### Business 4720 - Class 6

### Data Management in Python using Pandas

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### This Class

#### What You Will Learn:

- ► Introduction to Python
- Introduction to the the Numpy package
- ► Introduction to the Pandas package



### Intro to Python

### What is Python?

- Readability and simplicity
- Dynamic typing enhancing flexibility
- Extensive libraries
- Procedural, object-oriented, and functional programming
- Widely used in data analysis, AI, scientific computing, etc.
- Easy to learn
- Active community support

#### **Intro Tutorial:**

```
https://python.swaroopch.com/
https://github.com/swaroopch/byte-of-python/
releases/
```

# Running Python

- Interactive Python Shell (command line)
- 2 Jupyter Notebooks
- 3 PyCharm IDE



### Interactive Python Shell

- Similar to R
- Type "python" to launch Python interpreter
- Prompt is "> > >", type ENTER to execute a command
- ► Use quit () to exit
- Tip: Use a notepad app to assemble commands and to keep results



### Interactive Python Shell

```
joerg@joerg-samsung:~$ python
Python 3.10.12 (main, Nov 20 2023, 15:14:05) [GCC 11.4.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>>
```

# Jupyter Notebooks

- ► Interactive computing environment
- Notebook Interface
- Combine executable code, text, visualizations
- Create and share documents with live code, equations, and explanatory text
- Collaborative editing of notebooks (on web-based services)
- Popular for Python, but can handle other languages



### JupyterLabs Desktop

Welcome



# **Jupyterlab**

#### Start

New notebook...



Open File...



Connect...

#### Recent sessions

joerg /home

#### Jupyter News

Open Community Call

And Voici!

Plug your application into the Jupyter world

Voilà 0.5.0 : Homecoming

Bringing Modern JavaScript to the Jupyter Notebook

Desktop GIS software in the cloud with JupyterHub

Generative AI in Jupyter

European Commission Funds Jupyter Bug Bounty Program

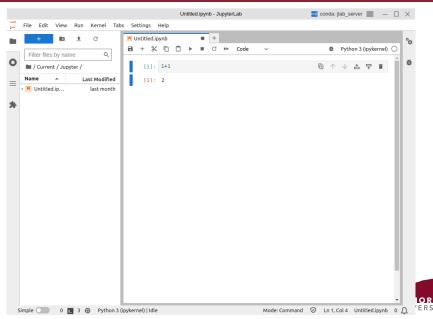
Announcing Jupyter Notebook 7

JupyterCon 2023 recordings now live on YouTube!





# JupyterLabs Desktop



# Jupyter Notebooks

- "Kernel" is the Python interpreter and environment that runs your code
- Enter code into empty cell
- ► Press CTRL-ENTER to execute a cell
- ► Merge, split, move, copy, delete cells
- Save, import, export notebooks

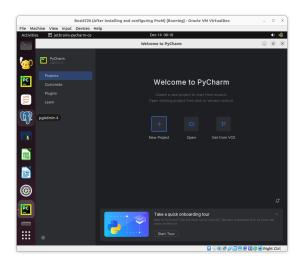


# PyCharm IDE

- ▶ When working with multiple Python files in your project
- Useful for programming (defining functions, classes; using control structures, etc.) rather than just scripting (executing a few Python commands one after the other)
- Contains built-in debugging tools



# PyCharm IDE





# Basic Python

### Printing a string:

```
print('hello world')
```

### String format method:

```
age = 19
name = 'Malina'
print('{0} is {1} years old'.format(name, age))
print('{name} is {age} years old'.format(name=name,age=age))
print('{} is {} years old'.format(name, age))
print(f'{name} is {age} years old')
print(name+' is '+str(age)+' years old')
```



# Basic Python

### Backslashes split and continue lines:

```
print('This is a very long \
string and needs a second line')
i = \
5
print(i)
```

### Multiline strings:

```
s = '''This is line 1
and here is line 2
and now this is line 3'''
print(s)
```



# Basic Python

### Python knows math:

```
2 + 2

2**4

13 // 3

-13 // 3

13 % 3

-25.5 % 2.25

3 < 5

3 > 5

3 == 5

(3 < 5) and (4 < 2)

(3 < 5) or not (4 < 2)
```

### Strings

### Python knows strings:

```
language = 'Innuktitut'
if language.startswith('Innu'):
    print('Yes, the string starts with "Innu"')
if 'u' in language:
    print('Yes, it contains the string "u"')
if language.find('nuk') != -1:
    print('Yes, it contains the string "nuk"')
# Joining and Splitting
delimiter = ' * '
mylist = ['Nain', 'Hopedale', 'Makkovik', 'Rigolet']
mystring = delimiter.join(mylist)
print (mystring)
thelist = mystring.split(delimiter)
print (thelist)
```

### Lists

#### Lists are ordered collections of items:

```
# Inuit deities
gods = ['Sedna', 'Nanook', 'Akna', 'Pinga']
print('There are', len(gods), 'deities:')
for item in gods:
    print(item, end=' ')
gods.append('Amaguq')
print('\nThe list of deities is now', gods)
gods.sort()
print ('The sorted list is', gods)
print('The first deity is', gods[0])
olditem = gods[0]
del gods[0]
print('I removed', olditem)
print('The list is now', gods)
```

### **Tuples**

### Tuples are immutable:

```
# Inuit Nunangat
regions = ('Inuvialuit', 'Nunavut', \
           'Nunavik', 'Nunatsiavut')
print('Number of regions is', len(regions))
all regions = 'NunatuKayummiut', 'Kalaallit', \
              'Inupiag', regions
print('Number of all Inuit regions:',len(all_regions))
print ('All Inuit regions are', all regions)
print('Regions in Inuit Nunangat are', all_regions[3])
print('First region in Inuit Nunangat is', \
      all regions[3][1])
print('Number of all Inuit regions is', \
      len(all_regions)-1+len(all_regions[3]))
```



### **Dictionaries**

- Key-value pairs
- Associative arrays
- ► Map

```
# Largest citites
C = {
    'Inuvialuit': 'Inuvik',
    'Nunavut': 'Iqaluit',
    'Nunavik': 'Kuujjuag',
    'Nunatsiavut': 'Nain'
print(c.kevs())
print(c.values())
print("Nunavik's largest city is", c['Nunavik'])
# Deleting a key-value pair
del c['Nunavut']
print('\nThere are {} cities left\n'.format(len(c)))
for region, city in c.items():
    print('{} is largest city of {}'.format(city, region))
# Adding a key-value pair
c['Nunavut'] = 'Iqaluit'
if 'Nunavut' in c:
    print("\nNunavut's largest city is", c['Nunavut'])
```

# Structured Data Types

### **Important**

- ► Indexing begins at 0 (different from R!)
- Can contain any data type

### Sequences

- List, tuples, strings are sequences
- ▶ Membership tests using in or not in
- Indexing and slicing



# Slicing

```
regions = ('Inuvialuit', 'Nunavut',
           'Nunavik', 'Nunatsiavut')
language = 'Innuktitut'
# Slicing on a tuple
print('Item 1 to 3 is', regions[1:3])
print('Item 2 to end is', regions[2:1)
print('Item 1 to -1 is', regions[1:-1])
print('Item start to end is', regions[:])
# Slicing on a string
print('characters 1 to 3 is', language[1:3])
print('characters 2 to end is', language[2:])
print('characters 1 to -1 is', language[1:-1])
print('characters start to end is', language[:])
# Slicing with step
print(regions[::1])
print(regions[::2])
print(regions[::3])
print(regions[::-1])
```

#### Lists

- 1 Create a list containing the numbers 1 to 10. Use list slicing to create a sublist with only the even numbers.
- 2 Using a for loop, sum all the items in the list.
- 3 Using a for loop, iterate over the list and print each number squared.
- Write a program to append the square of each number in the range [1:5] to a new list.

### **Tuples**

- 1 Create a tuple with different data types (string, int, float).
- 2 Demonstrate how tuples are immutable by attempting to change its first element.
- Write a program to convert the tuple into a list.



#### **Dictionaries**

- Create a dictionary with at least three key-value pairs, where the keys are strings and the values are numbers.
- 2 Write a Python script to add a new key-value pair to the dictionary and then print the updated dictionary.
- 3 Create a nested dictionary and demonstrate accessing elements at various levels.

# Numerical Data in Python with NumPy

### What is Numpy?

- High-performance scientific computing and data analysis.
- Multidimensional arrays
- Comprehensive mathematical function library
- ► Foundational package for other scientific libraries like SciPy, Pandas, Matplotlib, scikit-learn, scikit-image, etc.

#### Intro Tutorials:

```
https:
//numpy.org/doc/stable/user/quickstart.html
https://numpy.org/doc/stable/user/absolute_
beginners.html
```



# NumPy Array

### N-Dimensional Array, type "ndarray"

ndarray.ndim	Number of axes
ndarray.shape	Typle describing the size of each dimension (axis)
ndarray.size	Total number of elements
ndarray.dtype	The datatype of the elements, e.g. numpy.int32, numpy.int16, numpy.float32, numpy.float64
ndarray.itemsize	Number of bytes for each element



# NumPy Basics

```
# Import the numpy package
import numpy as np

# Create an array
a = np.arange(15).reshape(3, 5)
print(a.shape)
print(a.ndim)
print(a.dtype.name)
print(a.size)
print(type(a))
```



# NumPy Basics

```
# Create an array from Python lists and tuples
b = np.array([(1.5, 2., 3), (4, 5, 6)])
print(b)
# Elementwise operations
print(3 * b)
print(b + 5)
print(np.sqrt(b))
# Array operations
print(np.max(b))
print(np.max(b, axis=0))
print(np.max(b, axis=1))
print(np.std(b))
print (np.cov(b))
print(np.sum(b))
```



# NumPy Basics [cont'd]

```
# Create an array of zeros with shape (3,4)
x = np.zeros((3,4))
print(x)

# Create an array of ones with shape (2,3,4)
y = np.ones((2,3,4))
print(y)
```

# **Array Slicing**

► Each axis can be sliced using [:] or [::]



# Array Slicing & Iterators

```
c = np.array([[[ 0, 1, 2],
              [ 10, 12, 13]],
              [[100, 101, 102],
               [110, 112, 113]])
print (c.shape)
print(c[1, ...])
print(c[1, :, :])
print(c[..., 2])
print(c[:,:,2])
print(c[..., : , 1])
for row in b:
   print (row)
for element in b.flat:
   print(element)
```

# Array Reshaping

```
rg = np.random.default_rng(1)
a = np.floor(10 * rg.random((3, 4)))

print(a.shape)
print(a.flatten())
print(a.reshape(6, 2))
print(a.T)
print(a.T.shape)

b = np.floor(5 * rg.random((3, 4)))
print(np.vstack((a, b)))
print(np.hstack((b, a)))
```



# Array Indexing

# Unique Elements and Counts

- Create an array with random numbers in the shape indicated by the last four digits of your student number (if your student number contains a 0, use a 1 instead)
- 2 Construct a new array by swapping the first half of rows (axis 0) with the second half of rows (axis 0)
- 3 Calculate all covariance matrices formed by the last two axes of your array. Tip: Iterate over the first two axes/dimensions with a for loop
- 4 Subtract the mean of the array from each element in the array (mean normalization)
- 5 Select all elements that are greater than the overall mean
- 6 Sort the selected elements from the previous step



# Data Management with Pandas

#### What is Pandas?

- Open-source library for data analysis
- High-performance, easy-to-use data structures and data analysis tools
- ► Can handle tabular data, time series, matrix data, etc.
- Tools for data cleaning, transformation, and preparation
- ► Importing data from CSV, Excel, SQL databases, etc.
- Functions for aggregating, pivoting, joining, and sorting data

#### Intro Tutorial:

```
http:
```

//pandas.pydata.org/docs/user\_guide/10min.htm

#### Pandas Series

- ► 1-Dimensional *labeled* array
- Axis labels are called index

### Pandas Series [cont'd]

```
# Series behave like an ndarray
print(s.iloc[0])
print(s.iloc[:3])
print(s[s > s.median()])
print(s.iloc[[4, 3, 1]])
print(np.exp(s))
# Series behave like a dict.
print(s['a'])
print(s['e'])
print('e' in s)
print('f' in s)
# Series have a datatype and name
s.name = 'My First Series'
print(s.dtype)
```

#### Pandas Dataframe

- 2-Dimensional
- Columns may have different data types
- ► Conceptually a dict of pandas series

#### Pandas Dataframe Columns

```
print(df['one'])
df['three'] = df['one'] * df['two']
df['flag'] = df['one'] > 2
print(df)
del df['two']
three = df.pop('three')
df['foo'] = 'bar'
df['one_trunc'] = df['one'][:2]
df.insert(1, 'bar', df['one'])
print (df)
# Similar to 'mutate' in R/Dplyr
df = df.assign(four = df['one'] * np.sqrt(df['bar']))
print(df)
```

# Dataframe Indexing

Select column	df[col]	Series
Select row by label	df.loc[label]	Series
Select row by integer location	df.iloc[loc]	Series
Slice rows	df[::]	DataFrame
Select rows by boolean vector	df[bool]	DataFrame

## Dataframe Alignment and Arithmetic

Data is aligned on column labels and row indices

```
df = pd.DataFrame(np.random.randn(10, 4),
                   columns=["A", "B", "C", "D"])
df2 = pd.DataFrame(np.random.randn(7, 3),
                    columns=["A", "B", "C"])
print(df + df2)
# Elementwise operators
print(df * 5 + 2)
print (1/df)
print (df * * 4)
# Transpose
print (df.T)
# Using Numpy functions
print(np.exp(df))
print (np.asarray(df))
```

#### Pandas Dataframe

```
df.info()
df.head()
df.tail(3)

# Boolean reductions
(df > 0).all()
(df > 0).any()
(df > 0).any()
(df > 0).any()
# NaN's are not the same
df.iloc[0,0] = np.nan
(df+df == df*2).all()
(df + df).equals(df*2)
```



## Descriptive Statistics, Aggregation, & Strings

```
# Descriptive statistics
df.mean(0)
df.mean(1, skipna=False)
df std = (df - df.mean()) / df.std()
df.describe()
# Aggregation with 'agg'
df.agg(['sum', 'mean', 'std'], 0)
# Sort by values
df.sort values(by=['A', 'B'])
df.nsmallest(3, 'A')
df.nlargest(3, 'A')
# String functions with 'str'
s = pd.Series(
    ["A", "B", "C", "Aaba", "Baca", np.nan,
     "CABA", "dog", "cat"], dtype="string")
s.str.lower()
```

## Selection with Query

### Selection with Query [cont'd]

```
df = pd.DataFrame({'a': list('aabbccddeeff'),
                   'b': list('aaaabbbbcccc'),
                    'c': np.random.randint(5, size=12),
                    'd': np.random.randint(9, size=12)})
# Pure Python versus Query
df[df['a'].isin(df['b'])]
df.query('a in b')
df[~df['a'].isin(df['b'])]
df.query('a not in b')
df[df['b'].isin(df['a']) & (df['c'] < df['d'])]
df.query('a in b and c < d')
df[df['b'].isin(["a", "b", "c"])]
df.query('b == ["a", "b", "c"]')
df[df['c'].isin([1, 2])]
df.query('[1, 2] in c')
```

# Duplicate Data

```
df2 = df.copy()

df2.duplicated(['a', 'b'])

df2.drop_duplicates(['a', 'b'], keep='last')

df2.drop_duplicates(['a', 'b'], keep='first')
```

## Reading Data into Pandas

- Wide variety of format: CSV, JSON, Excel, SQL, ...
- https://pandas.pydata.org/docs/user\_guide/
  io.html#

```
rentals = pd.read_csv('pagila/rentals.csv')
rentals['rental date'] = \
     pd.to_datetime(rentals['rental_date'], utc=True)
rentals['return date'] = \
     pd.to datetime(rentals['return date'], utc=True)
rentals['payment_date'] = \
     pd.to datetime (rentals ['payment date'], utc=True)
rentals.info()
rentals.describe()
rentals index
rentals.columns
rentals.shape
```

#### Examine the NA's

## Pagila Database in Python]

Find all films and the actors that appeared in them, ordered by film category and year, for those films that are rated PG:

```
actors = pd.read_csv('pagila/actors.categories.csv')
result = pd.merge(rentals, actors, on='title',
          suffixes=(' customer', ' actor'),
          how='outer')
result = result[result['rating'] == 'PG']
result['actor'] = result['last_name_actor'] + \
          ', ' + result['first_name_actor']
result.rename(columns={'release year': 'year'},
          inplace=True)
result = result[['actor', 'title', 'category', 'year']]
result.drop_duplicates(['actor', 'title', 'category', 'year'],
          inplace=True)
result.sort values(['category', 'vear', 'title'],
          inplace=True)
grouped = result.groupby(['category', 'year', 'title'])
g_result = grouped['actor'].apply(list).reset_index()
print (g_result)
```



### Pagila Database in Python

Find the most popular actors in the rentals in each city:

```
addresses = pd.read_csv('pagila/addresses.csv')
addresses['phone'] = addresses['phone'].astype(str)
joined df = pd.merge(rentals, addresses,
                     left on='customer address',
                     right_on='address id')
joined df = pd.merge(joined df, actors,
                     on='title'.
                     suffixes=('_customer', '_actor'))
joined_df['actor'] = joined_df['last_name_actor'] + \
              ', ' + joined_df['first_name_actor']
grouped = joined df.groupby(['city', 'actor']). \
            size().reset_index(name='count')
grouped['ranking'] = grouped.groupby('city')['count']. \
            rank (method='min', ascending=False)
filtered = grouped[grouped['ranking'] < 4]</pre>
sorted df = filtered.sort values(
               by=['city', 'ranking', 'actor'])
print (sorted df.head(25))
```

## Pagila Database in Python

Find the customers who spend the most on rentals, and the number of rentals with the higest total rental payments for each category grouped by rental duration.

### Pagila Database in Pythonc [cont'd]

#### ... continued from previous slide ...

```
grouped data = selected data.groupby( \
  ['category', 'rental duration', 'customer']).agg( \
    payments=pd.NamedAgg('amount', 'sum'), \
    num rentals=pd.NamedAgg('amount', 'count')).reset index()
grouped_data['ranking'] = \
  grouped_data.groupby(['category', 'rental_duration']) \
    ['payments'].rank (method='min', ascending=False)
top entries = grouped data.loc[
    grouped_data.groupby(['category', 'rental_duration']) \
    ['ranking'].idxmin() \
print (top_entries)
```



### Pagila Database in Python

Get the top 5 and the bottom 5 grossing customers for each quarter.

```
full data['customer'] = \
    full data['first name customer'] + ' ' + \
    full data['last name customer']
full data['q'] = pd.to datetime( \
    full data['rental date']).dt.to period("0")
selected data = \
    full_data[['customer', 'q', 'amount', 'rental_date']]
grouped data = \
    selected data.groupby(['g', 'customer']) \
        .agg(payments=('amount', 'sum')).reset index()
distinct data = \
    grouped_data.drop_duplicates(
        subset=['customer', 'q', 'payments'])
```

## Pagila Database in Python [cont'd]

... continued from previous slide ...

```
distinct_data['rank_top'] = \
    distinct_data.groupby('q')['payments'] \
    .rank (method='min', ascending=False)
distinct data['rank bot'] = \
    distinct data.groupby('g')['payments'] \
    .rank (method='min', ascending=True)
filtered data = \
    distinct data[
        (distinct data['rank top'] < 6) |
        (distinct data['rank bot'] < 6)]
sorted data = \
    filtered data.sort values (
        by=['q', 'payments'],
        ascending=[True, False])
print (sorted data)
```

### Pagila Database in Python [cont'd]

Find the set of film titles by rental customer and the total number rentals for each customer

```
full_data['customer'] = \
    full_data['first_name_customer'] + ' ' + \
    full_data['last_name_customer']
selected data = full data[['customer', 'title']]
grouped data = \
    selected data.groupby('customer')['title'] \
        .apply(list).reset_index(name='titles')
grouped_data['rentals'] = \
     grouped_data['titles'].apply(len)
grouped data['unique titles'] = \
     grouped_data['titles'].apply(lambda x: list(set(x)))
grouped data = grouped data.drop(columns=['titles'])
sorted_data = grouped_data.sort_values(by='customer')
print (sorted data)
```

#### Hands-On Exercises

- Find all films with a rating of 'PG'
- 2 List all customers who live in Canada (with their address)
- 3 Find the average actual rental duration for all films
  - ► This requires date arithmetic
- Find the average overdue time for each customer
  - This requires date arithmetic
- 5 List all films that have never been rented
- 6 List the names of actors who have played in more than 15 films

