

Notebook3

Managing Data Frames with the dplyr package

```
#install.packages("dplyr")  
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

select()

- The `select()` function can be used to select columns of a data frame.

```
df = read.csv("data/iris.txt")  
head(df)
```

	Sepal.length	Sepal.width	Petal.length	Petal.width	Class
1	5.1	3.5	1.4	0.2	Iris-setosa
2	4.9	3.0	1.4	0.2	Iris-setosa
3	4.7	3.2	1.3	0.2	Iris-setosa
4	4.6	3.1	1.5	0.2	Iris-setosa
5	5.0	3.6	1.4	0.2	Iris-setosa
6	5.4	3.9	1.7	0.4	Iris-setosa

```
head(df[2:4])
```

	Sepal.width	Petal.length	Petal.width
1	3.5	1.4	0.2
2	3.0	1.4	0.2
3	3.2	1.3	0.2
4	3.1	1.5	0.2
5	3.6	1.4	0.2
6	3.9	1.7	0.4

```
#head(select(df, (2:4)))  
head(select(df, (Sepal.length:Petal.length)))
```

	Sepal.length	Sepal.width	Petal.length
1	5.1	3.5	1.4
2	4.9	3.0	1.4
3	4.7	3.2	1.3
4	4.6	3.1	1.5
5	5.0	3.6	1.4
6	5.4	3.9	1.7

- You can also omit variables using the `select()` function by using the negative sign

```
head(select(df, -(Sepal.length:Petal.width)))
```

	Class
1	Iris-setosa
2	Iris-setosa
3	Iris-setosa
4	Iris-setosa
5	Iris-setosa
6	Iris-setosa

The `select()` function also allows a special syntax that allows you to specify variable names based on patterns (`ends_with`, `start_with`, `contains`).

```
head(select(df, starts_with("p")))
```

	Petal.length	Petal.width
1	1.4	0.2
2	1.4	0.2
3	1.3	0.2
4	1.5	0.2
5	1.4	0.2
6	1.7	0.4

```
head(select(df, ends_with("h")))
```

	Sepal.length	Sepal.width	Petal.length	Petal.width
1	5.1	3.5	1.4	0.2
2	4.9	3.0	1.4	0.2
3	4.7	3.2	1.3	0.2
4	4.6	3.1	1.5	0.2
5	5.0	3.6	1.4	0.2
6	5.4	3.9	1.7	0.4

filter()

- The filter() function is used to extract subsets of rows from a data frame.

```
head(filter(df, Sepal.length>5))
```

	Sepal.length	Sepal.width	Petal.length	Petal.width	Class
1	5.1	3.5	1.4	0.2	Iris-setosa
2	5.4	3.9	1.7	0.4	Iris-setosa
3	5.4	3.7	1.5	0.2	Iris-setosa
4	5.8	4.0	1.2	0.2	Iris-setosa
5	5.7	4.4	1.5	0.4	Iris-setosa
6	5.4	3.9	1.3	0.4	Iris-setosa

arrange()

- The arrange() function is used to reorder rows of a data frame according to one of the variables/- columns.

```
head(arrange(df, Sepal.width))
```

	Sepal.length	Sepal.width	Petal.length	Petal.width	Class
1	5.0	2.0	3.5	1.0	Iris-versicolor
2	6.0	2.2	4.0	1.0	Iris-versicolor
3	6.2	2.2	4.5	1.5	Iris-versicolor
4	6.0	2.2	5.0	1.5	Iris-virginica
5	4.5	2.3	1.3	0.3	Iris-setosa
6	5.5	2.3	4.0	1.3	Iris-versicolor

```
head(arrange(df, desc(Sepal.width)))
```

	Sepal.length	Sepal.width	Petal.length	Petal.width	Class
1	5.7	4.4	1.5	0.4	Iris-setosa
2	5.5	4.2	1.4	0.2	Iris-setosa
3	5.2	4.1	1.5	0.1	Iris-setosa
4	5.8	4.0	1.2	0.2	Iris-setosa
5	5.4	3.9	1.7	0.4	Iris-setosa
6	5.4	3.9	1.3	0.4	Iris-setosa

rename()

Renaming a variable in a data frame in R.

```
head(rename(df, s1 = Sepal.width))
```

	Sepal.length	s1	Petal.length	Petal.width	Class
1	5.1	3.5	1.4	0.2	Iris-setosa
2	4.9	3.0	1.4	0.2	Iris-setosa
3	4.7	3.2	1.3	0.2	Iris-setosa
4	4.6	3.1	1.5	0.2	Iris-setosa
5	5.0	3.6	1.4	0.2	Iris-setosa
6	5.4	3.9	1.7	0.4	Iris-setosa

mutate()

- The mutate() function exists to compute transformations of variables in a data frame.
- to create new variables that are derived from existing variables

```
head(mutate(df, S1 = Sepal.width - mean(Sepal.width)))
```

	Sepal.length	Sepal.width	Petal.length	Petal.width	Class	S1
1	5.1	3.5	1.4	0.2	Iris-setosa	0.44266667
2	4.9	3.0	1.4	0.2	Iris-setosa	-0.05733333
3	4.7	3.2	1.3	0.2	Iris-setosa	0.14266667
4	4.6	3.1	1.5	0.2	Iris-setosa	0.04266667
5	5.0	3.6	1.4	0.2	Iris-setosa	0.54266667
6	5.4	3.9	1.7	0.4	Iris-setosa	0.84266667

```
head(transmute(df, S1 = Sepal.width - mean(Sepal.width)))
```

	S1
1	0.44266667
2	-0.05733333
3	0.14266667
4	0.04266667
5	0.54266667
6	0.84266667

%>%

The pipeline operator `%>%` is very handy for stringing together multiple dplyr functions in a sequence of operations

third(second(first(x)))

first(x) %>% second %>% third

```
head(df)
```

	Sepal.length	Sepal.width	Petal.length	Petal.width	Class
1	5.1	3.5	1.4	0.2	Iris-setosa
2	4.9	3.0	1.4	0.2	Iris-setosa
3	4.7	3.2	1.3	0.2	Iris-setosa
4	4.6	3.1	1.5	0.2	Iris-setosa
5	5.0	3.6	1.4	0.2	Iris-setosa
6	5.4	3.9	1.7	0.4	Iris-setosa

```
df1 <- df %>%
  filter(df[["Class"]] == "Iris-setosa") %>%
  select(c('Class', 'Sepal.length'))
```

```
head(df1)
```

```
      Class Sepal.length
1 Iris-setosa      5.1
2 Iris-setosa      4.9
3 Iris-setosa      4.7
4 Iris-setosa      4.6
5 Iris-setosa      5.0
6 Iris-setosa      5.4
```

group_by(), summarize()

```
group_by(df, Class)
```

```
# A tibble: 150 x 5
# Groups:   Class [3]
  Sepal.length Sepal.width Petal.length Petal.width Class
      <dbl>      <dbl>      <dbl>      <dbl> <chr>
1      5.1        3.5        1.4        0.2 Iris-setosa
2      4.9         3         1.4        0.2 Iris-setosa
3      4.7        3.2        1.3        0.2 Iris-setosa
4      4.6        3.1        1.5        0.2 Iris-setosa
5      5          3.6        1.4        0.2 Iris-setosa
6      5.4        3.9        1.7        0.4 Iris-setosa
7      4.6        3.4        1.4        0.3 Iris-setosa
8      5          3.4        1.5        0.2 Iris-setosa
9      4.4        2.9        1.4        0.2 Iris-setosa
10     4.9        3.1        1.5        0.1 Iris-setosa
# i 140 more rows
```

- Group_by() function alone will not give any output. It should be followed by summarise() function with an appropriate action to perform.

https://media.geeksforgeeks.org/wp-content/cdn-plloads/20210802175750/Sample_Superstore.csv

```
df = read.csv("data/Sample_Superstore.csv");
head(df)
```

```
Row.ID      State Region      Category Sub.Category   Sales   Profit
1      1  Kentucky  South    Furniture   Bookcases 261.9600 41.9136
```

2	2	Kentucky	South	Furniture	Chairs	731.9400	219.5820
3	3	California	West	Office Supplies	Labels	14.6200	6.8714
4	4	Florida	South	Furniture	Tables	957.5775	-383.0310
5	5	Florida	South	Office Supplies	Storage	22.3680	2.5164
6	6	California	West	Furniture	Furnishings	48.8600	14.1694

```
df_grp_region = df %>% group_by(Region) %>%
  summarise(total_sales = sum(Sales),
            .groups = 'drop')
```

```
df_grp_region
```

```
# A tibble: 4 x 3
  Region total_sales total_profits
  <chr>      <dbl>      <dbl>
1 Central    45502.      -850.
2 East       33798.      1712.
3 South      12344.      -424.
4 West       37782.      4925.
```

Grouping multiple columns

```
df_grp_reg_cat = df %>% group_by(Region, Category) %>%
df_grp_reg_cat
```

```
# A tibble: 12 x 4
  Region Category total_Sales total_Profit
  <chr>   <chr>      <dbl>      <dbl>
1 Central Furniture    13596.     -688.
2 Central Office Supplies  7762.      318.
3 Central Technology   24144.    -479.
4 East    Furniture    11658.   -1784.
5 East    Office Supplies 12794.    2537.
6 East    Technology    9346.     959.
7 South   Furniture    6329.    -552.
8 South   Office Supplies  3401.    -198.
9 South   Technology    2614.     326.
10 West   Furniture    14334.    -93.9
11 West   Office Supplies  7510.    1575.
12 West   Technology   15938.   3444.
```

```
df %>% group_by(Region) %>%
  summarise(n = n(), .groups = "keep") %>%
  group_vars()
```

```
[1] "Region"
```

```
df %>% group_by(Region) %>%
  summarise(n = n(), .groups = "drop") %>%
  group_vars()
```

```
character(0)
```

```
grp <- df %>% group_by(Region)
grp %>% group_keys()
```

```
# A tibble: 4 x 1
  Region
  <chr>
1 Central
2 East
3 South
4 West
```

```
grp1 <- df %>% group_by(Region, Category)
grp1 %>% group_keys()
```

```
# A tibble: 12 x 2
  Region Category
  <chr>   <chr>
1 Central Furniture
2 Central Office Supplies
3 Central Technology
4 East    Furniture
5 East    Office Supplies
6 East    Technology
7 South   Furniture
8 South   Office Supplies
```



```

9 South    Technology
10 West    Furniture
11 West    Office Supplies
12 West    Technology

```

```

# You can see which group each row belongs
grp %>% group_indices()

```

```

[1] 3 3 4 3 3 4 4 4 4 4 4 4 3 4 1 1 1 4 4 4 4 1 1 2 4 4 4 2 2 2 2 2 2 2 1 1 1
[38] 1 1 1 1 1 4 3 1 1 1 2 2 1 1 1 1 2 2 2 2 2 2 2 2 4 4 4 4 1 4 4 3 2 1 3 3
[75] 3 1 1 1 1 3 3 4 4 3 1 3 1 1 1 4 4 4 1 1 1 4 2 4 1 1 1 1 1 4 4 4 3 3 3 1 2
[112] 1 1 2 2 2 2 4 3 2 2 2 2 2 1 1 4 4 4 4 2 2 2 4 4 4 4 4 4 4 2 4 4 4 1 4 2 1
[149] 1 1 1 4 4 4 4 4 4 4 1 1 4 2 4 4 1 1 1 1 1 1 4 4 4 1 1 1 2 2 2 1 1 3 3 3
[186] 2 4 1 1 2 2 2 2 2 4 2 2 2 2 2 2 2 4 4 1 3 4 1 1 1 1 1 1 2 2 2 2 2 4 4 2 2 2
[223] 2 2 2 1 1 1 3 3 3 3 3 3 3 3 4 1 1 1 1 1 1 1 1 1 1 1 1 1 4 4 4 2 2 1 1 1 1 2
[260] 2 2 1 1 1 1 4 3 2 2 2 4 4 4 4 4 4 2 2 2 2 1 1 4 4 4 4 3 3 3 2 2 2 2 4 4 4
[297] 4 4 2 2 2 2 2 1 2 1 2 2 3 4 4 4 4 2 3 3 2 2 2 2 2 2 4 4 4 3 2 2 2 2 2 2 2
[334] 2 4 4 4 4 4 4 2 2 2 2 1 4 2 2 2 2 2 2 2 2 2 2 2 2 2 2 3 3 3 2 2 2 2 4 2 2 2
[371] 1 1 4 4 4 4 1 2 1 1 1 4 4 1 3 3 2 2 2 2 2 4 4 1 4 2 2 1 1 1 1 1 3 3 2 4 4
[408] 4 4 4 4 4 4 4 4 4 4 4 3 4 4 4 2 2 3 1 1 2 1 3 3 3 3 2 3 3 1 1 1 2 1 1 1 1
[445] 1 1 1 2 2 2 2 2 2 2 2 1 4 1 1 4 4 4 4 4 4 4 4 1 1 1 2 4 4 4 4 4 4 4 2 2 2
[482] 4 2 2 4 4 4 1 1 1 1 2 2 4 3 3 4 4 4 4

```

```

#which rows each group contains
grp %>% group_rows()

```

```

<list_of<integer>[4]>

```

```

[[1]]
[1] 15 16 17 22 23 35 36 37 38 39 40 41 42 45 46 47 50 51
[19] 52 53 67 72 76 77 78 79 85 87 88 89 93 94 95 99 100 101
[37] 102 103 110 112 113 125 126 145 148 149 150 151 159 160 165 166 167 168
[55] 169 170 171 175 176 177 181 182 188 189 204 207 208 209 210 211 212 226
[73] 227 228 238 239 240 241 242 243 244 245 246 247 248 249 255 256 257 258
[91] 262 263 264 265 281 282 304 306 345 371 372 377 379 380 381 384 394 398
[109] 399 400 401 402 426 427 429 437 438 439 441 442 443 444 445 446 447 456
[127] 458 459 468 469 470 488 489 490 491

```

```

[[2]]
[1] 24 28 29 30 31 32 33 34 48 49 54 55 56 57 58 59 60 61
[19] 62 71 97 111 114 115 116 117 120 121 122 123 124 131 132 133 141 147

```

```
[37] 162 178 179 180 186 190 191 192 193 194 196 197 198 199 200 201 213 214
[55] 215 216 217 220 221 222 223 224 225 253 254 259 260 261 268 269 270 277
[73] 278 279 280 290 291 292 293 299 300 301 302 303 305 307 308 314 317 318
[91] 319 320 321 322 327 328 329 330 331 332 333 334 341 342 343 344 347 348
[109] 349 350 351 352 353 354 355 356 357 358 362 363 364 365 367 368 369 370
[127] 378 387 388 389 390 391 396 397 405 423 424 428 434 440 448 449 450 451
[145] 452 453 454 455 471 479 480 481 483 484 492 493
```

```
[[3]]
```

```
[1] 1 2 4 5 13 44 70 73 74 75 80 81 84 86 107 108 109 119 183
[20] 184 185 205 229 230 231 232 233 234 235 236 267 287 288 289 309 315 316 326
[39] 359 360 361 385 386 403 404 419 425 430 431 432 433 435 436 495 496
```

```
[[4]]
```

```
[1] 3 6 7 8 9 10 11 12 14 18 19 20 21 25 26 27 43 63
[19] 64 65 66 68 69 82 83 90 91 92 96 98 104 105 106 118 127 128
[37] 129 130 134 135 136 137 138 139 140 142 143 144 146 152 153 154 155 156
[55] 157 158 161 163 164 172 173 174 187 195 202 203 206 218 219 237 250 251
[73] 252 266 271 272 273 274 275 276 283 284 285 286 294 295 296 297 298 310
[91] 311 312 313 323 324 325 335 336 337 338 339 340 346 366 373 374 375 376
[109] 382 383 392 393 395 406 407 408 409 410 411 412 413 414 415 416 417 418
[127] 420 421 422 457 460 461 462 463 464 465 466 467 472 473 474 475 476 477
[145] 478 482 485 486 487 494 497 498 499 500
```

```
# names of the grouping variables
grp %>% group_vars()
```

```
[1] "Region"
```

```
grp1 %>% group_vars()
```

```
[1] "Region" "Category"
```

Changing and adding to grouping variables

```
grp %>%
  group_by(Category) %>%
  tally()
```

```
# A tibble: 3 x 2
  Category      n
  <chr>      <int>
1 Furniture    110
2 Office Supplies 293
3 Technology    97
```

```
grp %>%
  group_by(Category, .add = TRUE) %>%
  tally()
```

```
# A tibble: 12 x 3
# Groups:   Region [4]
  Region Category      n
  <chr>   <chr>    <int>
1 Central Furniture     29
2 Central Office Supplies 78
3 Central Technology    28
4 East    Furniture     30
5 East    Office Supplies 100
6 East    Technology    26
7 South   Furniture     16
8 South   Office Supplies 28
9 South   Technology    11
10 West    Furniture     35
11 West    Office Supplies 87
12 West    Technology    32
```

```
grp %>%
  ungroup() %>%
  tally()
```

```
# A tibble: 1 x 1
      n
  <int>
1   500
```

```
grp1 %>%
  ungroup(Category) %>%
```

```
tally()
```

```
# A tibble: 4 x 2
  Region      n
  <chr>   <int>
1 Central  135
2 East    156
3 South   55
4 West   154
```

slice()- select rows within a group

```
grp %>%
  relocate(Region) %>%
  slice(2)
```

```
# A tibble: 4 x 7
# Groups:   Region [4]
  Region Row.ID State      Category      Sub.Category  Sales  Profit
  <chr>   <int> <chr>      <chr>      <chr>      <dbl>   <dbl>
1 Central    16 Texas      Office Supplies Binders      2.54   -3.82
2 East      28 Pennsylvania Furniture      Bookcases 3083. -1665.
3 South      2 Kentucky   Furniture      Chairs     732.   220.
4 West       6 California Furniture      Furnishings 48.9   14.2
```

slice_min() to select the smallest n values of a variable:

slice_max() to select the largest n values of a variable:

```
grp %>%
  relocate(Region,Sales) %>%
  slice_min(Sales,n =1)
```

```
# A tibble: 4 x 7
# Groups:   Region [4]
  Region Sales Row.ID State      Category      Sub.Category Profit
  <chr>   <dbl>   <int> <chr>      <chr>      <chr>      <dbl>
1 Central  1.25      76 Texas      Office Supplies Binders      -1.93
2 East    1.68     62 New York    Office Supplies Art         0.84
3 South   3.30    109 North Carolina Office Supplies Art         1.07
4 West    2.39    127 Arizona    Office Supplies Binders     -1.83
```

```
grp %>%
  relocate(Region,Sales) %>%
  slice_head(n=2)
```

```
# A tibble: 8 x 7
# Groups:   Region [4]
  Region Sales Row.ID State      Category      Sub.Category Profit
  <chr>   <dbl> <int> <chr>      <chr>      <chr>      <dbl>
1 Central 68.8      15 Texas      Office Supplies Appliances -124.
2 Central 2.54      16 Texas      Office Supplies Binders -3.82
3 East    71.4      24 Pennsylvania Furniture Chairs -1.02
4 East    3083.     28 Pennsylvania Furniture Bookcases -1665.
5 South   262.      1 Kentucky Furniture Bookcases 41.9
6 South   732.      2 Kentucky Furniture Chairs 220.
7 West    14.6      3 California Office Supplies Labels 6.87
8 West    48.9      6 California Furniture Furnishings 14.2
```

Useful Summary Functions

n() - returns the size of the current group.

sum(!is.na(x)) - To count the number of non-missing values.

n_distinct(x) - To count the number of distinct (unique) values.

mean(), *median()*, *sd()*, interquartile range- *IQR()*, mean absolute deviation - *mad()*

min(), *max()*

first(x), *nth()*, *last(x)*

min_rank(),

```
df %>%
  count(Category)
```

```
      Category    n
1      Furniture 110
2 Office Supplies 293
3      Technology  97
```

```
df %>%
  group_by(Category) %>%
  summarize(Count = n())
```

```
# A tibble: 3 x 2
  Category      Count
  <chr>         <int>
1 Furniture      110
2 Office Supplies 293
3 Technology      97
```

```
df %>%
  group_by(Category) %>%
  tally()
```

```
# A tibble: 3 x 2
  Category      n
  <chr>         <int>
1 Furniture      110
2 Office Supplies 293
3 Technology      97
```

```
df %>%
  group_by(Category) %>%
  tally(sort = TRUE)
```

```
# A tibble: 3 x 2
  Category      n
  <chr>         <int>
1 Office Supplies 293
2 Furniture      110
3 Technology      97
```

```
df %>%
  group_by(Category) %>%
  summarise(count = sum(Sales > 100))
```

```
# A tibble: 3 x 2
  Category      count
  <chr>         <int>
1 Furniture         72
2 Office Supplies   61
3 Technology        58
```

Ranking

```
x <- c(1,2,2,5)
```

```
row_number(x)
```

```
[1] 1 2 3 4
```

```
min_rank(x)
```

```
[1] 1 2 2 4
```

```
dense_rank(x)
```

```
[1] 1 2 2 3
```

```
df %>%
  select(Region, Sales) %>%
  group_by(Region) %>%
  mutate(rank = min_rank(Sales)) %>%
  arrange(Region,Sales)
```

```
# A tibble: 500 x 3
# Groups:   Region [4]
  Region Sales rank
  <chr>   <dbl> <int>
1 Central 1.25     1
2 Central 1.62     2
3 Central 1.79     3
4 Central 2.08     4
```

```

5 Central 2.2      5
6 Central 2.54     6
7 Central 2.91     7
8 Central 4.04     8
9 Central 4.79     9
10 Central 6.16    10
# i 490 more rows

```

```

df %>%
  select(Region, Sales) %>%
  group_by(Region) %>%
  mutate(rank = dense_rank(Sales)) %>%
  arrange(Region, Sales)

```

```

# A tibble: 500 x 3
# Groups:   Region [4]
  Region Sales rank
  <chr>   <dbl> <int>
1 Central 1.25     1
2 Central 1.62     2
3 Central 1.79     3
4 Central 2.08     4
5 Central 2.2      5
6 Central 2.54     6
7 Central 2.91     7
8 Central 4.04     8
9 Central 4.79     9
10 Central 6.16    10
# i 490 more rows

```

n_distinct(x)

Q. Find group with max rows.

```

df %>%
  group_by(Category) %>%
  summarize(Sub.Categories = n_distinct(Sub.Category))

```

```

# A tibble: 3 x 2
  Category Sub.Categories
  <chr>         <int>

```


1 Furniture	4
2 Office Supplies	9
3 Technology	4

Q. Find max sales Category wise.

```
df %>%
  group_by(Category) %>%
  summarize(count = max(Sales))
```

```
# A tibble: 3 x 2
  Category      count
  <chr>         <dbl>
1 Furniture    3083.
2 Office Supplies 4355.
3 Technology    8160.
```

```
names(df)
```

[1] "Row.ID"	"State"	"Region"	"Category"	"Sub.Category"
[6] "Sales"	"Profit"			

```
head(df[c("Category", "Sales")])
```

	Category	Sales
1	Furniture	261.9600
2	Furniture	731.9400
3	Office Supplies	14.6200
4	Furniture	957.5775
5	Office Supplies	22.3680
6	Furniture	48.8600

```
df %>%
  group_by(Category) %>%
  summarize(first = first(Sales))
```

```
# A tibble: 3 x 2
  Category      first
  <chr>         <dbl>
1 Furniture    262.
2 Office Supplies 14.6
3 Technology   907.
```

Grouping in R Variants

- **Group_by_all**: Allows the user to use every field in the data set.
- **Group_by_if**: Allows you to use an if function to group certain fields.
- **Group_Split**: This allows the user to separate the data into a list of Tibbles.
- **Group_Nest**: This returns a Tibble containing the grouped columns and the data from those respective groups.

```
df %>%
  group_by_all() %>%
  summarise(my_cnt = n(), .groups = 'keep') %>%
  arrange(desc(my_cnt)) %>% group_vars()
```

```
[1] "Row.ID"      "State"      "Region"     "Category"   "Sub.Category"
[6] "Sales"      "Profit"
```

group_split()

```
df_split <- df %>%
  group_by(Category) %>%
  group_split()

df_split[[3]]
```

```
# A tibble: 97 x 7
  Row.ID State      Region Category Sub.Category Sales Profit
  <int> <chr>      <chr>   <chr>   <chr>      <dbl> <dbl>
1      8 California West    Technology Phones      907.   90.7
2     12 California West    Technology Phones      911.   68.4
3     20 California West    Technology Phones     213.   16.0
4     27 California West    Technology Accessories   90.6  11.8
```

5	36	Texas	Central	Technology	Phones	1098.	123.
6	41	Texas	Central	Technology	Phones	371.	41.8
7	42	Illinois	Central	Technology	Phones	147.	16.6
8	45	Minnesota	Central	Technology	Accessories	46.0	19.8
9	48	Delaware	East	Technology	Accessories	45	4.95
10	49	Delaware	East	Technology	Phones	21.8	6.10

i 87 more rows

```
df_nest <- df %>%
  group_nest(Region);
```

```
df_nest
```

```
# A tibble: 4 x 2
  Region      data
  <chr>    <list<tibble[,6]>>
1 Central [135 x 6]
2 East   [156 x 6]
3 South  [55 x 6]
4 West   [154 x 6]
```