

# Assignment 4

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## 10.5

### 1.

Tibbles are data frames, and only print a limited number of rows and show the class on top of each column.

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 3.4.3
```

```
## -- Attaching packages ----- tidyverse 1.2.1 --
```

```
## v ggplot2 2.2.1      v purrr  0.2.4
```

```
## v tibble  1.4.2      v dplyr  0.7.4
```

```
## v tidyr   0.8.0      v stringr 1.2.0
```

```
## v readr   1.1.1      v forcats 0.3.0
```

```
## Warning: package 'ggplot2' was built under R version 3.4.3
```

```
## Warning: package 'tibble' was built under R version 3.4.3
```

```
## Warning: package 'tidyr' was built under R version 3.4.3
```

```
## Warning: package 'purrr' was built under R version 3.4.3
```

```
## Warning: package 'dplyr' was built under R version 3.4.3
```

```
## Warning: package 'stringr' was built under R version 3.4.3
```

```
## Warning: package 'forcats' was built under R version 3.4.3
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()     masks stats::lag()
```

```
mtcars
```

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4

```
## Lincoln Continental 10.4 8 460.0 215 3.00 5.424 17.82 0 0 3 4
## Chrysler Imperial 14.7 8 440.0 230 3.23 5.345 17.42 0 0 3 4
## Fiat 128 32.4 4 78.7 66 4.08 2.200 19.47 1 1 4 1
## Honda Civic 30.4 4 75.7 52 4.93 1.615 18.52 1 1 4 2
## Toyota Corolla 33.9 4 71.1 65 4.22 1.835 19.90 1 1 4 1
## Toyota Corona 21.5 4 120.1 97 3.70 2.465 20.01 1 0 3 1
## Dodge Challenger 15.5 8 318.0 150 2.76 3.520 16.87 0 0 3 2
## AMC Javelin 15.2 8 304.0 150 3.15 3.435 17.30 0 0 3 2
## Camaro Z28 13.3 8 350.0 245 3.73 3.840 15.41 0 0 3 4
## Pontiac Firebird 19.2 8 400.0 175 3.08 3.845 17.05 0 0 3 2
## Fiat X1-9 27.3 4 79.0 66 4.08 1.935 18.90 1 1 4 1
## Porsche 914-2 26.0 4 120.3 91 4.43 2.140 16.70 0 1 5 2
## Lotus Europa 30.4 4 95.1 113 3.77 1.513 16.90 1 1 5 2
## Ford Pantera L 15.8 8 351.0 264 4.22 3.170 14.50 0 1 5 4
## Ferrari Dino 19.7 6 145.0 175 3.62 2.770 15.50 0 1 5 6
## Maserati Bora 15.0 8 301.0 335 3.54 3.570 14.60 0 1 5 8
## Volvo 142E 21.4 4 121.0 109 4.11 2.780 18.60 1 1 4 2
```

## 2.

```
df <- data.frame(abc = 1, xyz = "a")
df$x
```

```
## [1] a
## Levels: a
```

```
df[, "xyz"]
```

```
## [1] a
## Levels: a
```

```
df[, c("abc", "xyz")]
```

```
##   abc xyz
## 1    1  a
```

```
tbl <- as_tibble(df)
tbl$x
```

```
## Warning: Unknown or uninitialised column: 'x'.
```

```
## NULL
```

```
tbl[, "xyz"]
```

```
## # A tibble: 1 x 1
##   xyz
##   <fct>
## 1 a
```

```
tbl[, c("abc", "xyz")]
```

```
## # A tibble: 1 x 2
##   abc xyz
##   <dbl> <fct>
## 1 1.00 a
```

### 3.

You can use the double bracket. You can't use the dollar sign, because `df$var` would look for a column named `var`.

### 4.

```
annoying <- tibble(  
  `1` = 1:10,  
  `2` = `1` * 2 + rnorm(length(`1`))  
)
```

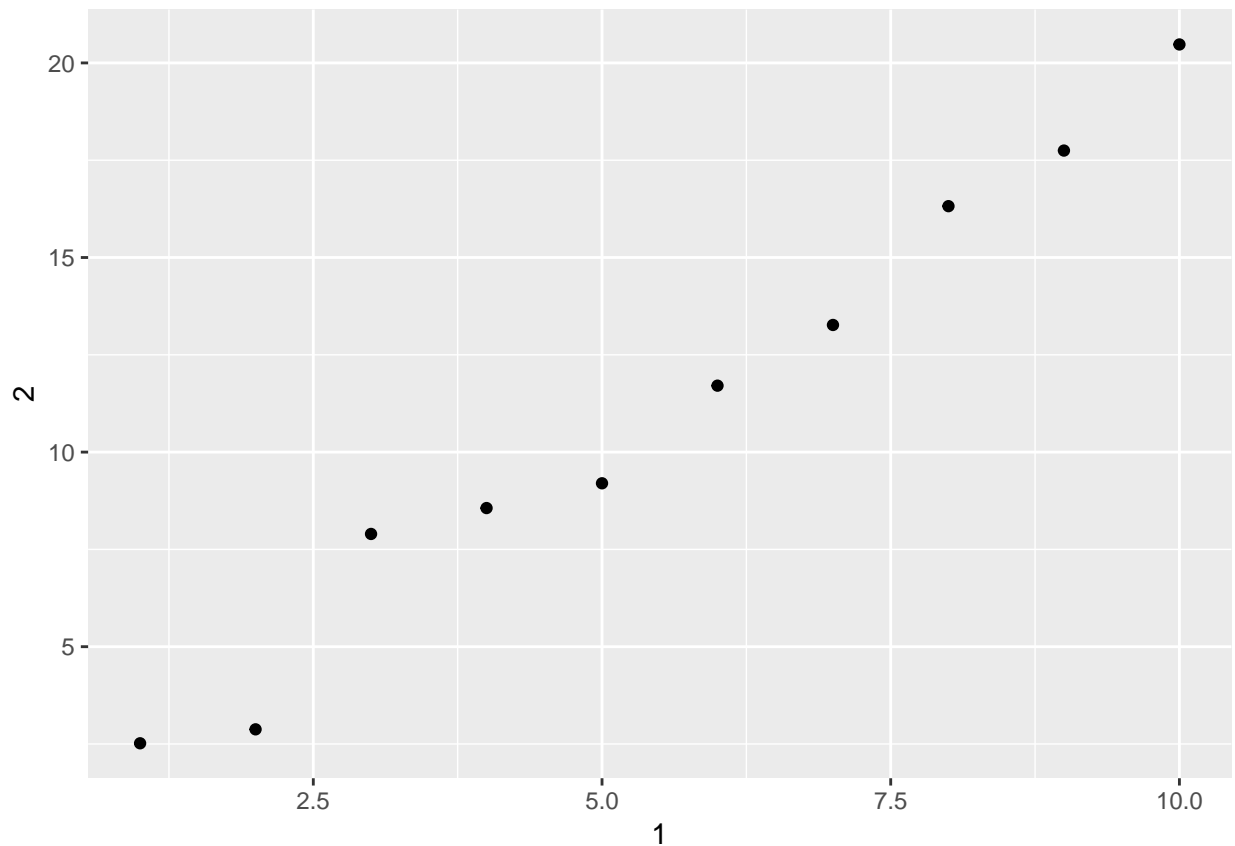
```
annoying[["1"]]
```

```
## [1] 1 2 3 4 5 6 7 8 9 10
```

```
annoying$`1`
```

```
## [1] 1 2 3 4 5 6 7 8 9 10
```

```
ggplot(annoying, aes(x = `1`, y = `2`)) +  
  geom_point()
```



```
annoying[["3"]] <- annoying$`2` / annoying$`1`  
annoying <- rename(annoying, one = `1`, two = `2`, three = `3`)  
glimpse(annoying)
```

```
## Observations: 10
## Variables: 3
## $ one    <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10
## $ two    <dbl> 2.517648, 2.876433, 7.896548, 8.561826, 9.198678, 11.707...
## $ three  <dbl> 2.517648, 1.438217, 2.632183, 2.140457, 1.839736, 1.9512...
```

## 5.

It converts named vectors to a data frame with names and values

```
enframe(c(a = 1, b = 2, c = 3))
```

```
## # A tibble: 3 x 2
##   name value
##   <chr> <dbl>
## 1 a     1.00
## 2 b     2.00
## 3 c     3.00
```

## 6.

The print function for tibbles is in `print.tbl_df`. The option `n_extra` determines the number of extra columns to print information for.

### 12.6.1

```
who1 <- who %>%
  gather(new_sp_m014:newrel_f65, key = "key", value = "cases", na.rm = TRUE)
glimpse(who1)
```

```
## Observations: 76,046
## Variables: 6
## $ country <chr> "Afghanistan", "Afghanistan", "Afghanistan", "Afghanis...
## $ iso2    <chr> "AF", "AF", "AF", "AF", "AF", "AF", "AF", "AF", "AF", ...
## $ iso3    <chr> "AFG", "AFG", "AFG", "AFG", "AFG", "AFG", "AFG", "AFG", "AFG"...
## $ year    <int> 1997, 1998, 1999, 2000, 2001, 2002, 2003, 2004, 2005, ...
## $ key     <chr> "new_sp_m014", "new_sp_m014", "new_sp_m014", "new_sp_m...
## $ cases   <int> 0, 30, 8, 52, 129, 90, 127, 139, 151, 193, 186, 187, 2...
```

```
who2 <- who1 %>%
  mutate(key = stringr::str_replace(key, "newrel", "new_rel"))
```

```
## Warning: package 'bindrcpp' was built under R version 3.4.3
```

```
who3 <- who2 %>%
  separate(key, c("new", "type", "sexage"), sep = "_")
who3
```

```
## # A tibble: 76,046 x 8
##   country iso2 iso3 year new type sexage cases
##   <chr>    <chr> <chr> <int> <chr> <chr> <chr> <int>
## 1 Afghanistan AF AFG 1997 new sp m014 0
## 2 Afghanistan AF AFG 1998 new sp m014 30
## 3 Afghanistan AF AFG 1999 new sp m014 8
```

```
## 4 Afghanistan AF AFG 2000 new sp m014 52
## 5 Afghanistan AF AFG 2001 new sp m014 129
## 6 Afghanistan AF AFG 2002 new sp m014 90
## 7 Afghanistan AF AFG 2003 new sp m014 127
## 8 Afghanistan AF AFG 2004 new sp m014 139
## 9 Afghanistan AF AFG 2005 new sp m014 151
## 10 Afghanistan AF AFG 2006 new sp m014 193
## # ... with 76,036 more rows
```

```
who3 %>%
  count(new)
```

```
## # A tibble: 1 x 2
##   new      n
##   <chr> <int>
## 1 new  76046
```

```
who4 <- who3 %>%
  select(-new, -iso2, -iso3)
who5 <- who4 %>%
  separate(sexage, c("sex", "age"), sep = 1)
who5
```

```
## # A tibble: 76,046 x 6
##   country      year type sex age cases
##   <chr>      <int> <chr> <chr> <chr> <int>
## 1 Afghanistan 1997 sp m 014 0
## 2 Afghanistan 1998 sp m 014 30
## 3 Afghanistan 1999 sp m 014 8
## 4 Afghanistan 2000 sp m 014 52
## 5 Afghanistan 2001 sp m 014 129
## 6 Afghanistan 2002 sp m 014 90
## 7 Afghanistan 2003 sp m 014 127
## 8 Afghanistan 2004 sp m 014 139
## 9 Afghanistan 2005 sp m 014 151
## 10 Afghanistan 2006 sp m 014 193
## # ... with 76,036 more rows
```

## 1

It is okay to treat missing values the same, and we don't lose any information by dropping them.

```
who1 %>%
  filter(cases == 0) %>%
  nrow()
```

```
## [1] 11080
```

```
gather(who, new_sp_m014:newrel_f65, key = "key", value = "cases") %>%
  group_by(country, year) %>%
  mutate(missing = is.na(cases)) %>%
  select(country, year, missing) %>%
  distinct() %>%
  group_by(country, year) %>%
  filter(n() > 1)
```

```
## # A tibble: 6,968 x 3
```

```
## # Groups:   country, year [3,484]
##   country   year missing
##   <chr>     <int> <lg1>
## 1 Afghanistan 1997 F
## 2 Afghanistan 1998 F
## 3 Afghanistan 1999 F
## 4 Afghanistan 2000 F
## 5 Afghanistan 2001 F
## 6 Afghanistan 2002 F
## 7 Afghanistan 2003 F
## 8 Afghanistan 2004 F
## 9 Afghanistan 2005 F
## 10 Afghanistan 2006 F
## # ... with 6,958 more rows
```

## 2

separate emits the warning “too few values”, and if we check the rows for keys beginning with “newrel\_”, we see that sexage is messing, and type = m014

```
who3a <- who1 %>%
  separate(key, c("new", "type", "sexage"), sep = "_")
```

```
## Warning: Expected 3 pieces. Missing pieces filled with `NA` in 2580 rows
## [73467, 73468, 73469, 73470, 73471, 73472, 73473, 73474, 73475, 73476,
## 73477, 73478, 73479, 73480, 73481, 73482, 73483, 73484, 73485, 73486, ...].
```

```
filter(who3a, new == "newrel") %>% head()
```

```
## # A tibble: 6 x 8
##   country   iso2 iso3  year new   type sexage cases
##   <chr>     <chr> <chr> <int> <chr> <chr> <chr> <int>
## 1 Afghanistan AF    AFG   2013 newrel m014 <NA>    1705
## 2 Albania     AL    ALB   2013 newrel m014 <NA>     14
## 3 Algeria     DZ    DZA   2013 newrel m014 <NA>     25
## 4 Andorra     AD    AND   2013 newrel m014 <NA>      0
## 5 Angola      AO    AGO   2013 newrel m014 <NA>    486
## 6 Anguilla    AI    AIA   2013 newrel m014 <NA>      0
```

## 3

```
select(who3, country, iso2, iso3) %>%
  distinct() %>%
  group_by(country) %>%
  filter(n() > 1)
```

```
## # A tibble: 0 x 3
## # Groups:   country [0]
## # ... with 3 variables: country <chr>, iso2 <chr>, iso3 <chr>
```

```
who5 %>%
  group_by(country, year, sex) %>%
  filter(year > 1995) %>%
  summarise(cases = sum(cases)) %>%
```

```
unite(country_sex, country, sex, remove = FALSE) %>%
  ggplot(aes(x = year, y = cases, group = country_sex, colour = sex)) +
  geom_line()
```

