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## GEOL 6670 Geophysical Inverse Theory

## Tutorial on Python

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## Why Python?

- Versatile
  - A personal story: My old workflow is to use C/Fortran for computing, then use Shell/Perl for I/O, and Matlab/GMT for plotting.
- Easy to read

```
def add(a, b):
    # Return the sum of two numbers.
    return a + b
```

- "Efficient"
  - I believe human time is more precious than machine time, and thus appreciate the fast feedback from Python in the development stage.
- Open-source
- Popular

**Fig: Trends of programming languages in [StackOverflow]**

([https://insights.stackoverflow.com/trends?](https://insights.stackoverflow.com/trends?tags=java%2Cc%2Cc%2B%2B%2Cpython%2Cc%23%2Cvb.net%2Cjavascript%2Casc)

[tags=java%2Cc%2Cc%2B%2B%2Cpython%2Cc%23%2Cvb.net%2Cjavascript%2Casc](https://insights.stackoverflow.com/trends?tags=java%2Cc%2Cc%2B%2B%2Cpython%2Cc%23%2Cvb.net%2Cjavascript%2Casc)  
c).

A google search of `python tutorial` yields ~1 billion results, thus I would not try to provide a list of useful resources. In addition, the tutorial aims to be useful for a diverse audience but will become more biased in the later part.

Some examples are provided only as counterparts to Menke's introduction to Matlab.

## Python syntax

### Variables and formats

In [1]: *# Single line comment*

"""

Multiple  
line  
comments.  
"""

*# Integer division*

5 // 3

-5 // 3

*# Exponentiation*

2\*\*3

*# Boolean*

True

not True

True or False

True and False

*# Comparison*

1 == 1

1 != 2

1 < 2

2 <= 2

1 < 2 < 3

x = 0

print('temperature data from {} to {}'.format(1.1, 2))

print('{0} and {1}'.format('spam', 'eggs'))

print('{1} and {0}'.format('spam', 'eggs'))

print('x = ' + str(x))

print(f'x = {x}')

print(f'x = {x:.3f}')

temperature data from 1.1 to 2

spam and eggs

eggs and spam

x = 0

x = 0

x = 0.000

## Data structures

```

In [ ]: # List
li = []
li.append(1)
li.append(2)
li.append(3)
li[1]
li[1:2]
len(li)
a, b, c = li
2 in li

# Tuple
tup = (1, 2, 3)
# tup[0] = 2    # Tuples are immutable!

# Dictionary
di = {'one': 1, 'two': 2, 'three': 3}
di['one']
list(di.keys())
list(di.values())
'one' in di
di.get('one')
di.update({'four': 4})

# Set
se = set({1, 1, 2, 2, 3})
se.add(4)
other_set = {3, 4}
se & other_set    # intersection
se | other_set    # union
se - other_set    # difference

```

## Control flows

```
In [ ]: x = 5
        if x > 1:
            print('x > 1')
        elif x < 1:
            print('x < 1')
        else:
            print('x == 1')

s = 0
for i in range(1, 10, 1):
    s += i
s

x = 0
while x < 2:
    print(x)
    x += 1

# try:
#     open('a/b')
# except FileNotFoundError as e:
#     raise e
# else:
#     pass
# finally:
#     pass
```

**I/O (cf. using [numpy](#), [pandas](#))**

```
In [ ]: fpath = '/Volumes/GoogleDrive/My Drive/6670/gda/data/global_temp.txt'
with open(fpath, 'r') as f:
    #     d = f.read()
    #     d = f.readline()

    #     d = f.readlines()
    d = list(f)

    #     for i, line in enumerate(f):
    #         print(i, line)

d

year, temperature = [], []
with open(fpath, 'r') as f:
    for line in f:
        y, t = line.split()
        year.append(int(y))
        temperature.append(float(t))

# year, temperature
```

## Functions

```
In [ ]: def function_name(arg1, *args, **kwargs):
        print(arg1)
        print(args)
        print(kwargs)

        return

arg1 = 1
args = (1, 2)
kwargs = {'a': 1}
# function_name(arg1)
function_name(arg1, *args, **kwargs)
```

## Modules

```
In [ ]: import math
```

```
In [ ]: a = 6
b = 8
c = math.sqrt(a**2 + b**2)
c
```

```
In [ ]: n = 3
        x = 4
        x0 = 1
        L = 6
        c = math.sin(n * math.pi * (x-x0) / L)
        c
```

```
In [ ]: from math import ceil
        ceil(3.2)
```

## Classes

```

In [ ]: class Human(object):
        def __init__(self, name, skills=[]):
            self.name = name
            self.skills = skills

        def solve_problems(self):
            return True

class Geologist(Human):
    def __init__(self, name, skills=['geology']):
        super().__init__(name=name, skills=skills)
#         super(Human, self).__init__(name=name, skills=skills)

    def solve_geological_problems(self):
        return True

class Physicist(Human):
    def __init__(self, name, skills=['physics']):
        super().__init__(name=name, skills=skills)

    def solve_physics_problems(self):
        return True

class Geophysicist(Human):
    def __init__(self, name, skills=['geophysics']):
        super().__init__(name=name, skills=skills)

    def solve_geophysical_problems(self):
        return True

class GeoPhysicist(Geologist, Physicist):
    def __init__(self, name, skills=['geology', 'physics']):
        Geologist.__init__(self, name=name, skills=skills)
        Physicist.__init__(self, name=name, skills=skills)

p1 = GeoPhysicist(name='anonymous')
p1.name
p1.skills
p1.solve_problems()
# p1.solve_geological_problems()
# p1.solve_geophysical_problems()

```

## Command line

```

In [ ]: pwd

```



```
In [ ]: cd /Volumes/GoogleDrive/My\ Drive/6670
```

```
In [ ]: # dir()
```

```
In [ ]: ls -lh
```

## Essential packages

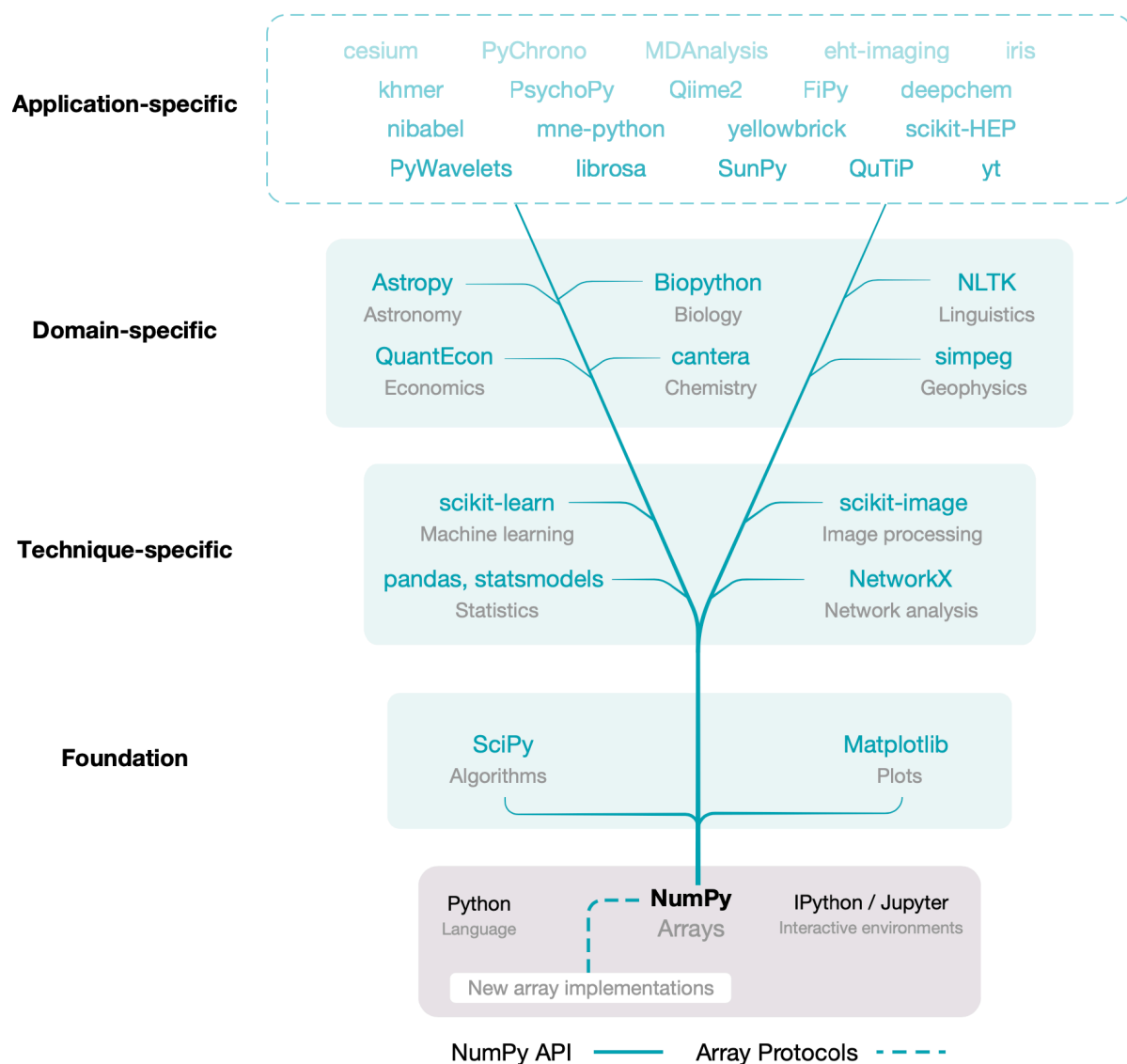
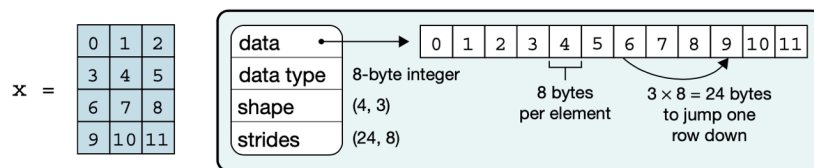
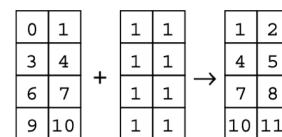
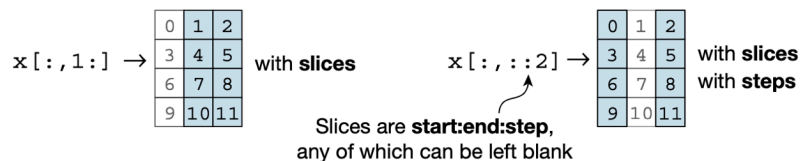
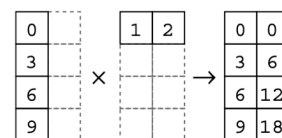
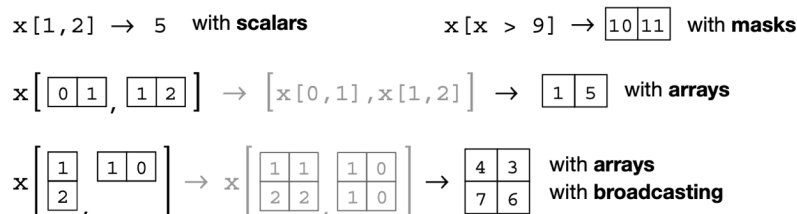
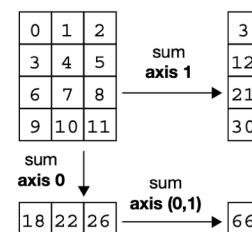


Fig: [Harris et al., 2020](<https://doi.org/10.1038/s41586-020-2649-2>).

## Array programming with NumPy

**a** Data structure**d** Vectorization**b** Indexing (view)**e** Broadcasting**c** Indexing (copy)**f** Reduction**Fig:** [Harris et al., 2020](https://doi.org/10.1038/s41586-020-2649-2).

```
In [ ]: import numpy as np
```

```
In [ ]: # Data structure
x = np.arange(12).reshape(4, 3)
x.dtype
x.shape
x.strides

# Indexing (view)
x[:, 1:]
x[:, ::2]

# Indexing (copy)
x[1, 2]
x[x > 9]
x[[0, 1], [1, 2]]
x[[[1], [2]], [1, 0]]

# Vectorization
a = np.column_stack([np.linspace(0, 9, 4), np.linspace(1, 10, 4)])
b = np.ones((4, 2))
a + b

# Broadcasting
a[:, 0].reshape(4, 1) * np.array([1, 2]).reshape(1, 2)

# Reduction
x.sum(axis=0)
x.sum(axis=1)
x.sum()
```

```
In [ ]: r = np.array([2, 4, 6]).reshape(1, 3)
c = np.array([1, 3, 5]).reshape(3, 1)

r, c
```

```
In [ ]: a = r[0, 1]
b = c[2, 0]
a, b
```

```
In [ ]: M = np.arange(1, 10, 1).reshape(3, 3)
M
```

```
In [ ]: c = M[0, 2]
c
```

```
In [ ]: M = np.array([
    1, 0, 2,
    0, 1, 0,
    2, 0, 1,
]).reshape(3, 3)
N = np.array([
    1, 0, -1,
    0, 2, 0,
    -1, 0, 3,
]).reshape(3, 3)
S = M + N
S
```

```
In [ ]: D = M - N
D
```

```
In [ ]: a = np.array([1, 3, 5]).reshape(3, 1)
b = np.array([2, 4, 6]).reshape(3, 1)
print(a.T @ b)
print(np.dot(a.T, b))
print(a.T.dot(b))
print(a @ b.T)
print(M @ a)
print(M @ N)
print(a * b)
```

- Loading data from a file

```
In [ ]: d = np.loadtxt(
    fpath,
    dtype={
        'names': ('year', 'temperature'),
        'formats': ('i4', 'f4')
    },
)
d
d['year']
```

## Scientific computing with SciPy

```
In [ ]: import scipy as sp
from scipy import linalg      # Linear algebra
```

```

In [ ]: # Matrix inverse
# A = np.array([
#     [1, 5, 13],
#     [2, 7, 17],
#     [3, 11, 19],
# ])
# b = np.array([1, 2, 3]).reshape(3, 1)
# print(A)

# B = linalg.inv(A)
# print(B)
# print(A @ B)
# print(B @ A)

# c, error, rank, s = linalg.lstsq(A, b)
# print(c, error)

# B = np.array([
#     [1, 3, 4],
#     [2, 3, 2],
#     [0, 0, 4],
# ])
# D = B @ linalg.inv(A)
# print(D)

# Eigenvalues and eigenvectors
M = np.array([
    [1, 2, 0],
    [2, 2, 0],
    [0, 0, 4],
])
print(M)
LAMBDA, V = linalg.eig(M)
print(LAMBDA, '\n', V)
print(V.T @ V)

```

- SVD (Singular Value Decomposition)

$$M = USV^H$$

where  $UU^H = U^H U = I$ ,  $VV^H = V^H V = I$ , and  $S$  is rectangular diagonal.

```
In [ ]: M = np.array([
    [1, 0, 0, 0, 2],
    [0, 0, 3, 0, 0],
    [0, 0, 0, 0, 0],
    [0, 2, 0, 0, 0],
])
U, s, Vh = linalg.svd(M)
print(U)
print(s)
print(Vh)
```

## Visualization with Matplotlib

```
In [ ]: import matplotlib.pyplot as plt

plt.rcParams.update({
    'figure.dpi': 300,
})
```

```
In [ ]: fig, ax = plt.subplots()
ax.plot(
    d['year'], d['temperature'], 'r-', linewidth=2,
    marker='o', markerfacecolor='none', markeredgcolor='k',
)
ax.set_xlim(1965, 2020)
ax.set_ylim(-0.5, 1)
ax.set_xlabel(r'Calendar year, $t_i$ (years)')
ax.set_ylabel(r'Temperature anomaly, $T_i$ ($^\circ$C)')
ax.set_title('Global temperature data', fontsize=14)
# fig.savefig('test.pdf')
```

## Specific packages

### Analysis of tabular data with Pandas

```
In [ ]: import pandas as pd
```

- Load data

```
In [ ]: df = pd.read_csv(
    fpath,
    sep='\s+',
    header=None,
    names=['year', 'temperature'],
    # parse_dates=['year'],
)
# df
```

- Plot data

```
In [ ]: ax = df.plot('year', 'temperature', kind='line', c='r')
df.plot.scatter('year', 'temperature', ax=ax, c='k')
ax.set_ylim(-.5, 1)
```

- Data analysis

```
In [ ]: df = pd.DataFrame(
    {
        "A": ["foo", "bar", "foo", "bar", "foo", "bar", "foo", "foo"],
        "B": ["one", "one", "two", "three", "two", "two", "one", "three"],
        "C": np.random.randn(8),
        "D": np.random.randn(8),
    }
)
df
df.groupby('A').sum()
```

## Machine learning with scikit-learn

### Lasso ( $l_1$ ) regression

$$\arg \min_m ||d - Gm||_2^2 + \alpha ||m||_1$$

```
In [ ]: from sklearn.linear_model import Lasso
```

```

In [ ]: np.random.seed(42)
n_samples, n_features = 50, 100
X = np.random.randn(n_samples, n_features)

idx = np.arange(n_features)
coef = -1**idx * np.exp(-idx/10)
coef[10:] = 0
y = np.dot(X, coef)

y += 0.01 * np.random.normal(size=n_samples)

X_train, y_train = X[:n_samples//2], y[:n_samples//2]
X_test, y_test = X[n_samples//2:], y[n_samples//2:]

alpha = 0.1
lasso = Lasso(alpha=alpha)
model = lasso.fit(X_train, y_train)
y_pred = model.predict(X_test)

fig, axes = plt.subplots(1, 2, figsize=(10, 4))
ax = axes[0]
ax.stem(
    np.where(coef)[0], coef[coef != 0], 'b',
    linefmt='--',
    label='True coefficients',
    markerfmt='bx',
)
ax.stem(
    np.where(lasso.coef_)[0], lasso.coef_[lasso.coef_ != 0],
    'k', markerfmt='kx', label='Lasso coefficients',
)
ax.legend(loc='best')

ax = axes[1]
ax.scatter(y_train, model.predict(X_train), label='Train', alpha=.5)
ax.scatter(y_test, y_pred, label='Test', alpha=.5)
ax.plot([-6, 6], [-6, 6], c='k', zorder=-1)
ax.set_xlabel('True')
ax.set_ylabel('Prediction')
ax.set_aspect(1)
ax.set_xlim(-6, 6)
ax.set_ylim(ax.get_xlim())
ax.legend(loc='best')

```

## Cartography with cartopy

```

In [ ]: import cartopy.crs as ccrs
import cartopy.feature as cfeature

```



```
In [ ]: fig, ax = plt.subplots(subplot_kw={'projection': ccrs.Mercator()})

ax.stock_img()

ax.gridlines(draw_labels=True, alpha=.1)
ax.add_feature(cfeature.COASTLINE, lw=1)
ax.add_feature(cfeature.BORDERS, lw=1)
ax.add_feature(cfeature.STATES, lw=1)
ax.set_extent([245, 260, 35, 43])

# x, y = np.meshgrid(
#     np.linspace(252, 256),
#     np.linspace(37, 41)
# )
# z = x + y
# im = ax.pcolormesh(
#     x, y, z,
#     transform=ccrs.PlateCarree(),    # required for plotting on maps
# )
# cb = fig.colorbar(im, orientation='horizontal')
# cb.set_label('Z (a.u.)')
```