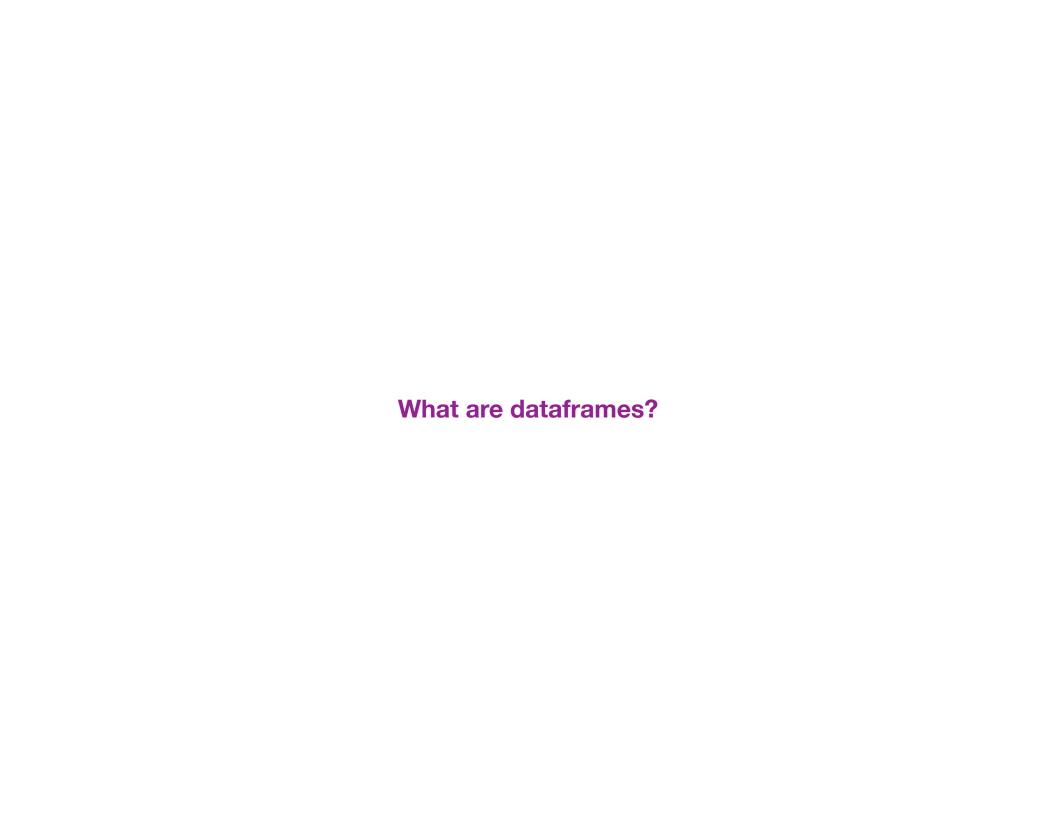
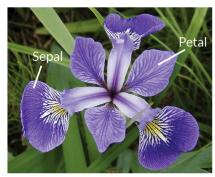
# **Introduction to dataframes**



QB Bootcamp, Day 2 Thursday, 2 September 2021 1:00pm - 1:30pm



#### Fisher's "Iris" dataset is a famous example dataset in statistics and dataviz







**Iris Versicolor** 

**Iris Setosa** 

Iris Virginica

R. A. Fisher (1936).

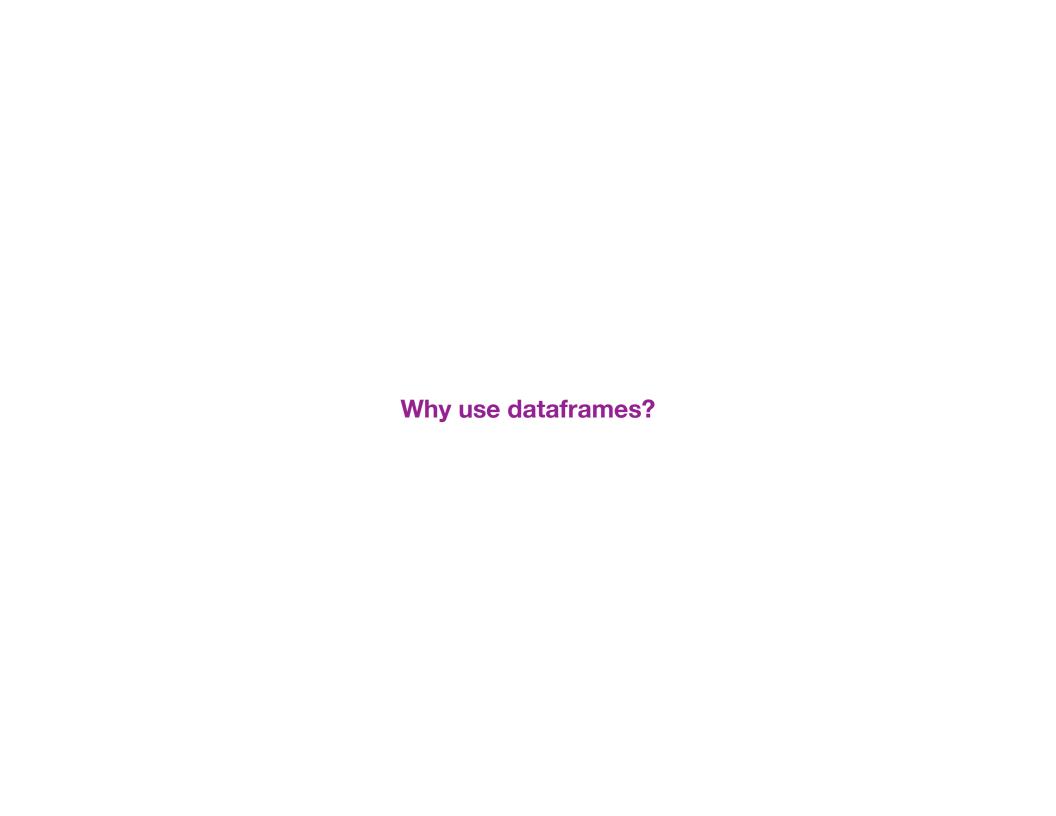
"The use of multiple measurements in taxonomic problems".

**Annals of Eugenics**. **7** (2): 179–188. (data collected by Edgar Anderson)

150 rows (50 per species)

|  |   | sepai_lengtn | sepai_width | petal_length | petal_width | species |
|--|---|--------------|-------------|--------------|-------------|---------|
|  | 0 | 5.1          | 3.5         | 1.4          | 0.2         | setosa  |
|  | 1 | 4.9          | 3.0         | 1.4          | 0.2         | setosa  |
|  | 2 | 4.7          | 3.2         | 1.3          | 0.2         | setosa  |
|  | 3 | 4.6          | 3.1         | 1.5          | 0.2         | setosa  |
|  | 4 | 5.0          | 3.6         | 1.4          | 0.2         | setosa  |

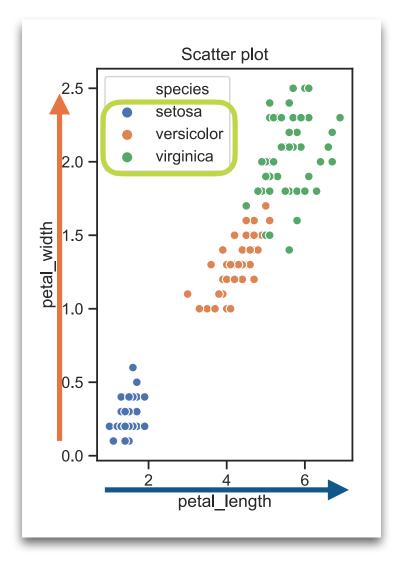
:



## **Dataframes greatly facilitate data visualization**

#### "aesthetics"

|    |              |             | $\mathcal{X}$ | У           | color   |
|----|--------------|-------------|---------------|-------------|---------|
|    | sepal_length | sepal_width | petal_length  | petal_width | species |
| 0  | 5.1          | 3.5         | 1.4           | 0.2         | setosa  |
| 1  | 4.9          | 3.0         | 1.4           | 0.2         | setosa  |
| 2  | 4.7          | 3.2         | 1.3           | 0.2         | setosa  |
| 3  | 4.6          | 3.1         | 1.5           | 0.2         | setosa  |
| 4  | 5.0          | 3.6         | 1.4           | 0.2         | setosa  |
| 5  | 5.4          | 3.9         | 1.7           | 0.4         | setosa  |
| 6  | 4.6          | 3.4         | 1.4           | 0.3         | setosa  |
| 7  | 5.0          | 3.4         | 1.5           | 0.2         | setosa  |
| 8  | 4.4          | 2.9         | 1.4           | 0.2         | setosa  |
| 9  | 4.9          | 3.1         | 1.5           | 0.1         | setosa  |
| 10 | 5.4          | 3.7         | 1.5           | 0.2         | setosa  |
| 11 | 4.8          | 3.4         | 1.6           | 0.2         | setosa  |
| 12 | 4.8          | 3.0         | 1.4           | 0.1         | setosa  |
| 13 | 4.3          | 3.0         | 1.1           | 0.1         | setosa  |
| 14 | 5.8          | 4.0         | 1.2           | 0.2         | setosa  |
|    |              |             |               |             |         |

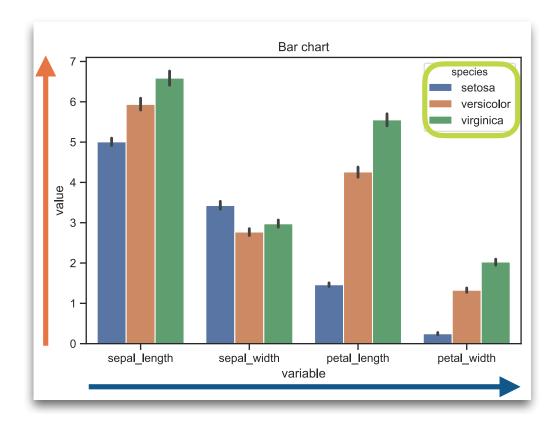


### **Dataframes facilitate important data-organizational transformations**

|   | sepal_length | sepal_width | petal_length | petal_width | species |                   | species | variable     | value |
|---|--------------|-------------|--------------|-------------|---------|-------------------|---------|--------------|-------|
| 0 | 5.1          | 3.5         | 1.4          | 0.2         | setosa  | 0                 | setosa  | sepal_length | 5.1   |
| 1 | 4.9          | 3.0         | 1.4          | 0.2         | setosa  | melt <sub>1</sub> | setosa  | sepal_length | 4.9   |
| 2 | 4.7          | 3.2         | 1.3          | 0.2         | setosa  | 2                 | setosa  | sepal_length | 4.7   |
| 3 | 4.6          | 3.1         | 1.5          | 0.2         | setosa  | 3                 | setosa  | sepal_length | 4.6   |
| 4 | 5.0          | 3.6         | 1.4          | 0.2         | setosa  | 4                 | setosa  | sepal_length | 5.0   |

individual = 1 flower

individual = 1 measurement

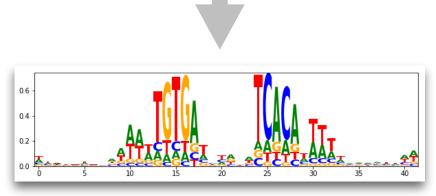




# We will start by working with a dataframe listing transcription factor binding sites in the database RegulonDB

|    | tf   | site   |
|----|------|--|
| 0  | AcrR | gcgttagattTACATACATTTGTGAATGTATGTAccatagcacg |
| 1  | AcrR | cgtgctatggTACATACATTCACAAATGTATGTAaatctaacgc |
| 2  | AcrR | catcggtcaaTTCATTCATTtgacttatac               |
| 3  | AcrR | tcactacacgCACATACAACggaggggggc               |
| 4  | AcrR | atttattaccGTCATTCATTTCTGAATGTCTGTTtacccctatt |
| 5  | AcrR | gctttacctcAAGTTAACTTgaggaattat               |
| 6  | AcrR | ataattcctcAAGTTAACTTgaggtaaagc               |
| 7  | Ada  | ttcagacgctGCGCTTTGCTTTCATATTCCGGTTgtcgcgacgg |
| 8  | Ada  | ggtcaccatcACGCAAAAACCAACAATCTTGCGCtttaattttt |
| 9  | Ada  | caacaatcttGCGCTTTAATTTTTTCGCTGACAaggaagcttt  |
| 10 | Ada  | cgcattacatTGCTGGATAAGAATGTTTTAGCAAtctctttctg |
| 11 | AgaR | ttcgtaaaacTTTCGTTTCATTTCGTTTTGcctattaacg     |
| 12 | AgaR | ttgcctattaACGCCTTTCTATTAAGCAAAtgcaagccca     |
| 13 | AgaR | tttcagtgacTTTCATTATGTTTCTTTTGTgaatcagatc     |
| 14 | AgaR | aaccattatcTTTCGTTTTATTTTTATCTCaccatgacgc     |

- Load TF binding site database a Pandas dataframe
- 2. Filter for TF of choice
- 3. Filter for binding sites of the most common length
- 4. Make a sequence logo



CRP logo (from 358 sites)

#### We will then parse our computed replication profiles in the form of a data frame

|    | chromosome | start | stop | reads |
|----|------------|-------|------|-------|
| 0  | chrl       | 1     | 31   | 2     |
| 1  | chrl       | 32    | 62   | 0     |
| 2  | chrl       | 63    | 93   | 1     |
| 3  | chrl       | 94    | 124  | 0     |
| 4  | chrl       | 125   | 155  | 3     |
| 5  | chrl       | 156   | 186  | 0     |
| 6  | chrl       | 187   | 217  | 0     |
| 7  | chrl       | 218   | 248  | 0     |
| 8  | chrl       | 249   | 279  | 0     |
| 9  | chrl       | 280   | 310  | 0     |
| 10 | chrl       | 311   | 341  | 0     |
| 11 | chrl       | 342   | 372  | 0     |
| 12 | chrl       | 373   | 403  | 0     |
| 13 | chrl       | 404   | 434  | 0     |
| 14 | chrl       | 435   | 465  | 1     |
|    |            |       |      |       |

- 1. Load a .bed file as a Pandas dataframe
- 2. Filter for the chromosome of choice
- 3. Smooth # reads as a function of position
- 4. Plot replication profile

