

# Working in the Tidyverse

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## Task 1

### Question A

```
#first we need to read in the tidyverse package
library(tidyverse)
```

Warning: package 'ggplot2' was built under R version 4.4.2

Warning: package 'lubridate' was built under R version 4.4.3

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.5
v forcats    1.0.0      v stringr    1.5.1
v ggplot2    3.5.1      v tibble     3.2.1
v lubridate  1.9.4      v tidyr      1.3.1
v purrr      1.0.2
```

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

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

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

```
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
#We cannot use read_csv because it can only read in comma and tab separated values (ours is '')
```

```
#Let's read it in with read_csv2 since it can handle ; delimited files
```

```
data <- read_csv2("data/data.txt", col_names = TRUE)
```

```
i Using "','" as decimal and "'.'" as grouping mark. Use `read_delim()` for more control.
Rows: 2 Columns: 3-- Column specification -----
Delimiter: ";"
dbl (3): x, y, z
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
#Print the output
data
```

```
# A tibble: 2 x 3
      x     y     z
  <dbl> <dbl> <dbl>
1     1     2     3
2     5     3     8
```

## Question B

```
#Read in the data
data2 <- read_delim("data/data2.txt", col_names = TRUE, delim = "6", col_types = "fdc")

#print the data
data2
```

```
# A tibble: 3 x 3
      x     y z
  <fct> <dbl> <chr>
1 1     2 3
2 5     3 8
3 7     4 2
```

## Task 2

### Question A

```
#Read in the trailblazer file
trailblazer <- read.csv("data/trailblazer.csv")

#Glimpse the data
glimpse(trailblazer)
```

```
Rows: 9
Columns: 11
$ Player      <chr> "Damian Lillard", "CJ McCollum", "Norman Powell", "Robert ~
$ Game1_Home  <int> 20, 24, 14, 8, 20, 5, 11, 2, 7
$ Game2_Home  <int> 19, 28, 16, 6, 9, 5, 18, 8, 11
$ Game3_Away  <int> 12, 20, NA, 0, 4, 8, 12, 5, 5
$ Game4_Home  <int> 20, 25, NA, 3, 17, 10, 17, 8, 9
$ Game5_Home  <int> 25, 14, 12, 9, 14, 9, 5, 3, 8
$ Game6_Away  <int> 14, 25, 14, 6, 13, 6, 19, 8, 8
$ Game7_Away  <int> 20, 20, 22, 0, 7, 0, 17, 7, 4
$ Game8_Away  <int> 26, 21, 23, 6, 6, 7, 15, 0, 0
$ Game9_Home  <int> 4, 27, 25, 19, 10, 0, 16, 2, 7
$ Game10_Home <int> 25, 7, 13, 12, 15, 6, 10, 4, 8
```

## Question B

```
#Pivot the data
trailblazer_longer <- trailblazer |>
  pivot_longer(cols = 2:11,
               names_to = "Location",
               values_to = "Points") |>
  separate_wider_delim(cols = "Location",
                       delim = "_",
                       names = c("Game", "Location"))

#Print first 5 rows
trailblazer_longer |>
  slice(1:5)
```

```
# A tibble: 5 x 4
  Player      Game Location Points
  <chr>      <chr> <chr>    <int>
1 Damian Lillard Game1 Home      20
2 Damian Lillard Game2 Home      19
```

3	Damian Lillard	Game3	Away	12
4	Damian Lillard	Game4	Home	20
5	Damian Lillard	Game5	Home	25

### Question C

On average, Jusuf Nurkic scored more points at home than away during the first 10 games

```
trailblazer_wider <- trailblazer_longer |>

#Create columns for home and away
pivot_wider(
  names_from = "Location",
  values_from = "Points") |>

#group so that the mean is calculated per player
group_by(Player) |>

#find means and difference for home vs away
mutate(mean_home = mean(Home, na.rm = TRUE),
       mean_away = mean(Away, na.rm = TRUE),
       mean_diff = mean_home - mean_away)

#Print the first 11 rows
trailblazer_wider |>
  slice(1:11)
```

```
# A tibble: 90 x 7
# Groups:   Player [9]
  Player      Game  Home  Away mean_home mean_away mean_diff
  <chr>      <chr> <int> <int>     <dbl>     <dbl>     <dbl>
1 Anfernee Simons Game1     11    NA      12.8      15.8      -2.92
2 Anfernee Simons Game2     18    NA      12.8      15.8      -2.92
3 Anfernee Simons Game3     NA     12      12.8      15.8      -2.92
4 Anfernee Simons Game4     17    NA      12.8      15.8      -2.92
5 Anfernee Simons Game5      5    NA      12.8      15.8      -2.92
6 Anfernee Simons Game6     NA     19      12.8      15.8      -2.92
7 Anfernee Simons Game7     NA     17      12.8      15.8      -2.92
8 Anfernee Simons Game8     NA     15      12.8      15.8      -2.92
9 Anfernee Simons Game9     16    NA      12.8      15.8      -2.92
10 Anfernee Simons Game10    10    NA      12.8      15.8      -2.92
# i 80 more rows
```

## Task 3

### Question A

1. Meaning of : There aren't any of this species on these islands, so there are no values so its an empty cell.
2. Meaning of <dbl [52]: There is a vector with 52 numeric (specifically double) elements
3. Meaning of : These variables are stored as lists

```
#read in the palmerpenguins package  
library(palmerpenguins)
```

Warning: package 'palmerpenguins' was built under R version 4.4.3

```
#run the code provided by colleagues  
penguins1 <- penguins |>  
select(species, island, bill_length_mm) |>  
pivot_wider(  
  names_from = island, values_from = bill_length_mm  
)
```

Warning: Values from `bill\_length\_mm` are not uniquely identified; output will contain list-cols.

```
* Use `values_fn = list` to suppress this warning.  
* Use `values_fn = {summary_fun}` to summarise duplicates.  
* Use the following dplyr code to identify duplicates.  
{data} |>  
  dplyr::summarise(n = dplyr::n(), .by = c(species, island)) |>  
  dplyr::filter(n > 1L)
```

```
view(penguins)
```

### Question B

```
#create the desired table (I dropped the grouping, this was unclear in the directions)  
penguins2 <- penguins |>  
  group_by(species) |>  
  summarise(  
    
```

```

    Bisco = sum(island == "Biscoe", na.rm = TRUE),
    Dream = sum(island == "Dream", na.rm = TRUE),
    Torgersen = sum(island == "Torgersen", na.rm = TRUE),
    .groups = "keep"
)

penguins2

```

```

# A tibble: 3 x 4
# Groups:   species [3]
  species    Bisco Dream Torgersen
  <fct>    <int> <int>    <int>
1 Adelie      44    56        52
2 Chinstrap    0    68         0
3 Gentoo    124     0         0

```