000   1st Qu.:1.200   versicolor:50
##   Median :5.900   Median :2.800   Median :4.350   Median :1.300   virginica  : 0
##   Mean    :5.936   Mean    :2.770   Mean    :4.260   Mean    :1.326
```



```
## 3rd Qu.:6.300 3rd Qu.:3.000 3rd Qu.:4.60 3rd Qu.:1.500
## Max. :7.000 Max. :3.400 Max. :5.10 Max. :1.800
##
## $virginica
## Sepal.Length Sepal.Width Petal.Length Petal.Width
## Min. :4.900 Min. :2.200 Min. :4.500 Min. :1.400
## 1st Qu.:6.225 1st Qu.:2.800 1st Qu.:5.100 1st Qu.:1.800
## Median :6.500 Median :3.000 Median :5.550 Median :2.000
## Mean :6.588 Mean :2.974 Mean :5.552 Mean :2.026
## 3rd Qu.:6.900 3rd Qu.:3.175 3rd Qu.:5.875 3rd Qu.:2.300
## Max. :7.900 Max. :3.800 Max. :6.900 Max. :2.500
## Species
## setosa : 0
## versicolor: 0
## virginica :50
##
##
##
```

1

```
iris |>
  split(iris$Species) |>
  lapply(summary)
```

```
## $setosa
## Sepal.Length Sepal.Width Petal.Length Petal.Width
## Min. :4.300 Min. :2.300 Min. :1.000 Min. :0.100
## 1st Qu.:4.800 1st Qu.:3.200 1st Qu.:1.400 1st Qu.:0.200
## Median :5.000 Median :3.400 Median :1.500 Median :0.200
## Mean :5.006 Mean :3.428 Mean :1.462 Mean :0.246
## 3rd Qu.:5.200 3rd Qu.:3.675 3rd Qu.:1.575 3rd Qu.:0.300
## Max. :5.800 Max. :4.400 Max. :1.900 Max. :0.600
## Species
## setosa :50
## versicolor: 0
## virginica : 0
##
##
##
## $versicolor
## Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## Min. :4.900 Min. :2.000 Min. :3.00 Min. :1.000 setosa : 0
## 1st Qu.:5.600 1st Qu.:2.525 1st Qu.:4.00 1st Qu.:1.200 versicolor:50
## Median :5.900 Median :2.800 Median :4.35 Median :1.300 virginica : 0
## Mean :5.936 Mean :2.770 Mean :4.26 Mean :1.326
## 3rd Qu.:6.300 3rd Qu.:3.000 3rd Qu.:4.60 3rd Qu.:1.500
## Max. :7.000 Max. :3.400 Max. :5.10 Max. :1.800
##
## $virginica
## Sepal.Length Sepal.Width Petal.Length Petal.Width
```

```
## Min. :4.900 Min. :2.200 Min. :4.500 Min. :1.400
## 1st Qu.:6.225 1st Qu.:2.800 1st Qu.:5.100 1st Qu.:1.800
## Median :6.500 Median :3.000 Median :5.550 Median :2.000
## Mean :6.588 Mean :2.974 Mean :5.552 Mean :2.026
## 3rd Qu.:6.900 3rd Qu.:3.175 3rd Qu.:5.875 3rd Qu.:2.300
## Max. :7.900 Max. :3.800 Max. :6.900 Max. :2.500
## Species
## setosa : 0
## versicolor: 0
## virginica :50
##
##
##
```

2

```
iris %>%
  split(iris$Species) %>%
  lapply(summary)
```

```
## $setosa
## Sepal.Length Sepal.Width Petal.Length Petal.Width
## Min. :4.300 Min. :2.300 Min. :1.000 Min. :0.100
## 1st Qu.:4.800 1st Qu.:3.200 1st Qu.:1.400 1st Qu.:0.200
## Median :5.000 Median :3.400 Median :1.500 Median :0.200
## Mean :5.006 Mean :3.428 Mean :1.462 Mean :0.246
## 3rd Qu.:5.200 3rd Qu.:3.675 3rd Qu.:1.575 3rd Qu.:0.300
## Max. :5.800 Max. :4.400 Max. :1.900 Max. :0.600
## Species
## setosa :50
## versicolor: 0
## virginica : 0
##
##
##
## $versicolor
## Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## Min. :4.900 Min. :2.000 Min. :3.00 Min. :1.000 setosa : 0
## 1st Qu.:5.600 1st Qu.:2.525 1st Qu.:4.00 1st Qu.:1.200 versicolor:50
## Median :5.900 Median :2.800 Median :4.35 Median :1.300 virginica : 0
## Mean :5.936 Mean :2.770 Mean :4.26 Mean :1.326
## 3rd Qu.:6.300 3rd Qu.:3.000 3rd Qu.:4.60 3rd Qu.:1.500
## Max. :7.000 Max. :3.400 Max. :5.10 Max. :1.800
##
## $virginica
## Sepal.Length Sepal.Width Petal.Length Petal.Width
## Min. :4.900 Min. :2.200 Min. :4.500 Min. :1.400
## 1st Qu.:6.225 1st Qu.:2.800 1st Qu.:5.100 1st Qu.:1.800
## Median :6.500 Median :3.000 Median :5.550 Median :2.000
## Mean :6.588 Mean :2.974 Mean :5.552 Mean :2.026
## 3rd Qu.:6.900 3rd Qu.:3.175 3rd Qu.:5.875 3rd Qu.:2.300
```

```
## Max.      :7.900    Max.      :3.800    Max.      :6.900    Max.      :2.500
##           Species
## setosa     : 0
## versicolor: 0
## virginica  :50
##
##
##
```

## Exercise 5

1

```
data_df = read.csv("data/irish_polls.csv")
data_df[data_df == "Not Available"] = NA
to_decimal = function(x) {
  return(sub("%", "", x))
}
data_df[, 10:21] = lapply(data_df[, 10:21], to_decimal)
data_df[, 10:21] = as.numeric(unlist(data_df[, 10:21]))/100
```

2

```
data_df = filter(data_df, (Fieldwork.End >= "2021-05-27") &
  (Fieldwork.End <= "2021-09-09"))
colna = c(colnames(data_df)[1:9], colnames(data_df)[10:21][!apply(is.na(data_df[,
  10:21]), 2, any)])
data_df = data_df[, colna]
```

3

```
data_df = mutate(data_df, Fieldwork.End = as.Date(Fieldwork.End,
  "%Y-%m-%d"))
data_df_long = melt(data_df, id.vars = colnames(data_df)[1:9])
head(data_df_long, 5)
```

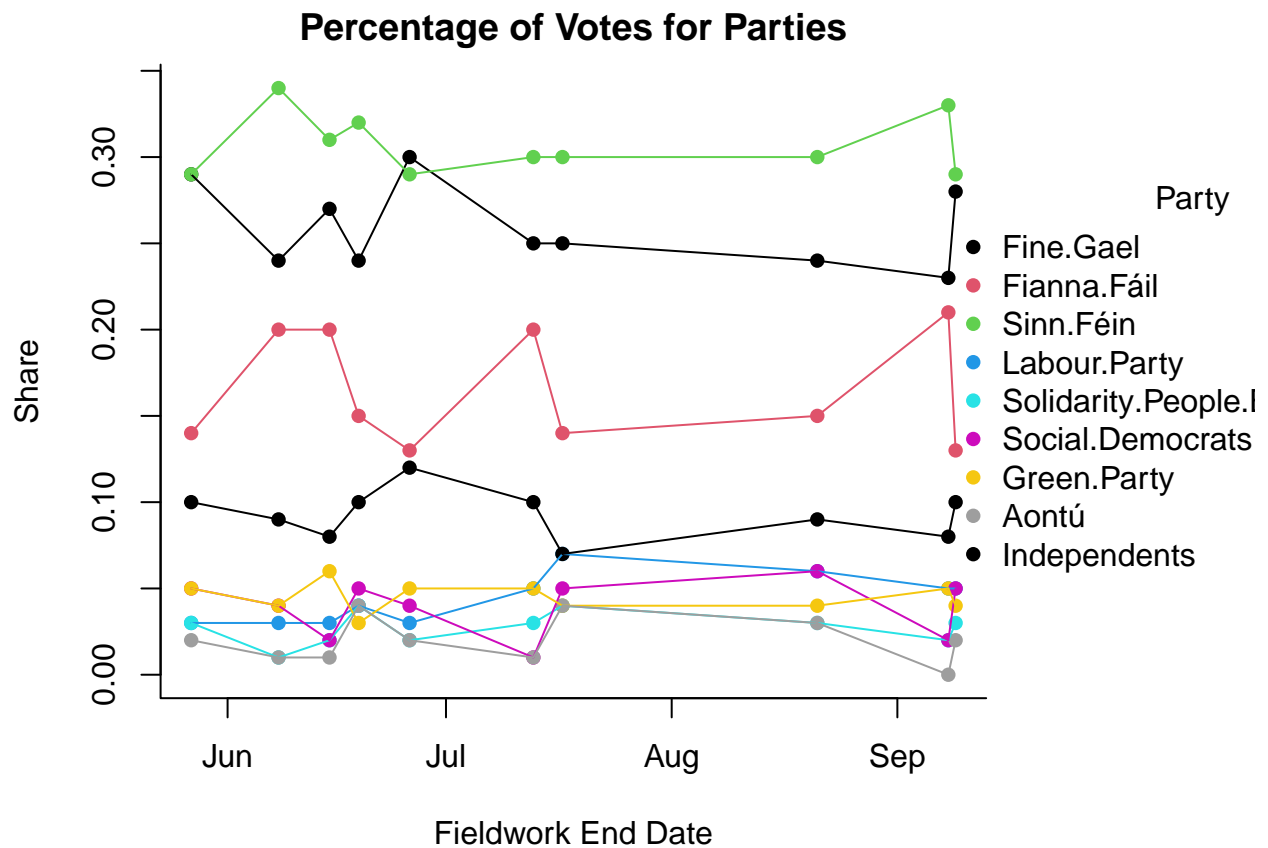
```
##           Polling.Firm      Commissioners Fieldwork.Start Fieldwork.End
## 1              Red C      Business Post      2021-09-03      2021-09-09
## 2 Behaviour and Attitudes The Sunday Times      2021-08-26      2021-09-08
## 3      Ireland Thinks Irish Mail on Sunday      2021-08-21      2021-08-21
## 4      Ireland Thinks Irish Mail on Sunday      2021-07-17      2021-07-17
## 5 Behaviour and Attitudes The Sunday Times      2021-07-01      2021-07-13
##           Scope Sample.Size Sample.Size.Qualification Participation Precision
## 1 National          1031              Provided              <NA>          1%
## 2 National           922              Provided              <NA>          1%
## 3 National          1203              Provided              <NA>          1%
```

```
## 4 National      1001      Provided      <NA>      1%
## 5 National      1001      Provided      <NA>      1%
##   variable value
## 1 Fine.Gael    0.28
## 2 Fine.Gael    0.23
## 3 Fine.Gael    0.24
## 4 Fine.Gael    0.25
## 5 Fine.Gael    0.25
```

```
data_df_long = rename(data_df_long, Party = variable, Share = value)
```

4

```
coun_n = length(unique(data_df_long$Party))
make.spaghetti(Fieldwork.End, Share, id = Party, group = Party,
  data = data_df_long, legend.title = "Party", col = 1:coun_n,
  cex.legend = 1, legend.inset = -0.55, xlab = "Fieldwork End Date",
  ylab = "Share", title = "Percentage of Votes for Parties")
```



5

```
mean_share = data_df_long |>
  group_by(Party) |>
  summarize(mean = mean(Share))
mean_share
```

```
## # A tibble: 9 x 2
##   Party          mean
##   <fct>         <dbl>
## 1 Fine.Gael      0.259
## 2 Fianna.Fáil    0.165
## 3 Sinn.Féin      0.307
## 4 Labour.Party   0.044
## 5 Solidarity.People.Before.Profit 0.027
## 6 Social.Democrats 0.039
## 7 Green.Party    0.045
## 8 Aontú          0.02
## 9 Independents   0.093
```

6

```
mean_share$Party[which.max(mean_share$mean)]
```

```
## [1] Sinn.Féin
## 9 Levels: Fine.Gael Fianna.Fáil Sinn.Féin ... Independents
```

Yes, the line of Sinn.Fein is in the highest position.