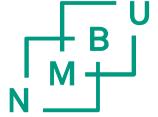


# DAT200 – Applied Machine Learning I

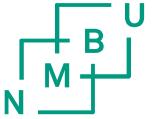
## Data Handling with Pandas

# Pandas for data handling (data wrangling)



- Assume that Anaconda Python distribution is installed on your computer
- [Pandas website and documentation](#)
- [Pandas community tutorials](#) (Official pandas website incl. videos)
- [Pandas getting started tutorials](#)
- [Searchable Pandas recipies](#)
- [DataFrames](#)
  - The primary pandas data structure
  - Two-dimensional size-mutable, potentially heterogeneous tabular data structure with labeled axes (rows and columns)
  - Arithmetic operations align on both row and column labels





# Some very common tasks with Pandas

- Create a Dataframe
  - <https://www.geeksforgeeks.org/different-ways-to-create-pandas-dataframe/>
- Load data from CSV-file into Dataframe
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_dataframe\\_importing\\_csv.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_dataframe_importing_csv.ipynb)
- Load data from web into Dataframe
  - `df = pd.read_csv('https://someplaceOnTheInternet.org/some.data')`

# Some very common tasks with Pandas – Part I

## ▪ Lecture exercise on Iris data

- Get data from: <https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data>
- Set column names to: sepal length, sepal width, petal length, petal width, types
- Set row names to: flower 1, flower 2, flower 3, ..., flower 150
- Hint: use input parameter `header=None` for `pd.read_csv` method
- Hint: use `DataFrame.index` to rename rows

Solution in file: `pandas_lectureExercise_part_I.py`

| Index     | sepal length | sepal width | petal length | petal width |
|-----------|--------------|-------------|--------------|-------------|
| flower 1  | 5.1          | 3.5         | 1.4          | 0.2         |
| flower 2  | 4.9          | 3           | 1.4          | 0.2         |
| flower 3  | 4.7          | 3.2         | 1.3          | 0.2         |
| flower 4  | 4.6          | 3.1         | 1.5          | 0.2         |
| flower 5  | 5            | 3.6         | 1.4          | 0.2         |
| flower 6  | 5.4          | 3.9         | 1.7          | 0.4         |
| flower 7  | 4.6          | 3.4         | 1.4          | 0.3         |
| flower 8  | 5            | 3.4         | 1.5          | 0.2         |
| flower 9  | 4.4          | 2.9         | 1.4          | 0.2         |
| flower 10 | 4.9          | 3.1         | 1.5          | 0.1         |
| flower 11 | 5.4          | 3.7         | 1.5          | 0.2         |

# Some very common tasks with Pandas – Part II



- Find Unique Values In Pandas Dataframes
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_find\\_unique\\_values.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_find_unique_values.ipynb)
- Grouping rows in Pandas
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_group\\_rows\\_by.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_group_rows_by.ipynb)
- Create a Column Based on a Conditional in Pandas
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_create\\_column\\_using\\_conditional.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_create_column_using_conditional.ipynb)

# Some very common tasks with Pandas – Part II

- Lecture exercise on iris data (continue from previous exercise)

- Find unique values for column `types` in your dataframe
- Compute the column mean for each type
- Create a new column in your dataframe named `sepal width >= 3` that contains `True` or `False`, depending on whether value in column `sepal width` is  $\geq 3.0$  (`True`) or  $< 3$  (`False`)
- Count how many times `sepal width` is  $\geq 3$  (you can use column `sepal width >= 3` for that)

| Index           | sepal length | sepal width | petal length | petal width |
|-----------------|--------------|-------------|--------------|-------------|
| Iris-setosa     | 5.006        | 3.418       | 1.464        | 0.244       |
| Iris-versicolor | 5.936        | 2.77        | 4.26         | 1.326       |
| Iris-virginica  | 6.588        | 2.974       | 5.552        | 2.026       |

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Solution in file: `pandas_lectureExercise_part_I_II.py`

| Index     | sepal length | sepal width | petal length | petal width | types           | sepal width >= 3 |
|-----------|--------------|-------------|--------------|-------------|-----------------|------------------|
| flower 51 | 7            | 3.2         | 4.7          | 1.4         | Iris-versicolor | True             |
| flower 52 | 6.4          | 3.2         | 4.5          | 1.5         | Iris-versicolor | True             |
| flower 53 | 6.9          | 3.1         | 4.9          | 1.5         | Iris-versicolor | True             |
| flower 54 | 5.5          | 2.3         | 4            | 1.3         | Iris-versicolor | False            |
| flower 55 | 6.5          | 2.8         | 4.6          | 1.5         | Iris-versicolor | False            |
| flower 56 | 5.7          | 2.8         | 4.5          | 1.3         | Iris-versicolor | False            |
| flower 57 | 6.3          | 3.3         | 4.7          | 1.6         | Iris-versicolor | True             |
| flower 58 | 4.9          | 2.4         | 3.3          | 1           | Iris-versicolor | False            |
| flower 59 | 6.6          | 2.9         | 4.6          | 1.3         | Iris-versicolor | False            |
| flower 60 | 5.2          | 2.7         | 3.9          | 1.4         | Iris-versicolor | False            |
| flower 61 | 5            | 2           | 3.5          | 1           | Iris-versicolor | False            |
| flower 62 | 5.9          | 3           | 4.2          | 1.5         | Iris-versicolor | True             |



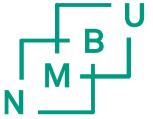
# Some very common tasks with Pandas – Part III

- Filter Pandas Dataframes
  - [https://github.com/chrisalbon/code\\_py/blob/master/filter\\_items\\_in\\_list\\_with\\_filter.ipynb](https://github.com/chrisalbon/code_py/blob/master/filter_items_in_list_with_filter.ipynb)
- Descriptive Statistics For Pandas Dataframe
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_dataframe\\_descriptive\\_stats.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_dataframe_descriptive_stats.ipynb)
- Count values in Pandas Dataframe
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_dataframe\\_count\\_values.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_dataframe_count_values.ipynb)
- Dropping Rows And Columns In Pandas Dataframe
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_dropping\\_column\\_and\\_rows.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_dropping_column_and_rows.ipynb)
- Search A Pandas Column For A Value
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_search\\_column\\_for\\_value.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_search_column_for_value.ipynb)
- Selecting Pandas DataFrame Rows Based On Conditions
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_selecting\\_rows\\_on\\_conditions.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_selecting_rows_on_conditions.ipynb)
- Sorting Rows In Pandas Dataframes
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_sorting\\_rows\\_dataframe.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_sorting_rows_dataframe.ipynb)

## Some very common tasks with Pandas – Part III



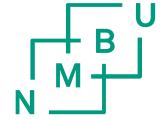
- Lecture exercise on iris data (continue from previous exercise)
  - Create three data subsets from original dataframe (one for setosa, one for versicolor, one for virginica). Use conditional row selection based on column `types`
  - Count how many times each class occurs (Answer: 50 of each class)



## Some very common tasks with Pandas – Part III

- Applying Operations Over Pandas Dataframes
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_apply\\_operations\\_to\\_dataframes.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_apply_operations_to_dataframes.ipynb)
- Pivot Tables In Pandas
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_pivot\\_tables.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_pivot_tables.ipynb)
- Selecting Pandas DataFrame Rows Based On Conditions
  - [https://github.com/chrisalbon/code\\_py/blob/master/pandas\\_selecting\\_rows\\_on\\_conditions.ipynb](https://github.com/chrisalbon/code_py/blob/master/pandas_selecting_rows_on_conditions.ipynb)

## Some very common tasks with Pandas – Part III



### ▪ Lecture exercise on iris data (continue from previous exercise)

- View last 10 rows of columns `sepal length` and `types`
- View rows where `sepal length > 5` and `petal width < 0.2`
- Make a new dataframe containing only rows where `petal width` is exactly 1.8
- Get descriptive statistics for the whole dataframe and afterward for column `petal length`
- Remove rows named `flower 55` and `flower 77`
- Remove column `sepal width >= 3`
- View all rows of `sepal length` where `petal width` is exactly 1.8
- Get values of the dataframe stored in a numpy array (in practice get rid of columns and rows)
- Remove column `types` and apply a function named `computation` to each cell in dataframe. Function `computation` should do the following: take the value of the cell, add 1 and multiply that by 3

