

Vectorization

1405

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Required Python Libraries for Today

Core

- NumPy, Pandas

```
import numpy as np
import pandas as pd
```

Visualization

- Matplotlib, Seaborn

```
from matplotlib import pyplot as plt
import seaborn as sns
```

Statistical learning

- scikit-learn, SciPy, statsmodels

```
import sklearn
import scipy
import statsmodel as sm
```

NumPy & Pandas Printing Format

It's better to print only *the first few decimal digits* of large real numbers. Set the print options to keep 3 decimal digits and supress scientific notation (for NumPy arrays and Pandas DataFrames):

- **NumPy**

```
np.set_printoptions(precision=3, suppress=True)
```

- **Pandas**

```
pd.options.display.float_format = '{:.3f}'.format
```

Vectorization & Advanced Techniques

- Technically, any operation you can perform on a list can also be done on a NumPy array by *iterating through its elements*
- However, *loops are inefficient* for large data sets, and so this is where matrix algebra and vectorized operations can help
- NumPy provides built-in support for vectorized operations, making computations *faster* and the code *more concise*

Apply Functions Element-Wise to Array

- Often, you need to apply more complicated functions to each element of on a NumPy array
- Two options:
 - Use built-in functions: **universal functions** (ufuncs)
 - Write your own customized function and **vectorize** it

Universal Functions (ufuncs)

- Built-in functions that perform element-wise operations on a NumPy array without explicitly looping through an entire array
 - A simple example: `np.sqrt(a)` returns a new array of the same size as `a` that contains the square root of each element of `a`
- *ufuncs are implemented in C (programming language)*
 - *Optimized for speed (e.g., avoids Python loops) and one reason NumPy is faster than plain-vanilla Python*

Commonly Used Universal Functions

ufunc Description	Python Implementation
Square root: $x \rightarrow \text{sqrt}(x)$	<code>np.sqrt(a)</code>
Exponential: $x \rightarrow \text{exp}(x)$	<code>np.exp(a)</code>
Natural log: $x \rightarrow \ln(x)$	<code>np.log(a)</code>
Largest integer no greater than: $x \rightarrow \text{floor}(x)$	<code>np.floor(a)</code>
Smallest integer no less than: $x \rightarrow \text{ceil}(x)$	<code>np.ceil(a)</code>
Array maximum: $(x_1, \dots, x_n) \rightarrow \max\{x_1, \dots, x_n\}$	<code>np.fmax(a1, a2, ..., an)</code>
Array minimum: $(x_1, \dots, x_n) \rightarrow \min\{x_1, \dots, x_n\}$	<code>np.fmin(a1, a2, ..., an)</code>
Check for missing value	<code>np.isnan(a)</code>
Check for membership	<code>np.isin(a, master)</code>

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Vectorize a Function & Apply to Array

- `np.vectorize()` can apply *any function* element-wise to an array: it generalizes ufuncs but without speed improvements

Step 1: Define a function

```
def f(x):  
    ...  
    return result
```

Step 2: Vectorize the function

```
vec_f = np.vectorize(f)
```

Step 3: Apply the vectorized function to an array

```
result = vec_f(a)
```