Basic Tabular Data Manipulation

Alejandro Schuler, adapted from Steve Bagley and based on R for Data Science by Hadley Wickham, updated to include GTEx sample data by Nicole Ferraro 2019, updated July 2021

- filter rows of a dataset based on conditions
- arrange rows of a dataset based on one or more columns
- select columns of a dataset
- mutate existing columns to create new columns
- group and summarize data by one or more columns
- use the pipe to combine multiple operations

dplyr

This section shows the basic data frame functions ("verbs") in the dplyr package (part of tidyverse).



dplyr verbs

Each operation takes a data frame and produces a new data frame.

- filter() picks out rows according to specified conditions
- select () picks out columns according to their names
- arrange () sorts the row by values in some column(s)
- mutate () creates new columns, often based on operations on other columns
- summarize () collapses many values in one or more columns down to one value per column

These can all be used in conjunction with $group_by$ () which changes the scope of each function from operating on the entire dataset to operating on it group-by-group. These six functions provide the "verbs" for a language of data manipulation.

All work similarly:

- 1. The first argument is a data frame.
- 2. The subsequent arguments describe what to do with the data frame, using the variable names (without quotes).
- 3. The result is a new data frame.

Together these properties make it easy to chain together multiple simple steps to achieve a complex result.

GTEx data

This is a subset of the Genotype Tissue Expression (GTEx) dataset

- The full dataset. Includes gene expression data, measured via RNA-sequencing, from 54 post-mortem tissues in ~800 individuals. Whole genome sequencing is also available for these individuals as part of the GTEx v8 release, available through dbGaP.
- The subsetted dataset. We are looking at expression data for just 78 individuals here, in four tissues including blood, heart, lung and liver.
- Data processing The expression values have been normalized and corrected for technical covariates and are now in the form of Z-scores, which indicate the distance of a given expression value from the mean across all measurements of that gene in that tissue.
- Goal. We will use the data here to illustrate different functions for data transformation, often focused on extracting individuals with extremely high or low expression values for a given gene as compared to the distribution across all samples.

NOTE: If copying the code, make sure there are no spaces in the download link (where it wraps to a new line).

```
# Read subsetted data from online file - make sure there are no spaces
gtex_data = read_tsv('https://raw.githubusercontent.com/alejandroschuler/r4ds-
courses/advance-2020/data/gtex.tissue.zscores.advance2020.txt')

# Check number of rows
nrow(gtex_data)
[1] 389922
```

Filter rows with filter()

Filter rows with filter()

- filter() lets you filter out rows of a dataset that meet a certain condition
- It takes two arguments: the dataset and the condition

```
filter(gtex data, Blood >= 12)
# A tibble: 12 x 7
  Gene
             Ind
                   Blood Heart Lung Liver NTissues
  <chr>
            <chr> <dbl> <dbl> <dbl> <dbl>
                                                <dbl>
1 AC012358.7 GTEX-VUSG 13.6 -1.43 1.22 -0.39
            GTEX-12696 13.6 NA
2 DCSTAMP
                                -0.57 - 0.91
 3 DIAPH2-AS1 GTEX-VUSG 12.2 -0.33 1.18 0.67
 4 DNASE2B
            GTEX-12696 14.4 -0.82 -0.92 0.35
 5 FFAR4
          GTEX-12696 12.9 -0.96 -0.67 0.18
 6 GAPDHP33 GTEX-UPK5 13.8 1.52 -1.48 -1.84
            GTEX-VUSG 12.2 1.67 0.78 0.09
7 GTF2A1L
8 GTF2IP14
           GTEX-11NV4 12.2 7.26 5.79 7.06
 9 KCNT1
            GTEX-1KANB 13.5 3.14 0.62 -0.37
                      15.7 -0.74 -0.44 -0.02
10 KLK3
             GTEX-147F4
11 NAPSA
             GTEX-1CB4J 12.3 -0.29 -0.44 -0.14
12 REN
             GTEX-U8XE 18.9 -0.57 NA
                                        0.09
```

Exercise

• What is the result of running this code?

```
nrow(gtex_data)
[1] 389922

filter(gtex_data, NTissues <= 2)
filter(gtex_data, Heart <= -5)
nrow(gtex_data)</pre>
```

• Remember, functions usually do not change their arguments!

```
low_expression_blood = filter(gtex_data, Blood <= -5)
low_expression_blood_heart = filter(low_expression_blood, Heart <= -5)
nrow(low_expression_blood_heart)
[1] 3</pre>
```

Combining constraints in filter

- This filters by the conjunction of the two constraints—both must be satisfied.
- Constraints appear as second (and third...) arguments, separated by commas.

Filtering out all rows

```
filter(gtex_data, NTissues > 5)
# A tibble: 0 x 7
# ... with 7 variables: Gene <chr>, Ind <chr>, Blood <dbl>, Heart <dbl>,
# Lung <dbl>, Liver <dbl>, NTissues <dbl>
```

• If the constraint is too severe, then you will select no rows, and produce a zero row sized tibble.

Comparison operators

- == and ! = test for equality and inequality (do not use = for equality)
- > and < test for greater-than and less-than
- >= and <= are greater-than-or-equal and less-than-or-equal
- these can also be used directly on vectors outside of data frames

```
c(1,5,-22,4) > 0
[1] TRUE TRUE FALSE TRUE
```

Aside: computers are not perfect, so be careful with checking equality

```
sqrt(2) ^ 2 == 2
[1] FALSE
1 / 49 * 49 == 1
[1] FALSE
```

You can use near () to check that two numbers are the same (up to "machine precision")

```
near(sqrt(2) ^ 2, 2)
[1] TRUE
near(1 / 49 * 49, 1)
[1] TRUE
```

Comparing to NA

• The other "gotcha" is that == cannot be used to compare to NA:

```
\begin{array}{rcl}
x &=& NA \\
x &==& NA \\
[1] & NA
\end{array}
```

- The result actually makes sense though, because I'm asking if "I don't know" is the same as "I don't know". Since either side could be any value, the right answer is "I don't know".
- To check if something is NA, use is . na ()

```
x = NA
is.na(x)
[1] TRUE
```

Logical conjunctions

Logic in R utilizes the following vocabulary: and (&), or (|), in (%in%), not (!), and equals (==)

• When OR (pipe symbol: |) is used to join two conditions, at least one condition must be true for the statement to be true. Otherwise it's false.

• When AND (ampersand symbol: &) is used to join two conditions, both conditions must be true for the statement to be true. Otherwise it's false. Separating conditions by a comma is the same as using & inside filter()

```
filter(gtex_data, Lung > 6 & Liver < -6)
# A tibble: 0 x 7
# ... with 7 variables: Gene <chr>, Ind <chr>, Blood <dbl>, Heart <dbl>,
# Lung <dbl>, Liver <dbl>, NTissues <dbl>
```

• Multiple conjunctions can describe complex logical conditions

Logical conjunctions

```
filter(gtex data, !(Blood < 6 | Lung < 6))</pre>
# A tibble: 5 \times 7
 Gene
             Ind
                  Blood Heart Lung Liver NTissues
           <chr>
                                             <dbl>
           GTEX-17HGU 6.61 0.65
1 CTAG2
                                 7.4
                                      2.85
         GTEX-X3Y1 10.3 7.46 8.12 3.67
2 GTF2IP14
       GTEX-X261 11.1
                            0.02 8.39 5.02
3 KLK3
4 RP11-1228E12.1 GTEX-1KANB 6.18 4.08
                                9.69 6.63
5 TDRD1
             GTEX-ZEX8 10.3
                            3.47
                                 6.19 0.3
```

• The exclamation point! means "NOT", which negates the logical condition

Logical conjunctions

```
filter (gtex data, NTissues %in% c(1,2)) # equivalent to filter (gtex data,
NTissues==1 | NTissues==2)
# A tibble: 132 x 7
                Blood Heart Lung Liver NTissues
  Gene
           Ind
  <dbl>
1 AC016757.3 GTEX-131YS 1.43 NA
                                   -0.07
2 ACTG1P1 GTEX-15RJE NA
                             -0.41 - 0.1
                         NA
3 ACVR2B-AS1 GTEX-1GN73 -0.86 NA
                                    0.46
                              -0.76
4 ADAMTSL1
           GTEX-1LGRB -0.48 NA
5 ADGRF5
           GTEX-ZPU1 -0.13 NA
                              NA
                                 -0.68
6 AOAH GTEX-11GSP -1.27 NA
                                 0.41
                              NA
7 ARV1 GTEX-12WSD -1.07 NA
                                 1.27
                              NA
                                 0.11
8 BORCS7 GTEX-X261 NA
                          0.93 NA
9 C16orf46 GTEX-ZEX8 -0.95 NA
                              NA 0.21
10 C4orf19 GTEX-ZVT3 -0.88 NA
                              NA
                                    0.82
# ... with 122 more rows
```

• %in% returns true for all elements of the thing on the left that are also elements of the thing on the right. This is actually shorthand for a match function (use help ('%in%') to learn more)

Caution! in (without the flanking percent signs) has a different meaning - we'll get to this later in the course.

Exercise: High expression

• How many individuals have high expression (Z > 3) for CTAG2 in both blood and lung?

Exercise: High expression

• How many individuals have high expression (Z > 3) for CTAG2 in both blood and lung?

```
nrow(filter(gtex_data, Gene=="CTAG2", Blood > 3, Lung > 3))
[1] 3
```

Exercise: Low expression

• Create a dataset of all individual-gene pairs with low expression (Z < -3) in blood and heart tissues

Exercise: Low expression

• Create a dataset of all individual-gene pairs with low expression (Z < -3) in blood and heart tissues

```
filter (gtex data, Blood < -3, Heart < -3)
# A tibble: 26 x 7
                  Ind Blood Heart Lung Liver NTissues
  Gene
  <dbl>
1 ABC7-42389800N19.1 GTEX-14E7W -4.24 -3.29 -3.55 -3.2
2 ALDH9A1
          GTEX-11NUK -3.45 -4.78 -4.15 -3.8
               GTEX-1MJIX -4.21 -3.76 -4.56 2.42
3 ARPC3
4 ATL3
              GTEX-ZF29 -5.07 -4.22 -2.77 -2.82
5 ATP5A1 GTEX-YFC4 -5.35 -6.05 -7.96 -4.4
6 CCDC163
             GTEX-1KANB -3.06 -3.05 -0.94 0.48
                  GTEX-ZEX8 -4.49 -3.11 -0.19 -0.9
7 COQ6
               GTEX-WFON -3.77 -6.42 -6.14 -4.11
8 CRYL1
9 CTPS2
             GTEX-1E2YA -3.92 -6.71 -4.06 -1.98
10 ENGASE
        GTEX-147JS -3.96 -3.17 -3.62 -3.22
# ... with 16 more rows
```

Exercise: High expression in all tissues

• Create a dataset of all individual-gene pairs with high expression (Z > 3) in all four tissues

Exercise: High expression in all tissues

• Create a dataset of all individual-gene pairs with high expression (Z > 3) in all four tissues

```
filter(gtex data, Blood > 3, Heart > 3, Lung > 3, Liver > 3)
# A tibble: 16 \times 7
                   Blood Heart Lung Liver NTissues
  Gene
                Ind
  <dbl>
1 AC004453.8 GTEX-12WSI 6.2
                                4.31
                                     4.57
                                          3.77
2 AC004453.8
            GTEX-14DAQ 7.76 6.98
                                     5.64 4.81
 3 CTAG2
               GTEX-YECK
                          4.73 8.37
                                     6.11 3.11
                                     5.9 4.76
 4 CTD-2525I3.5
               GTEX-14E1K 3.27
                               8.78
 5 EMC3-AS1
               GTEX-1KANB
                          5.33
                                     4.21 4.87
 6 GTF2IP14
               GTEX-11NV4 12.2
                               7.26 5.79 7.06
                               7.46 8.12 3.67
7 GTF2IP14
               GTEX-X3Y1 10.3
8 MTATP8P2
               GTEX-17EVP 6.35
                               5.16 5.31 5.44
 9 MTATP8P2
               GTEX-YECK
                          4.21
                               4.04 3.53 4.55
10 PSMA6
               GTEX-12WSI
                         4.23
                               3.92
                                     5.39 3.03
11 RP11-1228E12.1 GTEX-1KANB
                          6.18
                                4.08
                                     9.69
                                          6.63
12 RRP7BP
               GTEX-1AX9I
                          7.49
                               5.45
                                     5.8
                                          3.94
13 TMEM145
               GTEX-1JMLX
                          3.68
                               7.15 7.98 7.96
14 TRBV23-1
               GTEX-1A8FM 6.51
                               8.63
                                     3.15 4.82
15 TRBV3-1
               GTEX-1E2YA 5.04
                               3.64 3.34 4.64
16 TRMT2B
               GTEX-1CB4J
                          6.56 5.74 5.58 5.29
```

Exercise: High and low expression events

• Create a dataset of all high and low expression instances (Z > 3 or < -3) in any tissue (?abs may be helpful)

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• Create a dataset of all high and low expression instances (Z > 3 or < -3) in any tissue (?abs may be helpful)

```
filter(gtex data, abs(Blood) > 3 | abs(Heart) > 3 | abs(Lung) > 3 | abs(Liver) >
3)
# A tibble: 10,171 x 7
  Gene
          Ind Blood Heart Lung Liver NTissues
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl>
                                             <dbl>
1 A2ML1 GTEX-14E7W 0.55 -0.63
                               3.7 1.6
2 A2ML1 GTEX-1GF9V -0.04 1.44 -0.53 3.65
3 A3GALT2 GTEX-11TUW 0.15 -3.41 0.96 -0.83
4 A3GALT2 GTEX-1BAJH -0.7
                          0.91 0.16 3.61
5 A3GALT2 GTEX-XBEC 1.03 3.25 0.87 1.51
6 A3GALT2 GTEX-ZF29 -3.61 -2.02 -2.1 0.23
7 A3GALT2 GTEX-ZTPG 0.94 0.5 -1.23 3.22
8 AAMDC GTEX-12WSG -0.49 -3.38 -2.62 -1.67
 9 AAMDC GTEX-1JMLX 0.65 0.69 0.15 3.43
10 AANAT GTEX-ZTPG
                    1.02 0.02 0.55 3.78
# ... with 10,161 more rows
```

Exercise: getting rid of NAs

• Filter out any rows where the value for Heart is missing (value is NA)

Exercise: getting rid of NAs

• Filter out any rows where the value for Heart is missing (value is NA)

```
filter(gtex data, !is.na(Heart))
# A tibble: 383,941 x 7
  Gene Ind Blood Heart Lung Liver NTissues
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> 
                                             <dbl>
1 A2ML1 GTEX-11DXZ -0.14 -1.08 NA
                                    -0.66
2 A2ML1 GTEX-11GSP -0.5 0.53 0.76 -0.1
3 A2ML1 GTEX-11NUK -0.08 -0.4 -0.26 -0.13
4 A2ML1 GTEX-11NV4 -0.37 0.11 -0.42 -0.61
5 A2ML1 GTEX-11TT1 0.3 -1.11 0.59 -0.12
6 A2ML1 GTEX-11TUW 0.02 -0.47 0.29 -0.66
7 A2ML1 GTEX-11ZUS -1.07 -0.41 0.67 0.06
8 A2ML1 GTEX-11ZVC -0.27 -0.51 0.13 -0.75
 9 A2ML1 GTEX-1212Z -0.3 0.53 0.1 -0.48
10 A2ML1 GTEX-12696 -0.11 0.24 0.96 0.72
# ... with 383,931 more rows
```

Filtering by row number

• Use row_number() to filter specific rows. This is more useful once you have sorted the data in a particular order, which we will soon see how to do.

Sampling rows

• You can use sample_n () to get n randomly selected rows if you don't have a particular condition you would like to filter on.

- sample_frac() is similar
- Do ?sample n() to see how you can sample with replacement or with weights

Sort rows by a column with arrange()

Arrange rows with arrange()

- arrange () takes a data frame and a column, and sorts the rows by the values in that column (ascending order).
- Again, the first argument is the data frame and the other arguments tell the function what to do with it

```
arrange(gtex data, Blood)
# A tibble: \overline{3}89,922 \times 7
  Gene
               Ind
                          Blood Heart Lung Liver NTissues
   <chr>
              <chr>
                          <dbl> <dbl> <dbl> <dbl> <
                                                       <dbl>
                                -1.52 \quad -1.44 \quad -2.15
1 HBA2
              GTEX-11DXZ -9.44
2 MTATP6P1 GTEX-1KD5A -9.18 -10.1 -10.3 -9.52
 3 RP11-46D6.1 GTEX-14E1K -7.83 -3.94 -5.22 -4.49
 4 CYTH3
              GTEX-11NV4 -6.63 -0.6
                                      -0.37 - 1.32
 5 TRG-AS1
              GTEX-11NV4 - 6.47
                                2.39 -0.6 -0.22
 6 SMG1P1
              GTEX-11ZUS -6.26 -1.68 -1.41 -0.31
                                0.77 \quad 0.51 \quad -0.67
7 ZBTB10
              GTEX-VUSG -6.13
8 RPS29
              GTEX-1B8L1 -5.84 -0.8 -0.46 -0.17
 9 GHITM
              GTEX-WK11 -5.7 -7.24 -7.37 -4.06
              GTEX-VUSG -5.62
10 ZNF2
                                1.52
                                       0.61 0.13
# ... with 389,912 more rows
```

Arrange can sort by more than one column

• This is useful if there is a tie in sorting by the first column.

```
arrange(gtex data, NTissues, Blood)
# A tibble: \overline{3}89,922 \times 7
                     Blood Heart Lung Liver NTissues
  Gene
            Ind
                 <dbl> <dbl> <dbl> <dbl> <
  <chr> <chr>
                                              <dbl>
1 HEATR1 GTEX-1EWIQ -1.63 NA
                                   NA 0.49
2 FOXO1 GTEX-1BAJH -1.58 NA
                                   NA - 0.25
      GTEX-12WSI -1.57 NA
3 UCN
                                  NA - 0.48
4 GPR171 GTEX-132NY -1.53 NA
                                   NA - 1.03
5 UCN
        GTEX-WFON -1.46 NA
                                  NA - 0.15
6 KIAA1614 GTEX-12WSI -1.35 NA
                               NA - 0.46
7 ENTPD1-AS1 GTEX-11NUK -1.28 NA NA -0.54
8 TOP3B GTEX-1A32A -1.28 NA NA -0.76
9 AOAH GTEX-11GSP -1.27 NA
                                   NA 0.41
10 PRRX2
        GTEX-1A8FM -1.13 -1.2
                                   NA NA
# ... with 389,912 more rows
```

Use the desc function to arrange by descending values

```
arrange(gtex data, desc(Blood))
# A tibble: \overline{389}, 922 x 7
           Ind
                Blood Heart Lung Liver NTissues
  Gene
  <dbl>
1 REN GTEX-U8XE 18.9 -0.57 NA
                                     0.09
      GTEX-147F4 15.7 -0.74 -0.44 -0.02
2 KLK3
3 DNASE2B GTEX-12696 14.4 -0.82 -0.92 0.35
4 GAPDHP33 GTEX-UPK5 13.8 1.52 -1.48 -1.84
5 DCSTAMP GTEX-12696 13.6 NA
                             -0.57 - 0.91
6 AC012358.7 GTEX-VUSG 13.6 -1.43 1.22 -0.39
7 KCNT1
           GTEX-1KANB 13.5 3.14 0.62 -0.37
8 FFAR4 GTEX-12696 12.9 -0.96 -0.67 0.18
9 NAPSA GTEX-1CB4J 12.3 -0.29 -0.44 -0.14
10 DIAPH2-AS1 GTEX-VUSG 12.2 -0.33 1.18 0.67
# ... with 389,912 more rows
```

Exercise: top 5 high expression instances

Use arrange() and filter() to get the data for the 5 individual-gene pairs with the most extreme expression changes in blood

Exercise: top 5 high expression instances

Use arrange () and filter () to get the data for the 5 individual-gene pairs with the most extreme expression changes in blood

Exercise: top 5 high expression instances

Use arrange () and filter () to get the data for the 5 individual-gene pairs with the most extreme expression changes in blood

or

Select columns with select()

Select columns with select()

```
select(gtex data, Gene, Ind, Blood)
# A tibble: 389,922 x 3
  Gene Ind
                    Blood
   <chr> <chr>
                    <dbl>
1 A2ML1 GTEX-11DXZ -0.14
2 A2ML1 GTEX-11GSP -0.5
 3 A2ML1 GTEX-11NUK -0.08
 4 A2ML1 GTEX-11NV4 -0.37
 5 A2ML1 GTEX-11TT1
 6 A2ML1 GTEX-11TUW
 7 A2ML1 GTEX-11ZUS -1.07
 8 A2ML1 GTEX-11ZVC -0.27
 9 A2ML1 GTEX-1212Z -0.3
10 A2ML1 GTEX-12696 -0.11
# ... with 389,912 more rows
```

• The select function will return a subset of the tibble, using only the requested columns in the order specified.

Select columns with select()

• select() can also be used with handy helpers like starts_with() and contains()

• Use ?select to see all the possibilities

Select columns with select()

```
select(gtex data, contains("N"))
# A tibble: 389,922 x 4
   Gene Ind
                      Lung NTissues
   <chr> <chr>
                    <dbl>
                              <dbl>
1 A2ML1 GTEX-11DXZ NA
 2 A2ML1 GTEX-11GSP
 3 A2ML1 GTEX-11NUK -0.26
 4 A2ML1 GTEX-11NV4 -0.42
 5 A2MT<sub>1</sub>1 GTEX-11TT1
 6 A2ML1 GTEX-11TUW
                     0.29
  A2ML1 GTEX-11ZUS
                     0.67
 8 A2ML1 GTEX-11ZVC
                     0.13
 9 A2ML1 GTEX-1212Z
10 A2ML1 GTEX-12696 0.96
# ... with 389,912 more rows
```

- The quotes around the letter "N" make it a string. If we did not do this, \mathbb{R} would think it was looking for a variable called \mathbb{N} and not just the plain letter.
- We don't have to quote the names of columns (like Ind) because the tidyverse functions know that we are working within the dataframe and thus treat the column names like they are variables in their own right

select() subsets columns by name

• select () can also be used to select everything except for certain columns

select() subsets columns by name

• or even to select only columns that match a certain condition

```
select(gtex data, where(is.numeric))
# A tibble: 389,922 x 5
  Blood Heart Lung Liver NTissues
 <dbl> <dbl> <dbl> <dbl> <dbl>
                               <dbl>
1 -0.14 -1.08 NA
                     -0.66
2 -0.5 0.53 0.76 -0.1
 3 - 0.08 - 0.4 - 0.26 - 0.13
 4 -0.37 0.11 -0.42 -0.61
 5 0.3 -1.11 0.59 -0.12
 6 \quad 0.02 \quad -0.47 \quad 0.29 \quad -0.66
7 -1.07 -0.41 0.67 0.06
8 -0.27 -0.51 0.13 -0.75
 9 -0.3 0.53 0.1 -0.48
10 -0.11 0.24 0.96 0.72
# ... with 389,912 more rows
```

pull() is a friend of select()

• select () has a friend called pull () which returns a vector instead of a (one-column) data frame

```
select(gtex data, Gene)
# A tibble: 389,922 x 1
   Gene
   <chr>
 1 A2ML1
 2 A2ML1
 3 A2ML1
 4 A2ML1
 5 A2ML1
 6 A2ML1
 7 A2ML1
 8 A2ML1
 9 A2ML1
10 A2ML1
# ... with 389,912 more rows
pull(gtex data, Gene)
        "A2ML1"
                                                     "A2ML1"
                               "A2ML1"
        "A2ML1"
                               "A2ML1"
                                                     "A2ML1"
                               "A2ML1"
                                                     "A2ML1"
        "A2ML1"
                               "A2ML1"
                                                     "A2ML1"
        "A2ML1"
   [13] "A2ML1"
                               "A2ML1"
                                                     "A2ML1"
• • •
```

Rename column names with rename()

• select() can be used to rename variables, but it drops all variables not selected

• rename () is better suited for this because it keeps all the columns

Note: mutate(), can also change a column name (more on mutate() soon)

Exercise: select and filter

• Create a one-column dataframe of the heart expression Z-scores (Heart) of all individuals with data present (i.e. not NA) for gene WDR34 (Gene) in the gtex_data dataset.

Exercise: select and filter

• Create a one-column dataframe of the heart expression Z-scores (Heart) of all individuals with data present (i.e. not NA) for gene WDR34 (Gene) in the gtex data dataset.

Exercise: select and filter

• Create a one-column dataframe of the heart expression Z-scores (Heart) of all individuals with data present (i.e. not NA) for gene WDR34 (Gene) in the gtex data dataset.

```
select(filter(gtex_data, Gene == "WDR34", !is.na(Heart)), Heart)
# A tibble: 78 x 1
    Heart
        <dbl>
1     0.78
2     0.94
3     0.29
4     0.17
5     -0.6
6     0.46
7     -0.97
```

• What is wrong with this?

```
filter(select(gtex_data, Heart), Gene == "WDR34")
```

Exercise: select text columns

- Use select to subset the gtex data dataframe to just those columns that contain text data.
- Can you do this programmatically without specifying the names of each of the desired columns?
- Which base R function will help you determine if a column is textual or not? Use whatever tools you want to find out.

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Add new variables with mutate()

Add new variables with mutate()

- This uses mutate () to add a new column to which is the absolute value of Blood.
- The thing on the left of the = is a new name that you make up which you would like the new column to be called
- The expresssion on the right of the = defines what will go into the new column
- Warning! If the new variable name already exists, mutate () will overwrite the existing one

mutate() can create multiple new columns at once

• mutate () can create multiple columns at the same time and use multiple columns to define a single new one

```
mutate (gtex data, # the newlines make it more readable
     abs blood = abs (Blood),
     abs heart = abs(Heart),
     blood heart dif = abs blood - abs heart
# A tibble: 389,922 x 10
  Gene Ind Blood Heart Lung Liver NTissues abs blood abs heart
   <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> 
                                            <dbl>
                                                      <dbl>
                                                               <dbl>
                                        0.1
4 0.5
4 0.08
4
1 A2ML1 GTEX-11DXZ -0.14 -1.08 NA
                                    -0.66
                                                               1.08
                                                      0.5
2 A2ML1 GTEX-11GSP -0.5 0.53 0.76 -0.1
                                                               0.53
3 A2ML1 GTEX-11NUK -0.08 -0.4 -0.26 -0.13
                                                               0.4
4 A2ML1 GTEX-11NV4 -0.37 0.11 -0.42 -0.61
                                                               0.11
 5 A2ML1 GTEX-11TT1 0.3 -1.11 0.59 -0.12
                                                       0.3
                                                               1.11
                                                       0.02
                                                               0.47
 6 A2ML1 GTEX-11TUW 0.02 -0.47 0.29 -0.66
7 A2ML1 GTEX-11ZUS -1.07 -0.41
                               0.67 0.06
                                                      1.07
                                                               0.41
8 A2ML1 GTEX-11ZVC -0.27 -0.51 0.13 -0.75
                                                      0.27
                                                               0.51
 9 A2ML1 GTEX-1212Z -0.3 0.53 0.1 -0.48
                                                       0.3
                                                               0.53
10 A2ML1 GTEX-12696 -0.11 0.24 0.96 0.72
                                                       0.11
                                                                0.24
# ... with 389,912 more rows, and 1 more variable: blood heart dif <dbl>
```

• Note that we have also used two columns simultaneously (Blood and Heart) to create a new column)

mutate() for data type conversion

• Data is sometimes given to you in a form that makes it difficult to do operations on

```
df = tibble(number = c("1", "2", "3"))
df
# A tibble: 3 x 1
    number
    <chr>
1 1
2 2
3 3
mutate(df, number_plus_1 = number + 1)
Error: Problem with `mutate()` column `number_plus_1`.
i `number_plus_1 = number + 1`.
x non-numeric argument to binary operator
```

• mutate () is also useful for converting data types, in this case text to numbers

Exercise: mutate()

I want to identify genes that have large average expression changes across blood and liver. Can you compute the average of blood and liver expression changes across all gene-individual pairs? Compute the average manually (i.e. don't use the mean function).

Exercise: mutate()

I want to identify genes that have large average expression changes across blood and liver. Can you compute the average of blood and liver expression changes across all gene-individual pairs? Compute the average manually (i.e. don't use the mean function).

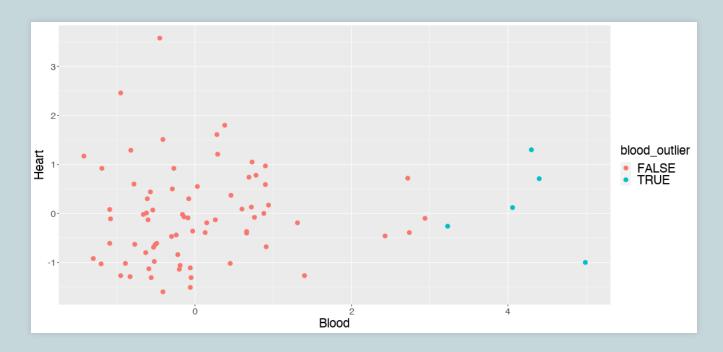
```
mutate(gtex data, avg blood liver = (Blood+Liver) /2)
# A tibble: 389,922 x 8
  Gene Ind Blood Heart Lung Liver NTissues avg blood liver
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl>
                                              <dbl>
                                                              <dbl>
1 A2ML1 GTEX-11DXZ -0.14 -1.08 NA
                                     -0.66
                                                             -0.4
2 A2ML1 GTEX-11GSP -0.5
                          0.53 \quad 0.76 \quad -0.1
                                                             -0.3
3 A2ML1 GTEX-11NUK -0.08 -0.4 -0.26 -0.13
                                                             -0.105
                                                             -0.49
4 A2ML1 GTEX-11NV4 -0.37 0.11 -0.42 -0.61
                                                             0.09
 5 A2ML1 GTEX-11TT1 0.3 -1.11 0.59 -0.12
 6 A2ML1 GTEX-11TUW
                   0.02 - 0.47
                                0.29 - 0.66
                                                             -0.32
7 A2ML1 GTEX-11ZUS -1.07 -0.41
                                                             -0.505
                                0.67 0.06
8 A2ML1 GTEX-11ZVC -0.27 -0.51
                                                             -0.51
                                0.13 - 0.75
 9 A2ML1 GTEX-1212Z -0.3 0.53 0.1 -0.48
                                                            -0.39
10 A2ML1 GTEX-12696 -0.11 0.24 0.96 0.72
                                                              0.305
# ... with 389,912 more rows
```

Exercise: mutate() and ggplot

Filter $gtex_data$ to only include measurements of the MYL1 gene. Then, use mutate to mark which gene-individual pairs have outlier MYL1 expression in blood, defined as Z > 3 or Z < -3. Then, produce a plot showing blood Z-scores vs heart Z-scores and color the blood gene expression outliers in a different color than the other points.

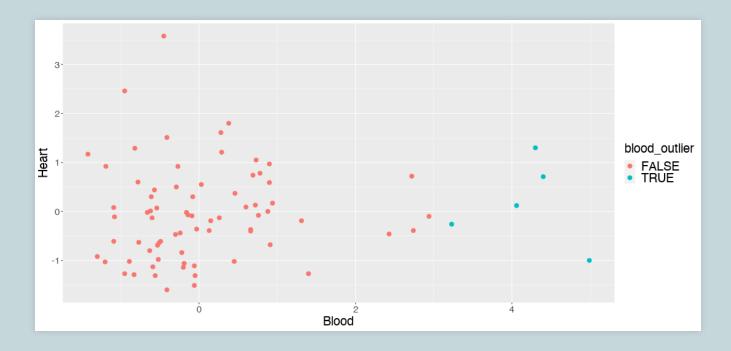
Exercise: mutate() and ggplot

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Exercise: mutate() and ggplot

Filter $gtex_data$ to only include measurements of the MYL1 gene. Then, use mutate to mark which gene-individual pairs have outlier MYL1 expression in blood, defined as Z > 3 or Z < -3. Then, produce a plot showing blood Z-scores vs heart Z-scores and color the blood gene expression outliers in a different color than the other points.



```
gene_data = filter(gtex_data, Gene == 'MYL1')
blood_outliers = mutate(gene_data, blood_outlier = abs(Blood)>3)
ggplot(blood_outliers) +
  geom_point(aes(x=Blood, y=Heart, color=blood_outlier))
```

Exercise: putting it together

I am interested in identifying individuals that have a large change in gene expression change for any gene between lung tissue and blood tissue, with higher expression in lung.

- 1. Produce a list of top 10 individual-gene pairs arranged by the expression change for lung compared to blood
- 2. Only consider individual-gene pairs measured in all four tissues
- 3. In the output, just show the gene, individual, lung expression, blood expression, and lung-blood differences

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- 2. Only consider individual-gene pairs measured in all four tissues
- 3. In the output, just show the gene, individual, lung expression, blood expression, and lung-blood differences

```
gtex data no change = filter(gtex data, NTissues == 4)
gtex data ratio = mutate(gtex data no change, lung blood dif = Lung - Blood)
sorted = arrange(gtex_data_ratio, desc(lung_blood_dif))
top 10 = filter(sorted, row number() <=10)
select (top 10, Gene, Ind, Lung, Blood, lung blood dif)
# A tibble: 10 x 5
  Gene Ind Lung Blood lung_blood_dif
 <chr> <chr> <chr> <dbl> <dbl> <dbl>
1 TMEM151A GTEX-ZEX8 12.0 -0.77 12.8
2 DSG3 GTEX-14PJO 10.3 -0.04 10.4
3 KNG1
             GTEX-12WSD 11.6 1.74
                                      9.82
4 NPIPA8
             GTEX-XXEK 9.23 0.03
                                       9.2
                                 8.82
5 ALOXE3
             GTEX-YFC4 7.5 -1.32
6 ETNPPL GTEX-1CB4J 8.47 -0.31
                                 8.78
      GTEX-13FTZ 7.79 -0.75 8.54
7 EGR4
8 IGHV2-26 GTEX-YECK 7.82 -0.63 8.45
9 LINC00342 GTEX-1A32A 6.01 -2.16
                                       8.17
10 RP11-1260E13.4 GTEX-11ZVC 6.96 -1.17
                                        8.13
```

Piping

Why pipe?

• In our last exercise, we used a number of different function applications to arrive at our answer. Shown below, we used temporary variables to keep our code clean.

```
gtex_data_no_change = filter(gtex_data, NTissues == 4)
gtex_data_ratio = mutate(gtex_data_no_change, lung_blood_dif = Lung - Blood)
sorted = arrange(gtex_data_ratio, desc(lung_blood_dif))
top_10 = filter(sorted, row_number() <= 10)
select(top_10, Gene, Ind, Lung, Blood, lung_blood_dif)</pre>
```

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• In our last exercise, we used a number of different function applications to arrive at our answer. Shown below, we used temporary variables to keep our code clean.

```
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top_10 = filter(sorted, row_number() <= 10)
select(top_10, Gene, Ind, Lung, Blood, lung_blood_dif)</pre>
```

• Compare that to the same code using nested calls (instead of storing in temporary variables):

```
select(
  filter(
    arrange(
       mutate(
         filter(
            gtex_data, NTissues == 4),
            lung_blood_dif = Lung - Blood),
            desc(lung_blood_dif)),
        row_number() <= 10),
        Gene, Ind, Lung, Blood, lung_blood_dif
)</pre>
```

What makes either of these hard to read or understand?

The pipe operator

• Tidyverse solves these problems with the pipe operator %>%

```
gtex_data %>%
  filter(NTissues == 4) %>%
  mutate(lung_blood_dif = Lung - Blood) %>%
  arrange(desc(lung_blood_dif)) %>%
  filter(row_number() <= 10) %>%
  select(Gene, Ind, Lung, Blood, lung_blood_dif)
```

The pipe operator

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```
gtex_data %>%
  filter(NTissues == 4) %>%
  mutate(lung_blood_dif = Lung - Blood) %>%
  arrange(desc(lung_blood_dif)) %>%
  filter(row_number() <= 10) %>%
  select(Gene, Ind, Lung, Blood, lung_blood_dif)
```

• How does this compare with our code before? What do you notice?

```
gtex_data_no_change = filter(gtex_data, NTissues == 4)
gtex_data_ratio = mutate(gtex_data_no_change, lung_blood_dif = Lung - Blood)
sorted = arrange(gtex_data_ratio, desc(lung_blood_dif))
top_10 = filter(sorted, row_number() <= 10)
select(top_10, Gene, Ind, Lung, Blood, lung_blood_dif)</pre>
```

Pipe details: What happens to an object when it gets "piped in"?

When df1 is piped into fun (x) (fun is just some fake function)

```
df1 %>% fun(x)
```

is converted into:

```
fun(df1, x)
```

- That is: the thing being piped in is used as the first argument of fun.
- The tidyverse functions are consistently designed so that the first argument is a data frame, and the result is a data frame, so you can push a dataframe all the way through a series of functions

Pipe details: What objects can be piped?

• The pipe works for all variables and functions (not just tidyverse functions)

Piping with an array

```
c(1,44,21,0,-4) %>%
sum() # instead of sum(c(1,44,21,0,-4))
[1] 62
```

Piping with a scalar

```
1 %>% `+`(1) # `+` is just a function that takes two arguments!
[1] 2
```

Piping with a data frame

Piping to another position

• The pipe typically pipes into the first argument of a function, but you can use . to represent the object you're piping into the function

```
# install.packages("slider")
library(slider)
mean %>%
   slide_vec(1:10, ., .before=2)
[1] 1.0 1.5 2.0 3.0 4.0 5.0 6.0 7.0 8.0 9.0
```

- Also notice how I've piped in a function to a function! (yes, functions are just objects like anything else in R)
- More about this in the functional programming section

Exercise: Pipe to ggplot

• Run this code to see what it does. Then rewrite it using the pipe operator and get it to produce the same output.

```
gene_data = filter(gtex_data, Gene == 'MYBL2')
outliers = mutate(gene_data, blood_outlier = abs(Blood) > 2)
ggplot(outliers) +
  geom_bar(aes(x=blood_outlier)) +
  scale_x_discrete("Class", labels=c("Other", "Outlier")) +
  ggtitle("How many individuals have outlier MYBL2 expression in blood?")
```

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  scale_x_discrete("Class", labels=c("Other", "Outlier")) +
  ggtitle("How many individuals have outlier MYBL2 expression in blood?")
```

```
gtex_data %>%
  filter(Gene == 'MYBL2') %>%
  mutate(blood_outlier = abs(Blood) > 2) %>%
ggplot() +
  geom_bar(aes(x=blood_outlier)) +
  scale_x_discrete("Class", labels=c("Other", "Outlier")) +
  ggtitle("How many individuals have outlier MYBL2 expression in blood?")
```

- summarize() boils down the data frame according to the conditions it gets. In this case, it creates a data frame with a single column called tissue_avg that contains the mean of the NTissues column
- As with mutate (), the name on the left of the = is something you make up that you would like the new column to be named.
- mutate() transforms columns into new columns of the same length, but summarize() collapses down the data frame into a single row
- Summaries are more useful when you apply them to subgoups of the data, which we will soon see how to do.

• note that you can also pass in multiple conditions that operate on multiple columns at the same time

• Summaries are more useful when you apply them to subgoups of the data

```
gtex data %>%
 group_by(Gene) %>%
 summarize(max blood = max(Blood))
# A tibble: 4,999 \times 2
  Gene
                   max blood
  <chr>
                        <dbl>
1 A2ML1
                         2.08
2 A3GALT2
                         2.77
3 A4GALT
                       2.78
4 AAMDC
                        NA
5 AANAT
                        1.71
                        2.52
6 AAR2
                        1.89
7 AARSD1
8 AB019441.29
                       2.31
 9 ABC7-42389800N19.1
                     1.98
10 ABCA5
                         2.3
# ... with 4,989 more rows
```

Multiple columns can be used to group the data simultaneously

```
gtex data %>%
 group by (Gene, Ind) %>%
 summarize(max blood = max(Blood))
# A tibble: 389,922 x 3
# Groups: Gene [4,999]
  Gene Ind
                 max blood
  <chr> <chr> <dbl>
1 A2ML1 GTEX-11DXZ -0.14
                  -0.5
2 A2ML1 GTEX-11GSP
                  -0.08
3 A2ML1 GTEX-11NUK
                  -0.37
4 A2ML1 GTEX-11NV4
                  0.3
 5 A2ML1 GTEX-11TT1
                  0.02
 6 A2ML1 GTEX-11TUW
7 A2ML1 GTEX-11ZUS
                  -1.07
                  -0.27
8 A2ML1 GTEX-11ZVC
                  -0.3
 9 A2ML1 GTEX-1212Z
10 A2ML1 GTEX-12696
                  -0.11
# ... with 389,912 more rows
```

• The result has the summary value for each unique combination of the grouping variables

Computing the number of rows in each group

• The n () function counts the number of rows in each group:

```
gtex data %>%
  filter(!is.na(Blood)) %>%
  group by (Gene) %>%
  summarize(how many = n())
# A tibble: 4,999 \times 2
   Gene
                       how many
   <chr>
                          <int>
1 A2MT<sub>1</sub>1
2 A3GALT2
 3 A4GALT
 4 AAMDC
 5 AANAT
 6 AAR2
 7 AARSD1
 8 AB019441.29
 9 ABC7-42389800N19.1
10 ABCA5
# ... with 4,989 more rows
```

• You can also use count (), which is just a shorthand for the same thing

```
gtex_data %>%
  filter(!is.na(Blood)) %>%
  group_by(Gene) %>%
  count()
```

Computing the number of distinct elements in a column, per group

• n distinct() counts the number of unique elements in a column

```
gtex data %>%
 group by (Ind) %>%
 summarize(n_genes = n_distinct(Gene))
# A tibble: 78 \times 2
  Ind
       n_genes
  <chr> <int>
1 GTEX-11DXZ 4999
2 GTEX-11GSP
             4999
             4999
3 GTEX-11NUK
             4999
4 GTEX-11NV4
             4999
 5 GTEX-11TT1
6 GTEX-11TUW
             4999
              4999
7 GTEX-11ZUS
             4999
8 GTEX-11ZVC
 9 GTEX-1212Z
             4999
10 GTEX-12696
             4999
# ... with 68 more rows
```

Exercise: top expression per tissue

- Ignoring NAs, what is the highest liver expression value seen for each gene in the gtex_data dataset?
- What about the lowest?

Exercise: top expression per tissue

- Ignoring NAs, what is the highest liver expression value seen for each gene in the gtex data dataset?
- What about the lowest?

```
gtex data %>%
 group by (Gene) %>%
 summarize(
   max liver = max(Liver, na.rm=T),
   min liver = min(Liver, na.rm=T)
# A tibble: 4,999 x 3
                 max liver min liver
  Gene
                   _<dbl> <dbl>
  <chr>
                     3.65 -1.94
1 A2ML1
                     3.61 -1.3
2 A3GALT2
                    2.22 -1.76
3 A4GALT
                   3.43 -2.62
4 AAMDC
                  3.78 -2.22
5 AANAT
                2.32 -3.23
 6 AAR2
                   2.77 - 2.75
7 AARSD1
8 AB019441.29 3.36 -1.53
9 ABC7-42389800N19.1 2.51 -3.2
                     3.27
                          -3.27
10 ABCA5
# ... with 4,989 more rows
```

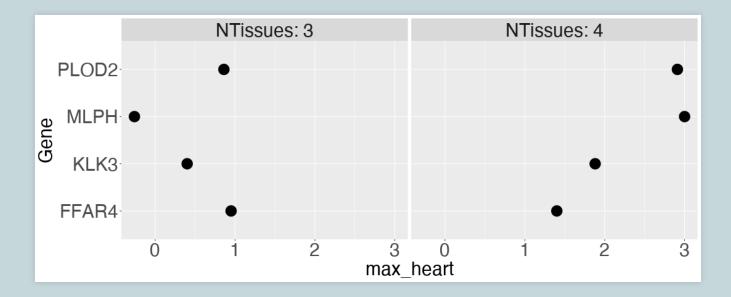
Exercise: summarize and plot

Recreate this plot.



Exercise: summarize and plot

Recreate this plot.



```
gtex_data %>%
  filter(Gene %in% c('FFAR4', 'KLK3', 'PLOD2', 'MLPH')) %>%
  group_by(Gene, NTissues) %>%
  summarize(max_heart = max(Heart ,na.rm=T)) %>%

ggplot() +
  geom_point(aes(y=Gene, x=max_heart)) +
  facet_grid(. ~ NTissues , labeller = label_both)
```

Grouped mutates and filters

Filtering grouped data

• filter() is aware of grouping. When used on a grouped dataset, it applies the filtering condition separately in each group

```
gtex data %>%
 group by (Gene) %>%
 filter(NTissues == max(NTissues))
# A tibble: 376,883 x 7
# Groups: Gene [4,999]
  Gene Ind
                   Blood Heart Lung Liver NTissues
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> 
                                              <dbl>
1 A2ML1 GTEX-11GSP -0.5
                          0.53
                                0.76 - 0.1
2 A2ML1 GTEX-11NUK -0.08 -0.4 -0.26 -0.13
 3 A2ML1 GTEX-11NV4 -0.37 0.11 -0.42 -0.61
 4 A2ML1 GTEX-11TT1
                    0.3 - 1.11
 5 A2ML1 GTEX-11TUW
                   0.02 - 0.47
 6 A2ML1 GTEX-11ZUS -1.07 -0.41
 7 A2ML1 GTEX-11ZVC -0.27 -0.51
                                0.13 - 0.75
8 A2ML1 GTEX-1212Z -0.3
                          0.53
 9 A2ML1 GTEX-12696 -0.11
                          0.24 0.96 0.72
10 A2ML1 GTEX-12WSD 0.53 0.36 0.2 0.51
# ... with 376,873 more rows
```

- Why do we get back multiple rows per class?
- This is an extremely convenient idiom for finding the rows that minimize or maximize a condition

Exercise: Max expression change in blood and lung

Which are the individual pairs that have both the max blood expression change and max lung expression change among all individuals with measurements for the same gene?

```
gtex data %>%
 group by (Gene) %>%
 filter (Blood == max (Blood), Lung==max (Lung))
# A tibble: 64 x 7
# Groups: Gene [64]
  Gene
            Ind Blood Heart Lung Liver NTissues
  <dbl>
1 A4GALT GTEX-12696 2.78 -1.02 2.31 -0.23
2 ABHD1 GTEX-VUSG 6.33 0.41 2.04 -0.04
3 AL162151.3 GTEX-WZTO
                       2.37 -0.19 4.23 -1.22
4 ANKRD36B
            GTEX-12WSD
                      2.72 0.74 2.66 1.22
        GTEX-12WSD
 5 APOA1
                       5.45 NA
                                       0.67
6 C14orf119 GTEX-11ZUS
                      2.51 0.76 1.85 -0.99
7 CD1D
            GTEX-1B996
                      3.05 2.85 2.78 2.1
8 CTB-131B5.2 GTEX-1GN73 6.29 -1.17
                                  5.51 - 0.96
9 CTC-448F2.6 GTEX-131YS 4.1
                            0.75 \quad 2.67 \quad -0.24
10 EVC
            GTEX-UPK5
                       4.31 0.21 2.6 -0.61
# ... with 54 more rows
```

Mutating grouped data

• mutate () is aware of grouping. When used on a grouped dataset, it applies the mutation separately in each group

```
gtex data %>%
 group by (Gene) %>%
 mutate(blood diff from min = Blood - min(Blood)) %>%
  select(Gene, Ind, Blood, blood_diff_from_min)
# A tibble: 389,922 x 4
# Groups: Gene [4,999]
  Gene Ind Blood blood diff from min
  <chr> <chr> <dbl>
                                      <dbl>
1 A2ML1 GTEX-11DXZ -0.14
                                     1.26
2 A2ML1 GTEX-11GSP -0.5
                                     0.9
 3 A2ML1 GTEX-11NUK -0.08
                                      1.32
                                      1.03
 4 A2ML1 GTEX-11NV4 -0.37
 5 A2ML1 GTEX-11TT1 0.3
                                      1.7
 6 A2ML1 GTEX-11TUW
                   0.02
                                    1.42
7 A2ML1 GTEX-11ZUS -1.07
                                  0.33
8 A2ML1 GTEX-11ZVC -0.27
                                   1.13
 9 A2ML1 GTEX-1212Z -0.3
                                      1.1
10 A2ML1 GTEX-12696 -0.11
                                       1.29
# ... with 389,912 more rows
```

• As always, mutate does not change the number of rows in the dataset

Data Transformation with dplyr:: cheat sheet dplyr functions work with pipes and expect tidy data. In tidy data: Manipulate Cases Manipulate Variables **EXTRACT CASES** EXTRACT VARIABLES Row functions return a subset of rows as a new table. Column functions return a set of columns as a new vector or table. Each variable is in Each observation, or x %>% f(y) its own column case, is in its own row becomes f(x, y) pull(.data, var = -1) Extract column values as filter(.data, ...) Extract rows that meet logical a vector. Choose by name or index. criteria. filter(iris, Sepal.Length > 7) pull(iris, Sepal.Length) **Summarise Cases** distinct(.data, ..., .keep_all = FALSE) Remove select(.data, ...) Extract columns as a table. Also select_if(). rows with duplicate values. These apply summary functions to columns to create a new distinct(iris, Species) select(iris, Sepal.Length, Species) table of summary statistics. Summary functions take vectors as input and return one value (see back) sample_frac(tbl, size = 1, replace = FALSE, weight = NULL, .env = parent.frame()) Randomly Use these helpers with select (), summary function select fraction of rows. e.g. select(iris, starts_with("Sepal")) sample_frac(iris, 0.5, replace = TRUE) summarise(.data, ...) contains(match) num_range(prefix, range) :, e.g. mpg:cyl sample_n(tbl, size, replace = FALSE, weight = Compute table of summaries. ends with(match) one of(...) -, e.g, -Species NULL, .env = parent.frame()) Randomly select summarise(mtcars, avg = mean(mpg)) matches(match) starts_with(match) size rows. $sample_n(iris, 10, replace = TRUE)$ count(x, ..., wt = NULL, sort = FALSE) slice(.data, ...) Select rows by position. MAKE NEW VARIABLES Count number of rows in each group defined slice(iris, 10:15) by the variables in ... Also tally(). These apply vectorized functions to columns. Vectorized funs take count(iris, Species) top_n(x, n, wt) Select and order top n entries (by vectors as input and return vectors of the same length as output group if grouped data). top_n(iris, 5, Sepal.Width) VARIATIONS vectorized function summarise_all() - Apply funs to every column. summarise_at() - Apply funs to specific columns. **mutate(**.data, ...**)** Logical and boolean operators to use with filter() Compute new column(s) summarise_if() - Apply funs to all cols of one type. mutate(mtcars, gpm = 1/mpg) is.na() xor() !is.na() transmute(.data, ...) **Group Cases** See ?base::logic and ?Comparison for help. Compute new column(s), drop others. transmute(mtcars, qpm = 1/mpq)Use group_by() to create a "grouped" copy of a table. dplyr functions will manipulate each "group" separately and mutate_all (.tbl, .funs, ...) Apply funs to every then combine the results. ARRANGE CASES column. Use with funs(). Also mutate_if(). mutate_all(faithful, funs(log(.), log2(.))) arrange(.data, ...) Order rows by values of a mutate if(iris, is.numeric, funs(log(.))) column or columns (low to high), use with desc() to order from high to low. group_by(cyl) %>% mutate_at(.tbl, .cols, .funs, ...) Apply funs to arrange(mtcars, mpg) specific columns. Use with funs(), vars() and summarise(avg = mean(mpg) arrange(mtcars, desc(mpg)) the helper functions for select(). mutate_at(iris, vars(-Species), funs(log(.))) ADD CASES ungroup(x,...) add_column(.data, ..., .before = NULL, .after = Returns ungrouped copy NULL) Add new column(s). Also add_count(), add_row(.data, ..., .before = NULL, .after = NULL) Returns copy of table of table. add_tally(). add_column(mtcars, new = 1:32) Add one or more rows to a table. grouped by .. ungroup(g_iris) add_row(faithful, eruptions = 1, waiting = 1) g_iris <- group_by(iris, Species) rename(.data, ...) Rename columns. rename(iris, Length = Sepal.Length) RStudio* is a trademark of RStudio, Inc. • CC BY SA RStudio • info@rstudio.com • 844-448-1212 • rstudio.com • Learn more with browseVignettes(package = c("dplyr", "tibble")) • dplyr 0.7.0 • tibble 1.2.0 • Updated: 2017-03

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