

CptS 475/575: Data Science, Fall 2018

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Load the data into R, and check the first few rows for abnormalities. You will likely notice several.

```
msleep <- read.csv("https://scads.eecs.wsu.edu/wp-content/uploads/2017/10/msleep_ggplot2.csv")
```

```
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
```

```
summary(msleep)
```

```
##
##      name      genus      vore
## African elephant      : 1 Panthera : 3   carni :19
## African giant pouched rat: 1  Sperophilus: 3   herbi :32
## African striped mouse   : 1 Equus      : 2   insecti: 5
## Arctic fox              : 1 Vulpes     : 2   omni    :20
## Arctic ground squirrel  : 1 Acinonyx  : 1   NA's    : 7
## Asian elephant          : 1 Aotus      : 1
## (Other)                 :77 (Other)     :71
##
##      order      conservation sleep_total sleep_rem
## Rodentia      :22 cd          : 2   Min.    : 1.90   Min.    :0.100
## Carnivora      :12 domesticated:10   1st Qu.: 7.85   1st Qu.:0.900
## Primates       :12 en          : 4   Median :10.10   Median :1.500
## Artiodactyla: 6 lc           :27   Mean    :10.43   Mean    :1.875
## Soricomorpha: 5 nt           : 4   3rd Qu.:13.75   3rd Qu.:2.400
## Cetacea        : 3 vu          : 7   Max.    :19.90   Max.    :6.600
## (Other)        :23 NA's       :29
##
##      sleep_cycle      awake      brainwt      bodywt
## Min.    :0.1167   Min.    : 4.10   Min.    :0.00014   Min.    : 0.005
## 1st Qu.:0.1833   1st Qu.:10.25   1st Qu.:0.00290   1st Qu.: 0.174
## Median :0.3333   Median :13.90   Median :0.01240   Median : 1.670
## Mean    :0.4396   Mean    :13.57   Mean    :0.28158   Mean    :166.136
## 3rd Qu.:0.5792   3rd Qu.:16.15   3rd Qu.:0.12550   3rd Qu.: 41.750
## Max.    :1.5000   Max.    :22.10   Max.    :5.71200   Max.    :6654.000
## NA's     :51
## NA's     :27
```

Use `select()` to print the head of the columns with a title including “sleep”.

```
head(msleep)
```

```
##
##      name      genus vore      order conservation
## 1      Cheetah  Acinonyx carni   Carnivora      lc
## 2 Owl monkey   Aotus    omni    Primates    <NA>
```

```
## 3      Mountain beaver Aplodontia herbi      Rodentia      nt
## 4 Greater short-tailed shrew Blarina omni Soricomorpha      lc
## 5              Cow      Bos herbi Artiodactyla domesticated
## 6      Three-toed sloth Bradypus herbi      Pilosa      <NA>
##  sleep_total sleep_rem sleep_cycle awake brainwt bodywt
## 1      12.1      NA      NA 11.9      NA 50.000
## 2      17.0      1.8      NA 7.0 0.01550 0.480
## 3      14.4      2.4      NA 9.6      NA 1.350
## 4      14.9      2.3 0.1333333 9.1 0.00029 0.019
## 5      4.0      0.7 0.6666667 20.0 0.42300 600.000
## 6      14.4      2.2 0.7666667 9.6      NA 3.850
```

a). Use filter() to count the number of animals which weigh over 50 kilograms and sleep more than 6 hours a day.

```
filter(msleep, sleep_total >6, bodywt > 50)
```

```
##      name      genus      vore      order conservation
## 1 Gray seal Haliochoerus      carni      Carnivora      lc
## 2 Human      Homo      omni      Primates      <NA>
## 3 Chimpanzee      Pan      omni      Primates      <NA>
## 4 Tiger      Panthera      carni      Carnivora      en
## 5 Jaguar      Panthera      carni      Carnivora      nt
## 6 Lion      Panthera      carni      Carnivora      vu
## 7 Giant armadillo Priodontes insecti      Cingulata      en
## 8 Pig      Sus      omni Artiodactyla domesticated
##  sleep_total sleep_rem sleep_cycle awake brainwt bodywt
## 1      6.2      1.5      NA 17.8 0.325 85.000
## 2      8.0      1.9 1.500000 16.0 1.320 62.000
## 3      9.7      1.4 1.416667 14.3 0.440 52.200
## 4      15.8      NA      NA 8.2      NA 162.564
## 5      10.4      NA      NA 13.6 0.157 100.000
## 6      13.5      NA      NA 10.5      NA 161.499
## 7      18.1      6.1      NA 5.9 0.081 60.000
## 8      9.1      2.4 0.500000 14.9 0.180 86.250
```

b).Use piping (%>%), select() and arrange() to print the name, order, sleep time and bodyweight of the animals with the top 6 sleep times, in order of sleep time.

```
msleep %>%
  select(name, order, sleep_total,bodywt) %>%
  arrange(order, sleep_total) %>%
  filter(sleep_total >= 17.4)
```

```
##      name      order sleep_total bodywt
## 1 Big brown bat      Chiroptera      19.7 0.023
## 2 Little brown bat      Chiroptera      19.9 0.010
## 3 Long-nosed armadillo      Cingulata      17.4 3.500
## 4 Giant armadillo      Cingulata      18.1 60.000
## 5 North American Opossum Didelphimorphia      18.0 1.700
## 6 Thick-tailed opossum Didelphimorphia      19.4 0.370
```

c).Use mutate to add two new columns to the dataframe; wt_ratio with the ratio of brain size to body weight, rem_ratio with the ratio of rem sleep to sleep time. If you think they might be useful, feel free to extract more features than these, and describe what they are

```
msleep %>%
  mutate(rem_ratio = sleep_rem / sleep_total,
         wt_ratio = brainwt/bodywt ) %>%
  head
```

```
##           name      genus  vore      order conservation
## 1      Cheetah  Acinonyx carni   Carnivora          lc
## 2      Owl monkey    Aotus  omni    Primates        <NA>
## 3      Mountain beaver Aplodontia herbi   Rodentia          nt
## 4 Greater short-tailed shrew Blarina  omni Soricomorpha          lc
## 5      Cow      Bos  herbi Artiodactyla domesticated
## 6      Three-toed sloth  Bradypus herbi    Pilosa        <NA>
##  sleep_total sleep_rem sleep_cycle awake brainwt  bodywt rem_ratio
## 1      12.1      NA      NA  11.9      NA  50.000      NA
## 2      17.0      1.8      NA   7.0 0.01550   0.480 0.1058824
## 3      14.4      2.4      NA   9.6      NA   1.350 0.1666667
## 4      14.9      2.3  0.1333333   9.1 0.00029   0.019 0.1543624
## 5       4.0      0.7  0.6666667  20.0 0.42300 600.000 0.1750000
## 6      14.4      2.2  0.7666667   9.6      NA   3.850 0.1527778
##      wt_ratio
## 1      NA
## 2 0.03229167
## 3      NA
## 4 0.01526316
## 5 0.00070500
## 6      NA
```

d). Use `group_by()` and `summarize()` to display the average, min and max sleep times for each order. Remember to use `ungroup()` when you are done.

```
msleep %>%
  group_by(order) %>%
  summarise(avg_sleep = mean(sleep_total),
            min_sleep = min(sleep_total),
            max_sleep = max(sleep_total),
            total = n())
```

```
## # A tibble: 19 x 5
##   order      avg_sleep min_sleep max_sleep total
##   <fct>      <dbl>    <dbl>    <dbl> <int>
## 1 Afrosoricida    15.6      15.6     15.6     1
## 2 Artiodactyla     4.52       1.9      9.1     6
## 3 Carnivora      10.1       3.5     15.8    12
## 4 Cetacea        4.5       2.7      5.6     3
## 5 Chiroptera     19.8      19.7     19.9     2
## 6 Cingulata      17.8      17.4     18.1     2
## 7 Didelphimorphia 18.7       18      19.4     2
## 8 Diprotodontia   12.4      11.1     13.7     2
## 9 Erinaceomorpha  10.2      10.1     10.3     2
## 10 Hyracoidea     5.67       5.3      6.3     3
## 11 Lagomorpha      8.4       8.4      8.4     1
## 12 Monotremata     8.6       8.6      8.6     1
## 13 Perissodactyla  3.47       2.9      4.4     3
## 14 Pilosa        14.4      14.4     14.4     1
## 15 Primates      10.5       8       17     12
```

## 16	Proboscidea	3.6	3.3	3.9	2
## 17	Rodentia	12.5	7	16.6	22
## 18	Scandentia	8.9	8.9	8.9	1
## 19	Soricomorpha	11.1	8.4	14.9	5

ungroup(msleep)

##	name	genus	vore	order
## 1	Cheetah	Acinonyx	carni	Carnivora
## 2	Owl monkey	Aotus	omni	Primates
## 3	Mountain beaver	Aplodontia	herbi	Rodentia
## 4	Greater short-tailed shrew	Blarina	omni	Soricomorpha
## 5	Cow	Bos	herbi	Artiodactyla
## 6	Three-toed sloth	Bradypus	herbi	Pilosa
## 7	Northern fur seal	Callorhinus	carni	Carnivora
## 8	Vesper mouse	Calomys	<NA>	Rodentia
## 9	Dog	Canis	carni	Carnivora
## 10	Roe deer	Capreolus	herbi	Artiodactyla
## 11	Goat	Capri	herbi	Artiodactyla
## 12	Guinea pig	Cavis	herbi	Rodentia
## 13	Grivet	Cercopithecus	omni	Primates
## 14	Chinchilla	Chinchilla	herbi	Rodentia
## 15	Star-nosed mole	Condylura	omni	Soricomorpha
## 16	African giant pouched rat	Cricetomys	omni	Rodentia
## 17	Lesser short-tailed shrew	Cryptotis	omni	Soricomorpha
## 18	Long-nosed armadillo	Dasypus	carni	Cingulata
## 19	Tree hyrax	Dendrohyrax	herbi	Hyracoidea
## 20	North American Opossum	Didelphis	omni	Didelphimorphia
## 21	Asian elephant	Elephas	herbi	Proboscidea
## 22	Big brown bat	Eptesicus	insecti	Chiroptera
## 23	Horse	Equus	herbi	Perissodactyla
## 24	Donkey	Equus	herbi	Perissodactyla
## 25	European hedgehog	Erinaceus	omni	Erinaceomorpha
## 26	Patas monkey	Erythrocebus	omni	Primates
## 27	Western american chipmunk	Eutamias	herbi	Rodentia
## 28	Domestic cat	Felis	carni	Carnivora
## 29	Galago	Galago	omni	Primates
## 30	Giraffe	Giraffa	herbi	Artiodactyla
## 31	Pilot whale	Globicephalus	carni	Cetacea
## 32	Gray seal	Haliochoerus	carni	Carnivora
## 33	Gray hyrax	Heterohyrax	herbi	Hyracoidea
## 34	Human	Homo	omni	Primates
## 35	Mongoose lemur	Lemur	herbi	Primates
## 36	African elephant	Loxodonta	herbi	Proboscidea
## 37	Thick-tailed opossum	Lutreolina	carni	Didelphimorphia
## 38	Macaque	Macaca	omni	Primates
## 39	Mongolian gerbil	Meriones	herbi	Rodentia
## 40	Golden hamster	Mesocricetus	herbi	Rodentia
## 41	Vole	Microtus	herbi	Rodentia
## 42	House mouse	Mus	herbi	Rodentia
## 43	Little brown bat	Myotis	insecti	Chiroptera
## 44	Round-tailed muskrat	Neofiber	herbi	Rodentia
## 45	Slow loris	Nyctibeus	carni	Primates
## 46	Degu	Octodon	herbi	Rodentia
## 47	Northern grasshopper mouse	Onychomys	carni	Rodentia

## 48		Rabbit	Oryctolagus	herbi	Lagomorpha		
## 49		Sheep	Ovis	herbi	Artiodactyla		
## 50		Chimpanzee	Pan	omni	Primates		
## 51		Tiger	Panthera	carni	Carnivora		
## 52		Jaguar	Panthera	carni	Carnivora		
## 53		Lion	Panthera	carni	Carnivora		
## 54		Baboon	Papio	omni	Primates		
## 55		Desert hedgehog	Paraechinus	<NA>	Erinaceomorpha		
## 56		Potto	Perodicticus	omni	Primates		
## 57		Deer mouse	Peromyscus	<NA>	Rodentia		
## 58		Phalanger	Phalanger	<NA>	Diprotodontia		
## 59		Caspian seal	Phoca	carni	Carnivora		
## 60		Common porpoise	Phocoena	carni	Cetacea		
## 61		Potoroo	Potorous	herbi	Diprotodontia		
## 62		Giant armadillo	Priodontes	insecti	Cingulata		
## 63		Rock hyrax	Procavia	<NA>	Hyracoidea		
## 64		Laboratory rat	Rattus	herbi	Rodentia		
## 65		African striped mouse	Rhabdomys	omni	Rodentia		
## 66		Squirrel monkey	Saimiri	omni	Primates		
## 67		Eastern american mole	Scalopus	insecti	Soricomorpha		
## 68		Cotton rat	Sigmodon	herbi	Rodentia		
## 69		Mole rat	Spalax	<NA>	Rodentia		
## 70		Arctic ground squirrel	Spermophilus	herbi	Rodentia		
## 71		Thirteen-lined ground squirrel	Spermophilus	herbi	Rodentia		
## 72		Golden-mantled ground squirrel	Spermophilus	herbi	Rodentia		
## 73		Musk shrew	Suncus	<NA>	Soricomorpha		
## 74		Pig	Sus	omni	Artiodactyla		
## 75		Short-nosed echidna	Tachyglossus	insecti	Monotremata		
## 76		Eastern american chipmunk	Tamias	herbi	Rodentia		
## 77		Brazilian tapir	Tapirus	herbi	Perissodactyla		
## 78		Tenrec	Tenrec	omni	Afrosoricida		
## 79		Tree shrew	Tupaia	omni	Scandentia		
## 80		Bottle-nosed dolphin	Tursiops	carni	Cetacea		
## 81		Genet	Genetta	carni	Carnivora		
## 82		Arctic fox	Vulpes	carni	Carnivora		
## 83		Red fox	Vulpes	carni	Carnivora		
##	conservation	sleep_total	sleep_rem	sleep_cycle	awake	brainwt	bodywt
## 1	lc	12.1	NA	NA	11.90	NA	50.000
## 2	<NA>	17.0	1.8	NA	7.00	0.01550	0.480
## 3	nt	14.4	2.4	NA	9.60	NA	1.350
## 4	lc	14.9	2.3	0.1333333	9.10	0.00029	0.019
## 5	domesticated	4.0	0.7	0.6666667	20.00	0.42300	600.000
## 6	<NA>	14.4	2.2	0.7666667	9.60	NA	3.850
## 7	vu	8.7	1.4	0.3833333	15.30	NA	20.490
## 8	<NA>	7.0	NA	NA	17.00	NA	0.045
## 9	domesticated	10.1	2.9	0.3333333	13.90	0.07000	14.000
## 10	lc	3.0	NA	NA	21.00	0.09820	14.800
## 11	lc	5.3	0.6	NA	18.70	0.11500	33.500
## 12	domesticated	9.4	0.8	0.2166667	14.60	0.00550	0.728
## 13	lc	10.0	0.7	NA	14.00	NA	4.750
## 14	domesticated	12.5	1.5	0.1166667	11.50	0.00640	0.420
## 15	lc	10.3	2.2	NA	13.70	0.00100	0.060
## 16	<NA>	8.3	2.0	NA	15.70	0.00660	1.000
## 17	lc	9.1	1.4	0.1500000	14.90	0.00014	0.005

## 18	lc	17.4	3.1	0.3833333	6.60	0.01080	3.500
## 19	lc	5.3	0.5	NA	18.70	0.01230	2.950
## 20	lc	18.0	4.9	0.3333333	6.00	0.00630	1.700
## 21	en	3.9	NA	NA	20.10	4.60300	2547.000
## 22	lc	19.7	3.9	0.1166667	4.30	0.00030	0.023
## 23	domesticated	2.9	0.6	1.0000000	21.10	0.65500	521.000
## 24	domesticated	3.1	0.4	NA	20.90	0.41900	187.000
## 25	lc	10.1	3.5	0.2833333	13.90	0.00350	0.770
## 26	lc	10.9	1.1	NA	13.10	0.11500	10.000
## 27	<NA>	14.9	NA	NA	9.10	NA	0.071
## 28	domesticated	12.5	3.2	0.4166667	11.50	0.02560	3.300
## 29	<NA>	9.8	1.1	0.5500000	14.20	0.00500	0.200
## 30	cd	1.9	0.4	NA	22.10	NA	899.995
## 31	cd	2.7	0.1	NA	21.35	NA	800.000
## 32	lc	6.2	1.5	NA	17.80	0.32500	85.000
## 33	lc	6.3	0.6	NA	17.70	0.01227	2.625
## 34	<NA>	8.0	1.9	1.5000000	16.00	1.32000	62.000
## 35	vu	9.5	0.9	NA	14.50	NA	1.670
## 36	vu	3.3	NA	NA	20.70	5.71200	6654.000
## 37	lc	19.4	6.6	NA	4.60	NA	0.370
## 38	<NA>	10.1	1.2	0.7500000	13.90	0.17900	6.800
## 39	lc	14.2	1.9	NA	9.80	NA	0.053
## 40	en	14.3	3.1	0.2000000	9.70	0.00100	0.120
## 41	<NA>	12.8	NA	NA	11.20	NA	0.035
## 42	nt	12.5	1.4	0.1833333	11.50	0.00040	0.022
## 43	<NA>	19.9	2.0	0.2000000	4.10	0.00025	0.010
## 44	nt	14.6	NA	NA	9.40	NA	0.266
## 45	<NA>	11.0	NA	NA	13.00	0.01250	1.400
## 46	lc	7.7	0.9	NA	16.30	NA	0.210
## 47	lc	14.5	NA	NA	9.50	NA	0.028
## 48	domesticated	8.4	0.9	0.4166667	15.60	0.01210	2.500
## 49	domesticated	3.8	0.6	NA	20.20	0.17500	55.500
## 50	<NA>	9.7	1.4	1.4166667	14.30	0.44000	52.200
## 51	en	15.8	NA	NA	8.20	NA	162.564
## 52	nt	10.4	NA	NA	13.60	0.15700	100.000
## 53	vu	13.5	NA	NA	10.50	NA	161.499
## 54	<NA>	9.4	1.0	0.6666667	14.60	0.18000	25.235
## 55	lc	10.3	2.7	NA	13.70	0.00240	0.550
## 56	lc	11.0	NA	NA	13.00	NA	1.100
## 57	<NA>	11.5	NA	NA	12.50	NA	0.021
## 58	<NA>	13.7	1.8	NA	10.30	0.01140	1.620
## 59	vu	3.5	0.4	NA	20.50	NA	86.000
## 60	vu	5.6	NA	NA	18.45	NA	53.180
## 61	<NA>	11.1	1.5	NA	12.90	NA	1.100
## 62	en	18.1	6.1	NA	5.90	0.08100	60.000
## 63	lc	5.4	0.5	NA	18.60	0.02100	3.600
## 64	lc	13.0	2.4	0.1833333	11.00	0.00190	0.320
## 65	<NA>	8.7	NA	NA	15.30	NA	0.044
## 66	<NA>	9.6	1.4	NA	14.40	0.02000	0.743
## 67	lc	8.4	2.1	0.1666667	15.60	0.00120	0.075
## 68	<NA>	11.3	1.1	0.1500000	12.70	0.00118	0.148
## 69	<NA>	10.6	2.4	NA	13.40	0.00300	0.122
## 70	lc	16.6	NA	NA	7.40	0.00570	0.920
## 71	lc	13.8	3.4	0.2166667	10.20	0.00400	0.101

```
## 72      lc      15.9      3.0      NA  8.10      NA      0.205
## 73      <NA>      12.8      2.0  0.1833333 11.20 0.00033      0.048
## 74 domesticated      9.1      2.4  0.5000000 14.90 0.18000      86.250
## 75      <NA>      8.6      NA      NA 15.40 0.02500      4.500
## 76      <NA>      15.8      NA      NA  8.20      NA      0.112
## 77      vu      4.4      1.0  0.9000000 19.60 0.16900     207.501
## 78      <NA>      15.6      2.3      NA  8.40 0.00260      0.900
## 79      <NA>      8.9      2.6  0.2333333 15.10 0.00250      0.104
## 80      <NA>      5.2      NA      NA 18.80      NA     173.330
## 81      <NA>      6.3      1.3      NA 17.70 0.01750      2.000
## 82      <NA>      12.5      NA      NA 11.50 0.04450      3.380
## 83      <NA>      9.8      2.4  0.3500000 14.20 0.05040      4.230
```

e). Make a copy of your dataframe, and use `group_by()` and `mutate()` to impute the missing brain weights as the average `wt_ratio` for that animal's order times the animal's weight. Make a 2 second copy of your dataframe, but this time use `group_by()` and `mutate()` to impute missing brain weights with the average brain weight for that animal's order. What assumptions do these data filling methods make? Which is the best way to impute the data, or do you see a better way, and why? You may impute or remove other variables as you find appropriate. Briefly explain your decisions.

```
Firstcopy=data.frame(msleep)
head(Firstcopy)
```

```
##           name      genus vore      order conservation
## 1      Cheetah  Acinonyx  carni    Carnivora          lc
## 2      Owl monkey   Aotus  omni    Primates        <NA>
## 3  Mountain beaver Aplodontia herbi   Rodentia          nt
## 4 Greater short-tailed shrew  Blarina  omni Soricomorpha      lc
## 5      Cow      Bos  herbi Artiodactyla domesticated
## 6  Three-toed sloth  Bradypus herbi     Pilosa        <NA>
##  sleep_total sleep_rem sleep_cycle awake brainwt  bodywt
## 1      12.1      NA      NA 11.9      NA  50.000
## 2      17.0      1.8      NA  7.0 0.01550   0.480
## 3      14.4      2.4      NA  9.6      NA   1.350
## 4      14.9      2.3  0.1333333  9.1 0.00029   0.019
## 5       4.0      0.7  0.6666667 20.0 0.42300 600.000
## 6      14.4      2.2  0.7666667  9.6      NA   3.850
```

```
Firstcopy %>%
  group_by(order) %>%
  mutate(brainwt= ifelse(is.na(brainwt), mean(brainwt, na.rm=TRUE), brainwt))
```

```
## # A tibble: 83 x 11
## # Groups:   order [19]
##   name genus vore order conservation sleep_total sleep_rem sleep_cycle
##   <fct> <fct> <fct> <fct> <fct>          <dbl>      <dbl>      <dbl>
## 1 Chee~ Acin~ carni Carn~ lc              12.1      NA          NA
## 2 Owl ~ Aotus omni Prim~ <NA>              17        1.8          NA
## 3 Moun~ Aplo~ herbi Rode~ nt              14.4      2.4          NA
## 4 Grea~ Blar~ omni Sori~ lc              14.9      2.3      0.133
## 5 Cow  Bos  herbi Arti~ domesticated      4        0.7      0.667
## 6 Thre~ Brad~ herbi Pilo~ <NA>              14.4      2.2      0.767
## 7 Nort~ Call~ carni Carn~ vu              8.7      1.4      0.383
## 8 Vesp~ Calo~ <NA> Rode~ <NA>              7        NA          NA
## 9 Dog  Canis carni Carn~ domesticated     10.1      2.9      0.333
## 10 Roe ~ Capr~ herbi Arti~ lc              3        NA          NA
```

```
## # ... with 73 more rows, and 3 more variables: awake <dbl>, brainwt <dbl>,
## #   bodywt <dbl>

secondcopy=data.frame(msleep)

secondcopy %>%
  group_by(order) %>%
  mutate(bodywt= ifelse(is.na(bodywt), mean(bodywt, na.rm=TRUE), bodywt))

## # A tibble: 83 x 11
## # Groups:   order [19]
##   name genus vore order conservation sleep_total sleep_rem sleep_cycle
##   <fct> <fct> <fct> <fct> <fct>          <dbl>      <dbl>      <dbl>
## 1 Chee~ Acin~ carni Carn~ lc          12.1        NA        NA
## 2 Owl ~ Aotus omni Prim~ <NA>         17         1.8        NA
## 3 Moun~ Aplo~ herbi Rode~ nt          14.4        2.4        NA
## 4 Grea~ Blar~ omni Sori~ lc          14.9        2.3        0.133
## 5 Cow   Bos   herbi Arti~ domesticated    4         0.7        0.667
## 6 Thre~ Brad~ herbi Pilo~ <NA>         14.4        2.2        0.767
## 7 Nort~ Call~ carni Carn~ vu          8.7        1.4        0.383
## 8 Vesp~ Calo~ <NA> Rode~ <NA>         7         NA        NA
## 9 Dog   Canis carni Carn~ domesticated   10.1        2.9        0.333
## 10 Roe ~ Capr~ herbi Arti~ lc           3         NA        NA
## # ... with 73 more rows, and 3 more variables: awake <dbl>, brainwt <dbl>,
## #   bodywt <dbl>
```

Exercise 2

For this question, you will first need to read section 12.6 in the R for Data Science book, here (<http://r4ds.had.co.nz/tidy-data.html#case-study>). Grab the dataset from the tidyr package, and tidy it as shown in the case study before answering the following questions

```
readdata = read.csv("assignment 3/TB_notification.csv")
summary(readdata)
```

```
##           country      iso2      iso3      iso_numeric
## Afghanistan : 38 AD : 38 ABW : 38 Min. : 4.0
## Albania      : 38 AE : 38 AFG : 38 1st Qu.:212.0
## Algeria      : 38 AF : 38 AGO : 38 Median :430.0
## American Samoa: 38 AG : 38 AIA : 38 Mean :431.7
## Andorra      : 38 AI : 38 ALB : 38 3rd Qu.:646.0
## Angola       : 38 (Other):7842 AND : 38 Max. :894.0
## (Other)      :7842 NA's : 38 (Other):7842
## g_whoregion   year      new_sp      new_sn
## AFR:1755      Min. :1980 Min. : 0 Min. : 0.0
## AMR:1688      1st Qu.:1989 1st Qu.: 99 1st Qu.: 60.0
## EMR: 836      Median :1999 Median : 1054 Median : 521.5
## EUR:2027      Mean :1999 Mean : 9880 Mean : 8252.9
## SEA: 396      3rd Qu.:2008 3rd Qu.: 5012 3rd Qu.: 2423.5
## WPR:1368      Max. :2017 Max. :642321 Max. :932998.0
##              NA's :4166 NA's :4504
##           new_su      new_ep      new_oth      ret_rel
## Min. : 0 Min. : 0 Min. : 0.00 Min. : 0.0
## 1st Qu.: 0 1st Qu.: 46 1st Qu.: 0.00 1st Qu.: 3.0
## Median : 3 Median : 373 Median : 0.00 Median : 92.5
## Mean : 1230 Mean : 3228 Mean : 61.07 Mean : 1122.4
```


##	3rd Qu.:	205	3rd Qu.:	1843	3rd Qu.:	0.00	3rd Qu.:	442.8
##	Max.	:787338	Max.	:298831	Max.	:7342.00	Max.	:112508.0
##	NA's	:5235	NA's	:3473	NA's	:6674	NA's	:4708
##	ret_taf		ret_tad		ret_oth		newret_oth	
##	Min.	: 0.00	Min.	: 0.0	Min.	: 0	Min.	: 0.0
##	1st Qu.:	0.00	1st Qu.:	0.0	1st Qu.:	0	1st Qu.:	0.0
##	Median :	5.00	Median :	15.0	Median :	12	Median :	0.0
##	Mean :	272.86	Mean :	614.9	Mean :	1437	Mean :	207.8
##	3rd Qu.:	76.75	3rd Qu.:	125.2	3rd Qu.:	241	3rd Qu.:	2.0
##	Max.	:39840.00	Max.	:77618.0	Max.	:101832	Max.	:40659.0
##	NA's	:5852	NA's	:5838	NA's	:5828	NA's	:6487
##	new_labconf		new_clindx		ret_rel_labconf		ret_rel_clindx	
##	Min.	: 0	Min.	: 0.0	Min.	: 0	Min.	: 0.0
##	1st Qu.:	123	1st Qu.:	28.5	1st Qu.:	3	1st Qu.:	0.0
##	Median :	1154	Median :	357.0	Median :	92	Median :	1.0
##	Mean :	12656	Mean :	10288.0	Mean :	1410	Mean :	716.6
##	3rd Qu.:	5278	3rd Qu.:	2240.5	3rd Qu.:	459	3rd Qu.:	67.0
##	Max.	:817239	Max.	:599786.0	Max.	:124679	Max.	:140820.0
##	NA's	:4838	NA's	:7039	NA's	:7079	NA's	:7178
##	ret_rel_ep		ret_nrel		notif_foreign		c_newinc	
##	Min.	: 0.00	Min.	: 0.0	Min.	: 0.0	Min.	: 0
##	1st Qu.:	0.00	1st Qu.:	1.0	1st Qu.:	0.0	1st Qu.:	234
##	Median :	1.00	Median :	56.0	Median :	5.0	Median :	2162
##	Mean :	67.81	Mean :	1446.6	Mean :	304.8	Mean :	21488
##	3rd Qu.:	25.00	3rd Qu.:	413.8	3rd Qu.:	111.0	3rd Qu.:	9616
##	Max.	:2734.00	Max.	:172282.0	Max.	:9527.0	Max.	:1786681
##	NA's	:7189	NA's	:7070	NA's	:6363	NA's	:539
##	new_sp_m04		new_sp_m514		new_sp_m014		new_sp_m1524	
##	Min.	: 0.000	Min.	: 0.00	Min.	: 0.00	Min.	: 0.0
##	1st Qu.:	0.000	1st Qu.:	0.00	1st Qu.:	0.00	1st Qu.:	9.0
##	Median :	0.000	Median :	1.00	Median :	5.00	Median :	90.0
##	Mean :	8.287	Mean :	42.23	Mean :	83.77	Mean :	1016.3
##	3rd Qu.:	3.000	3rd Qu.:	13.00	3rd Qu.:	37.00	3rd Qu.:	503.5
##	Max.	:655.000	Max.	:1594.00	Max.	:5001.00	Max.	:78278.0
##	NA's	:6996	NA's	:6985	NA's	:4899	NA's	:4863
##	new_sp_m2534		new_sp_m3544		new_sp_m4554		new_sp_m5564	
##	Min.	: 0.0	Min.	: 0	Min.	: 0	Min.	: 0.0
##	1st Qu.:	14.0	1st Qu.:	13	1st Qu.:	12	1st Qu.:	8.0
##	Median :	150.5	Median :	131	Median :	102	Median :	63.0
##	Mean :	1404.7	Mean :	1317	Mean :	1105	Mean :	801.2
##	3rd Qu.:	716.0	3rd Qu.:	584	3rd Qu.:	440	3rd Qu.:	279.2
##	Max.	:84003.0	Max.	:90830	Max.	:82921	Max.	:63814.0
##	NA's	:4866	NA's	:4853	NA's	:4849	NA's	:4854
##	new_sp_m65		new_sp_mu		new_sp_f04		new_sp_f514	
##	Min.	: 0.0	Min.	: 0.00	Min.	: 0.000	Min.	: 0.0
##	1st Qu.:	8.0	1st Qu.:	0.00	1st Qu.:	0.000	1st Qu.:	0.0
##	Median :	53.0	Median :	0.00	Median :	0.000	Median :	2.0
##	Mean :	683.2	Mean :	10.85	Mean :	6.511	Mean :	59.2
##	3rd Qu.:	233.0	3rd Qu.:	0.00	3rd Qu.:	2.000	3rd Qu.:	24.0
##	Max.	:70376.0	Max.	:7417.00	Max.	:620.000	Max.	:3132.0
##	NA's	:4863	NA's	:7153	NA's	:6995	NA's	:6982
##	new_sp_f014		new_sp_f1524		new_sp_f2534		new_sp_f3544	
##	Min.	: 0.0	Min.	: 0.0	Min.	: 0.0	Min.	: 0.0
##	1st Qu.:	1.0	1st Qu.:	7.0	1st Qu.:	9.0	1st Qu.:	6.0

## Median :	7.0	Median :	66.0	Median :	84.0	Median :	57.0
## Mean :	114.4	Mean :	826.4	Mean :	917.6	Mean :	640.6
## 3rd Qu.:	51.0	3rd Qu.:	421.0	3rd Qu.:	476.5	3rd Qu.:	308.0
## Max. :	8576.0	Max. :	53975.0	Max. :	49887.0	Max. :	34698.0
## NA's :	4897	NA's :	4877	NA's :	4871	NA's :	4872
## new_sp_f4554		new_sp_f5564		new_sp_f65		new_sp_fu	
## Min. :	0.0	Min. :	0.0	Min. :	0	Min. :	0.000
## 1st Qu.:	4.0	1st Qu.:	3.0	1st Qu.:	4	1st Qu.:	0.000
## Median :	38.0	Median :	25.0	Median :	30	Median :	0.000
## Mean :	445.9	Mean :	314.0	Mean :	284	Mean :	4.137
## 3rd Qu.:	211.0	3rd Qu.:	146.8	3rd Qu.:	129	3rd Qu.:	0.000
## Max. :	23977.0	Max. :	18203.0	Max. :	21339	Max. :	2559.000
## NA's :	4867	NA's :	4876	NA's :	4874	NA's :	7155
## new_sn_m04		new_sn_m514		new_sn_m014		new_sn_m1524	
## Min. :	0.0	Min. :	0.0	Min. :	0.0	Min. :	0.0
## 1st Qu.:	0.0	1st Qu.:	0.0	1st Qu.:	1.0	1st Qu.:	2.0
## Median :	3.0	Median :	4.0	Median :	9.0	Median :	15.5
## Mean :	155.6	Mean :	144.4	Mean :	308.8	Mean :	513.0
## 3rd Qu.:	18.0	3rd Qu.:	29.0	3rd Qu.:	61.0	3rd Qu.:	102.0
## Max. :	15147.0	Max. :	8438.0	Max. :	22355.0	Max. :	60246.0
## NA's :	7115	NA's :	7116	NA's :	7025	NA's :	7040
## new_sn_m2534		new_sn_m3544		new_sn_m4554		new_sn_m5564	
## Min. :	0.0	Min. :	0.0	Min. :	0.0	Min. :	0.0
## 1st Qu.:	2.0	1st Qu.:	2.0	1st Qu.:	2.0	1st Qu.:	2.0
## Median :	23.0	Median :	19.0	Median :	19.0	Median :	16.0
## Mean :	653.7	Mean :	837.9	Mean :	520.8	Mean :	448.6
## 3rd Qu.:	135.5	3rd Qu.:	132.0	3rd Qu.:	127.5	3rd Qu.:	102.0
## Max. :	50282.0	Max. :	250051.0	Max. :	57181.0	Max. :	64972.0
## NA's :	7048	NA's :	7045	NA's :	7043	NA's :	7049
## new_sn_m65		new_sn_m15plus		new_sn_mu		new_sn_f04	
## Min. :	0.0	Min. :	0.0	Min. :	0.0	Min. :	0.0
## 1st Qu.:	2.0	1st Qu.:	15.0	1st Qu.:	0.0	1st Qu.:	0.0
## Median :	20.5	Median :	124.5	Median :	0.0	Median :	3.0
## Mean :	460.4	Mean :	3480.0	Mean :	246.7	Mean :	139.4
## 3rd Qu.:	111.8	3rd Qu.:	886.0	3rd Qu.:	0.0	3rd Qu.:	14.0
## Max. :	74282.0	Max. :	361435.0	Max. :	66885.0	Max. :	14084.0
## NA's :	7050	NA's :	7014	NA's :	7294	NA's :	7118
## new_sn_f514		new_sn_f014		new_sn_f1524		new_sn_f2534	
## Min. :	0	Min. :	0	Min. :	0.0	Min. :	0.0
## 1st Qu.:	0	1st Qu.:	1	1st Qu.:	1.0	1st Qu.:	2.0
## Median :	5	Median :	8	Median :	12.0	Median :	18.0
## Mean :	146	Mean :	292	Mean :	407.9	Mean :	466.3
## 3rd Qu.:	27	3rd Qu.:	58	3rd Qu.:	89.0	3rd Qu.:	103.2
## Max. :	7322	Max. :	21406	Max. :	35518.0	Max. :	28753.0
## NA's :	7119	NA's :	7030	NA's :	7048	NA's :	7054
## new_sn_f3544		new_sn_f4554		new_sn_f5564			
## Min. :	0.00	Min. :	0.00	Min. :	0.0		
## 1st Qu.:	1.00	1st Qu.:	1.00	1st Qu.:	1.0		
## Median :	11.00	Median :	10.00	Median :	8.0		
## Mean :	506.60	Mean :	271.16	Mean :	213.4		
## 3rd Qu.:	82.25	3rd Qu.:	76.75	3rd Qu.:	56.0		
## Max. :	148811.00	Max. :	23869.00	Max. :	26085.0		
## NA's :	7050	NA's :	7052	NA's :	7053		
## new_sn_f65		new_sn_f15plus		new_sn_fu		new_sn_sexunk04	

## Min. :	0.0	## Min. :	0	## Min. :	0.0	## Min. :	0.00
## 1st Qu.:	1.0	## 1st Qu.:	9	## 1st Qu.:	0.0	## 1st Qu.:	22.75
## Median :	13.0	## Median :	80	## Median :	0.0	## Median :	135.50
## Mean :	230.8	## Mean :	2184	## Mean :	179.4	## Mean :	467.33
## 3rd Qu.:	74.0	## 3rd Qu.:	608	## 3rd Qu.:	0.0	## 3rd Qu.:	752.75
## Max. :	29630.0	## Max. :	170327	## Max. :	47305.0	## Max. :	1667.00
## NA's :	7051	## NA's :	7026	## NA's :	7297	## NA's :	8046
## new_sn_sexunk514		## new_sn_sexunk014		## new_sn_sexunk15plus		## new_ep_m04	
## Min. :	0.0	## Min. :	0	## Min. :	0.0	## Min. :	0
## 1st Qu.:	31.0	## 1st Qu.:	66	## 1st Qu.:	639.8	## 1st Qu.:	0
## Median :	112.5	## Median :	328	## Median :	1445.5	## Median :	0
## Mean :	517.2	## Mean :	3010	## Mean :	20955.2	## Mean :	0
## 3rd Qu.:	598.8	## 3rd Qu.:	2133	## 3rd Qu.:	4650.8	## 3rd Qu.:	0
## Max. :	4438.0	## Max. :	36673	## Max. :	303530.0	## Max. :	0
## NA's :	8046	## NA's :	8035	## NA's :	8036	## NA's :	8062
## new_ep_m514		## new_ep_m014		## new_ep_m1524		## new_ep_m2534	
## Min. :	0.00	## Min. :	0.0	## Min. :	0.0	## Min. :	0.0
## 1st Qu.:	0.00	## 1st Qu.:	0.0	## 1st Qu.:	1.0	## 1st Qu.:	1.0
## Median :	3.00	## Median :	6.0	## Median :	11.0	## Median :	13.0
## Mean :	82.31	## Mean :	128.6	## Mean :	158.3	## Mean :	201.2
## 3rd Qu.:	32.00	## 3rd Qu.:	56.5	## 3rd Qu.:	88.0	## 3rd Qu.:	124.0
## Max. :	4369.00	## Max. :	7869.0	## Max. :	8558.0	## Max. :	11843.0
## NA's :	7122	## NA's :	7032	## NA's :	7044	## NA's :	7050
## new_ep_m3544		## new_ep_m4554		## new_ep_m5564		## new_ep_m65	
## Min. :	0.00	## Min. :	0.00	## Min. :	0.00	## Min. :	0.00
## 1st Qu.:	1.00	## 1st Qu.:	1.00	## 1st Qu.:	1.00	## 1st Qu.:	1.00
## Median :	10.50	## Median :	8.50	## Median :	7.00	## Median :	10.00
## Mean :	272.73	## Mean :	108.12	## Mean :	72.17	## Mean :	78.94
## 3rd Qu.:	91.25	## 3rd Qu.:	63.25	## 3rd Qu.:	46.50	## 3rd Qu.:	55.00
## Max. :	105825.00	## Max. :	5875.00	## Max. :	3957.00	## Max. :	3061.00
## NA's :	7046	## NA's :	7050	## NA's :	7055	## NA's :	7052
## new_ep_m15plus		## new_ep_mu		## new_ep_f04		## new_ep_f514	
## Min. :	0.0	## Min. :	0.00	## Min. :	0.00	## Min. :	0.00
## 1st Qu.:	7.0	## 1st Qu.:	0.00	## 1st Qu.:	0.00	## 1st Qu.:	0.00
## Median :	64.0	## Median :	0.00	## Median :	1.00	## Median :	3.00
## Mean :	939.5	## Mean :	46.04	## Mean :	34.31	## Mean :	76.49
## 3rd Qu.:	576.5	## 3rd Qu.:	0.00	## 3rd Qu.:	14.00	## 3rd Qu.:	28.00
## Max. :	105825.0	## Max. :	16676.00	## Max. :	3300.00	## Max. :	4055.00
## NA's :	7019	## NA's :	7290	## NA's :	7124	## NA's :	7124
## new_ep_f014		## new_ep_f1524		## new_ep_f2534		## new_ep_f3544	
## Min. :	0.00	## Min. :	0.0	## Min. :	0.0	## Min. :	0.0
## 1st Qu.:	0.00	## 1st Qu.:	1.0	## 1st Qu.:	1.0	## 1st Qu.:	1.0
## Median :	5.00	## Median :	9.0	## Median :	12.0	## Median :	9.0
## Mean :	112.89	## Mean :	149.2	## Mean :	189.5	## Mean :	241.7
## 3rd Qu.:	50.25	## 3rd Qu.:	78.0	## 3rd Qu.:	95.0	## 3rd Qu.:	77.0
## Max. :	6960.00	## Max. :	7866.0	## Max. :	10759.0	## Max. :	101015.0
## NA's :	7038	## NA's :	7049	## NA's :	7049	## NA's :	7049
## new_ep_f4554		## new_ep_f5564		## new_ep_f65		## new_ep_f15plus	
## Min. :	0.00	## Min. :	0.00	## Min. :	0.00	## Min. :	0.0
## 1st Qu.:	1.00	## 1st Qu.:	1.00	## 1st Qu.:	0.00	## 1st Qu.:	6.0
## Median :	8.00	## Median :	6.00	## Median :	10.00	## Median :	58.0
## Mean :	93.78	## Mean :	63.04	## Mean :	72.31	## Mean :	863.8
## 3rd Qu.:	56.00	## 3rd Qu.:	42.00	## 3rd Qu.:	51.00	## 3rd Qu.:	463.0
## Max. :	6759.00	## Max. :	4684.00	## Max. :	2548.00	## Max. :	101015.0

##	NA's :7053	NA's :7053	NA's :7056	NA's :7024
##	new_ep_fu	new_ep_sexunk04	new_ep_sexunk514	new_ep_sexunk014
##	Min. : 0.00	Min. : 0.0	Min. : 0.0	Min. : 0.0
##	1st Qu.: 0.00	1st Qu.: 16.0	1st Qu.: 35.0	1st Qu.: 74.5
##	Median : 0.00	Median : 66.0	Median : 157.0	Median : 383.0
##	Mean : 40.81	Mean : 455.4	Mean : 911.4	Mean : 3003.1
##	3rd Qu.: 0.00	3rd Qu.: 694.0	3rd Qu.: 774.0	3rd Qu.: 1454.0
##	Max. :21246.00	Max. :2604.0	Max. :5376.0	Max. :34062.0
##	NA's :7297	NA's :8045	NA's :8045	NA's :8035
##	new_ep_sexunk15plus	new_ep_sexunkageunk	rel_in_agesex_flg	
##	Min. : 0	Min. :0	Min. :0.00	
##	1st Qu.: 548	1st Qu.:0	1st Qu.:1.00	
##	Median : 1119	Median :0	Median :1.00	
##	Mean : 14838	Mean :0	Mean :0.82	
##	3rd Qu.: 4771	3rd Qu.:0	3rd Qu.:1.00	
##	Max. :200528	Max. :0	Max. :1.00	
##	NA's :8036	NA's :8062	NA's :7051	
##	newrel_m04	newrel_m514	newrel_m014	newrel_m1524
##	Min. : 0.0	Min. : 0.0	Min. : 0.0	Min. : 0.0
##	1st Qu.: 1.0	1st Qu.: 2.0	1st Qu.: 4.0	1st Qu.: 14.0
##	Median : 12.0	Median : 23.0	Median : 36.5	Median : 162.5
##	Mean : 375.3	Mean : 616.8	Mean : 956.0	Mean : 2530.9
##	3rd Qu.: 101.2	3rd Qu.: 157.5	3rd Qu.: 247.5	3rd Qu.: 883.2
##	Max. :15727.0	Max. :38668.0	Max. :52377.0	Max. :197736.0
##	NA's :7142	NA's :7143	NA's :7098	NA's :7126
##	newrel_m2534	newrel_m3544	newrel_m4554	
##	Min. : 0.00	Min. : 0	Min. : 0.0	
##	1st Qu.: 24.75	1st Qu.: 21	1st Qu.: 20.0	
##	Median : 249.00	Median : 226	Median : 208.5	
##	Mean : 3269.35	Mean : 3181	Mean : 3010.7	
##	3rd Qu.: 1148.25	3rd Qu.: 952	3rd Qu.: 805.5	
##	Max. :208747.00	Max. :219012	Max. :206433.0	
##	NA's :7126	NA's :7125	NA's :7124	
##	newrel_m5564	newrel_m65	newrel_m15plus	
##	Min. : 0.0	Min. : 0.0	Min. : 0	
##	1st Qu.: 16.0	1st Qu.: 13.5	1st Qu.: 136	
##	Median : 158.0	Median : 128.0	Median : 1373	
##	Mean : 2459.7	Mean : 2167.0	Mean : 16502	
##	3rd Qu.: 593.8	3rd Qu.: 576.0	3rd Qu.: 5331	
##	Max. :168943.0	Max. :127287.0	Max. :1109208	
##	NA's :7124	NA's :7123	NA's :7085	
##	newrel_mu	newrel_f04	newrel_f514	newrel_f014
##	Min. : 0.00	Min. : 0.0	Min. : 0.0	Min. : 0.0
##	1st Qu.: 0.00	1st Qu.: 1.0	1st Qu.: 2.0	1st Qu.: 4.0
##	Median : 0.00	Median : 10.0	Median : 24.0	Median : 39.0
##	Mean : 98.03	Mean : 302.3	Mean : 654.9	Mean : 921.3
##	3rd Qu.: 0.00	3rd Qu.: 83.0	3rd Qu.: 159.0	3rd Qu.: 230.5
##	Max. :23928.00	Max. :11212.0	Max. :47757.0	Max. :55828.0
##	NA's :7231	NA's :7145	NA's :7144	NA's :7099
##	newrel_f1524	newrel_f2534	newrel_f3544	newrel_f4554
##	Min. : 0	Min. : 0.0	Min. : 0.00	Min. : 0.0
##	1st Qu.: 11	1st Qu.: 15.0	1st Qu.: 10.25	1st Qu.: 9.0
##	Median : 119	Median : 172.0	Median : 124.50	Median : 93.0
##	Mean : 2070	Mean : 2217.2	Mean : 1682.52	Mean : 1360.4

## 3rd Qu.:	670	3rd Qu.:	780.5	3rd Qu.:	579.50	3rd Qu.:	450.5
## Max.:	:176341	Max.:	:141461.0	Max.:	:95103.00	Max.:	:70606.0
## NA's:	:7125	NA's:	:7123	NA's:	:7124	NA's:	:7123
## newrel_f5564		newrel_f65		newrel_f15plus		newrel_fu	
## Min.:	: 0	Min.:	: 0.0	Min.:	: 0.0	Min.:	: 0.00
## 1st Qu.:	: 8	1st Qu.:	: 9.0	1st Qu.:	: 72.5	1st Qu.:	: 0.00
## Median:	: 72	Median:	: 78.0	Median:	: 766.0	Median:	: 0.00
## Mean:	: 1059	Mean:	: 1015.2	Mean:	: 9470.0	Mean:	: 64.72
## 3rd Qu.:	: 339	3rd Qu.:	: 400.2	3rd Qu.:	: 3568.5	3rd Qu.:	: 0.00
## Max.:	:54259	Max.:	:53551.0	Max.:	:571905.0	Max.:	:14494.00
## NA's:	:7124	NA's:	:7124	NA's:	:7083	NA's:	:7237
## newrel_sexunk04		newrel_sexunk514		newrel_sexunk014		newrel_sexunk15plus	
## Min.:	: 0.0	Min.:	: 0.0	Min.:	: 0	Min.:	: 0
## 1st Qu.:	: 0.0	1st Qu.:	: 0.0	1st Qu.:	: 0	1st Qu.:	: 0
## Median:	: 0.0	Median:	: 0.0	Median:	: 0	Median:	: 0
## Mean:	: 431.6	Mean:	: 506.1	Mean:	: 3657	Mean:	: 55454
## 3rd Qu.:	: 229.2	3rd Qu.:	: 347.2	3rd Qu.:	: 641	3rd Qu.:	: 7094
## Max.:	:3239.0	Max.:	:3720.0	Max.:	:64726	Max.:	:1179179
## NA's:	:8050	NA's:	:8050	NA's:	:8047	NA's:	:8047
## newrel_sexunkageunk		rdx_data_available		newinc_rdx		rdxsurvey_newinc	
## Min.:	: 0	Min.:	: 0.00	Min.:	: 0	Min.:	: 14.0
## 1st Qu.:	: 0	1st Qu.:	: 0.00	1st Qu.:	: 35	1st Qu.:	: 200.8
## Median:	: 0	Median:	:60.00	Median:	: 236	Median:	: 870.5
## Mean:	: 1232	Mean:	:37.58	Mean:	: 7110	Mean:	: 4517.5
## 3rd Qu.:	: 301	3rd Qu.:	:60.00	3rd Qu.:	: 1206	3rd Qu.:	: 5187.2
## Max.:	:11716	Max.:	:61.00	Max.:	:720051	Max.:	:16315.0
## NA's:	:8050	NA's:	:7471	NA's:	:7709	NA's:	:8066
## rdxsurvey_newinc_rdx		rdst_new		rdst_ret		rdst_unk	
## Min.:	: 14.0	Min.:	: 0	Min.:	: 0	Min.:	: 0.0
## 1st Qu.:	: 74.0	1st Qu.:	: 13	1st Qu.:	: 3	1st Qu.:	: 0.0
## Median:	: 561.5	Median:	: 233	Median:	: 69	Median:	: 0.0
## Mean:	:1263.2	Mean:	: 3978	Mean:	: 2056	Mean:	: 1339.6
## 3rd Qu.:	:1750.8	3rd Qu.:	: 1291	3rd Qu.:	: 461	3rd Qu.:	: 31.5
## Max.:	:3916.0	Max.:	:537180	Max.:	:283400	Max.:	:218231.0
## NA's:	:8066	NA's:	:7173	NA's:	:7184	NA's:	:7219
## conf_rrmdr		conf_mdr		rr_sldst		all_conf_xdr	
## Min.:	: 0.0	Min.:	: 0.00	Min.:	: 0.00	Min.:	: 0.00
## 1st Qu.:	: 1.0	1st Qu.:	: 0.00	1st Qu.:	: 0.00	1st Qu.:	: 0.00
## Median:	: 27.0	Median:	: 12.50	Median:	: 6.00	Median:	: 0.00
## Mean:	: 724.9	Mean:	: 393.01	Mean:	: 343.57	Mean:	: 46.05
## 3rd Qu.:	: 178.5	3rd Qu.:	: 87.75	3rd Qu.:	: 58.75	3rd Qu.:	: 3.00
## Max.:	:39009.0	Max.:	:25971.00	Max.:	:26832.00	Max.:	:3661.00
## NA's:	:7286	NA's:	:6528	NA's:	:7524	NA's:	:7509
## unconf_rrmdr_tx		conf_rrmdr_tx		unconf_mdr_tx		conf_mdr_tx	
## Min.:	: 0.00	Min.:	: 0.0	Min.:	: 0.00	Min.:	: 0.0
## 1st Qu.:	: 0.00	1st Qu.:	: 1.0	1st Qu.:	: 0.00	1st Qu.:	: 0.0
## Median:	: 0.00	Median:	: 19.0	Median:	: 0.00	Median:	: 7.0
## Mean:	: 28.45	Mean:	: 623.5	Mean:	: 36.74	Mean:	: 277.4
## 3rd Qu.:	: 0.00	3rd Qu.:	: 135.0	3rd Qu.:	: 1.00	3rd Qu.:	: 62.0
## Max.:	:5301.00	Max.:	:35950.0	Max.:	:3344.00	Max.:	:21093.0
## NA's:	:7328	NA's:	:7294	NA's:	:7400	NA's:	:7030
## conf_xdr_tx		mdrxdr_bdq_used		mdrxdr_bdq_tx		mdrxdr_dlm_used	
## Min.:	: 0.00	Min.:	:0.000	Min.:	: 0.0	Min.:	:0.000
## 1st Qu.:	: 0.00	1st Qu.:	:0.000	1st Qu.:	: 1.0	1st Qu.:	:0.000

```

## Median : 0.00 Median :0.000 Median : 6.0 Median :0.000
## Mean : 33.88 Mean :0.324 Mean : 128.5 Mean :0.247
## 3rd Qu.: 1.00 3rd Qu.:0.000 3rd Qu.: 22.0 3rd Qu.:0.000
## Max. :2882.00 Max. :3.000 Max. :8240.0 Max. :3.000
## NA's :6774 NA's :7285 NA's :7928 NA's :7685
## mdrxdr_dlm_tx mdx_shortreg_used mdx_shortreg_tx
## Min. : 0.00 Min. :0.000 Min. : 0.00
## 1st Qu.: 2.00 1st Qu.:0.000 1st Qu.: 5.75
## Median : 8.00 Median :0.000 Median : 33.00
## Mean : 27.77 Mean :0.354 Mean : 116.92
## 3rd Qu.: 40.50 3rd Qu.:1.000 3rd Qu.: 80.00
## Max. :140.00 Max. :3.000 Max. :3474.00
## NA's :8031 NA's :7486 NA's :7950
## mdx_tx_adverse_events newrel_tbhiv_flg newrel_hivtest newrel_hivpos
## Min. : 0.00 Min. :0.000 Min. : 0 Min. : 0
## 1st Qu.: 0.00 1st Qu.:0.000 1st Qu.: 107 1st Qu.: 2
## Median : 0.00 Median :1.000 Median : 1894 Median : 67
## Mean : 72.31 Mean :0.702 Mean : 19310 Mean : 2618
## 3rd Qu.: 9.00 3rd Qu.:1.000 3rd Qu.: 8428 3rd Qu.: 724
## Max. :3635.00 Max. :1.000 Max. :1271416 Max. :157505
## NA's :7653 NA's :7691 NA's :7514 NA's :7519
## newrel_art hivtest hivtest_pos
## Min. : 0.0 Min. : 0 Min. : 0.0
## 1st Qu.: 1.0 1st Qu.: 32 1st Qu.: 1.0
## Median : 67.5 Median : 482 Median : 28.0
## Mean : 2350.8 Mean : 9528 Mean : 2091.6
## 3rd Qu.: 581.8 3rd Qu.: 4175 3rd Qu.: 413.5
## Max. :133116.0 Max. :1034712 Max. :211128.0
## NA's :7572 NA's :6019 NA's :6048
## hiv_cpt hiv_art hiv_tbscr
## Min. : 0.0 Min. : 0.0 Min. : 0.0
## 1st Qu.: 0.0 1st Qu.: 0.0 1st Qu.: 18.2
## Median : 3.0 Median : 8.0 Median : 317.0
## Mean : 2029.7 Mean : 1286.3 Mean : 35321.0
## 3rd Qu.: 194.8 3rd Qu.: 190.5 3rd Qu.: 4697.8
## Max. :161561.0 Max. :141755.0 Max. :1324386.0
## NA's :6514 NA's :6419 NA's :7312
## hiv_reg hiv_ipt hiv_reg_new hiv_ipt_reg_all
## Min. : 0 Min. : 0 Min. : 0 Min. : 0
## 1st Qu.: 114 1st Qu.: 0 1st Qu.: 164 1st Qu.: 110
## Median : 1429 Median : 11 Median : 986 Median : 1158
## Mean : 65784 Mean : 6011 Mean : 33983 Mean : 39795
## 3rd Qu.: 13659 3rd Qu.: 571 3rd Qu.: 5804 3rd Qu.: 41156
## Max. :4277683 Max. :551787 Max. :1091549 Max. :170022
## NA's :7145 NA's :7167 NA's :7659 NA's :8062
## hiv_tbdetect hiv_reg_new2
## Min. : 9 Min. : 0.0
## 1st Qu.: 1086 1st Qu.: 166.2
## Median : 3916 Median : 893.0
## Mean : 207674 Mean : 24412.9
## 3rd Qu.: 50609 3rd Qu.: 5602.2
## Max. :1480908 Max. :1091549.0
## NA's :8062 NA's :7666

```

a) Explain why this line `> mutate(key = stringr::str_replace(key, "newrel", "new_rel"))` is necessary to properly tidy the data.

This dataset contains 1). It looks like country, iso2 and iso3 are three variables that redundantly specify the country. 2). We don't know about what all other columns are yet, but given the structure in the variables name (new_sp_m014, new_ep_m014, new_ep_f014) these are likely to be values, not variables.

In this dataset, we need to make some minor changes to fix the format of the columns name because the names are slightly inconsistent. As we seen in the statement we have newrel instead of new_rel (its difficult to spot this here but if we don't fix it we will get the errors in subsequent steps). So, using the idea of replacing the characters "newrel" with "new_rel". This makes all variable names consistent.

What happens if you neglect the mutate() step?

First Solution We can neglect the mutate step only if we know that all cases are new and we just parse the case type after the 3rd character. But we may not know that so better to mutate.

Second Solution

The separate() function emits the warning "too few values". If we check the rows for keys beginning with "newrel_", we see that sexage is missing, and type = m014.

- b) How many entries are removed from the dataset when you set na.rm to true in the gather command (in this dataset). How else could those NA values be handled? Among these options, which do you think is the best way to handle those missing values for this dataset, and why?

How many entries are removed from the dataset when you set na.rm to true in the gather command (in this dataset)

To give this question answer, i would need to know more about the data generation process. There are zero's in the data, which means they may explicitly be indicating no cases. To get the zero's in the dataset, below is the r command.

How else could those NA values be handled? Among these options, which do you think is the best way to handle those missing values for this dataset, and why?

There are Two R functions which deal with the NA values using Fill argument.

1). In Spread(), all NA values are replaced by the fill value. The fill argument only takes in one value. 2). In complete(), all NA values are under different variables can be replaced by different values. The fill argument takes in a list that specifies the values to replace NA for different variables.

Considering the best way to handle missing values for this dataset is using Gather() and Spread() function because we have the count for the individuals columns who has TRUE(NA) and False(value) counts. Now, these missing value could be informative. After analysing the dataset, I have found that most countries have loads of missing values ! we can decide to remove all the missing values from dataset using readdata very easily with na.omit(). In the following commands, I showed the whole process for getting the NA values and omitting the NA values.

- c) Explain the difference between an explicit and implicit missing value, in general. Can you find any implicit missing values in this dataset, if so where?

In the dataset, a value can be missing in the two possible ways. 1). Explicitly which means dataset flagged with "NA" values which we have in this dataset as i showed in the previous example. 2). Implicitly which means simply nothing present in the dataset. for example, it could be one or more empty row or has zero in the country column in this dataset.

Implicitly missing values in this dataset

- d) Looking at the features (country, year, var, sex, age, cases) in the tidied data, are they all appropriately typed? Are there any features you think would be better suited as a different type? Why or why not?

- e) Explain in your own words what a gather operation is, and give an example of a situation when it might be useful. Do the same for spread.

Gather operation will take multiple columns and collapse them into key-value pairs, duplicating all other column needed.

Spread operation function spreads a key-value pair across multiple columns.

- f) Generate an informative visualization, which shows something about the data. Give a brief description of what it shows, and why you thought it was interesting.