

Deep Learning - lab 2

An introduction to Numpy, Matplotlib, Scipy and TensorFlow

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Installing libraries

Third-party libraries can be imported easily in your code:



Numerical Python
<https://numpy.org/>

Visualization with Python
<https://matplotlib.org/>

Example:

```
pip install numpy matplotlib pandas scipy tensorflow  
# Or  
conda install numpy matplotlib pandas scipy tensorflow
```

Numpy basics

Arrays

Let's suppose we have the matrix (32bit floats):

$$a = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ -1 & 0 & 3 \end{pmatrix}$$

We allocate this matrix with:

```
import numpy as np
a = np.array([[1, 0, 0], [0, 1, 2], [-1, 0, 3]], dtype=np.float32)

type(a) # <class 'numpy.ndarray'>
a.shape # returns (3,3)
a.ndim  # 2
a.dtype # dtype('float32')
a.flatten() # [1, 0, 0, 0, 1, 2, -1, 0, 3]
```

Numpy provides several dtypes: int32, int64, complex32, complex64, float32, float64...

Arrays

```
import numpy as np

# building array with zeros:
z = np.zeros(5, dtype=np.int32)
# array([0, 0, 0, 0, 0], dtype=int32)

# building array with ones
o = np.ones((5,1), dtype=np.float64)

# reshaping array
w = o.reshape(1, 5)
# array([[1., 1., 1., 1., 1.]])
```

Math operations

```
import numpy as np
a = np.array([1, 2, 3, 5])
b = np.ones(4)

c = a + b
# array([2., 3., 4., 6.])

d = c ** 2
e = 2 * np.sin(d)
# array([-1.51360499,  0.82423697, -0.57580663, -1.98355771])

e > 0
# array([False,  True, False, False])
```

Linear algebra

```
import numpy as np

A = np.eye(2) #  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ 
B = np.array([[2, 1], [3, 4]])

# elementwise product
A * B #  $\begin{bmatrix} 2 & 0 \\ 0 & 4 \end{bmatrix}$ 

# matrix product
A @ B # or A.dot(B)
#  $\begin{bmatrix} 2 & 1 \\ 3 & 4 \end{bmatrix}$ 

# transpose matrix
B.T
#  $\begin{bmatrix} 2 & 3 \\ 1 & 4 \end{bmatrix}$ 
```

Slicing

```
import numpy as np
```

```
C = np.array([[2, 1], [3, 4]])
```

```
m = 0
```

```
C[m, m]    # 2
```

```
C[:, m]    # [2, 3]
```

```
C[m, :]    # [2, 1]
```

```
C[m, -1]   # 1
```

```
C[-1, m]   # 3
```

```
C.flatten()[0]    # 2
```

```
C.flatten()[-1]   # 4
```

```
C.flatten()[0:2]  # [2, 1]
```


Copy

```
import numpy as np
```

```
a = np.array([[2, 1], [3, 4]])
```

```
# no copy
```

```
b = a # a and b are the same ndarray
```

```
# shallow copy
```

```
c = a.view()
```

```
c[0, 0] = -1 # a's data change
```

```
c = c.reshape(4) # a's shape doesn't change
```

```
# deep copy
```

```
d = a.copy()
```

Input/Output

```
import numpy as np

# store array
a = np.array([[2, 1], [3, 4]])
np.save("myarray.npy", a)    # binary format
np.savetxt("myarray.txt", a) # or text format

# read from file
b = np.load("myarray.npy")
b = np.loadtxt("myarray.txt")
```

Matplotlib basics

Basics

```
import matplotlib.pyplot as plt
import numpy as np

x = np.linspace(0, 2, 100) # 100 points linearly spaced [0,2]

plt.plot(x, x**2, label="quadratic")

plt.title("Quadratic curve")
plt.xlabel("x label")
plt.ylabel("y label")
plt.legend()

plt.savefig("myplot.png")
plt.show() # keeps the plot open if script
```

https://matplotlib.org/cheatsheets/_images/cheatsheets-1.png

Pandas basics

```
import pandas as pd
import numpy as np

s = pd.Series([1, 3, 5, np.nan, 6, 8])
print(s)
# 0    1.0
# 1    3.0
# 2    5.0
# 3    NaN
# 4    6.0
# 5    8.0
# dtype: float64
```

Dataframes

```
import pandas as pd
import numpy as np
df = pd.DataFrame({
    "A": 1.0,
    "B": pd.Series(1, index=list(range(4)), dtype="float32"),
    "C": np.array([3] * 4, dtype="int32"),
    "D": pd.Categorical(["test", "train", "test", "train"]),
    "E": "foo",})
```

```
# A    B    C    D    E
# 0  1.0  1.0  3  test  foo
# 1  1.0  1.0  3  train foo
# 2  1.0  1.0  3  test  foo
# 3  1.0  1.0  3  train foo
```


Viewing table

```
df.head()    # head of the table
df.tail(3)   # tail (last 3 lines) of the table

df.index
# Int64Index([0, 1, 2, 3], dtype='int64')

df.columns
# Index(['A', 'B', 'C', 'D'], dtype='object')

df.to_numpy() # export dataframe to numpy.ndarray
df.describe() # export mean, std, min, metrics for each column
df.mean()     # computes mean
df.dropna()   # drop NaN (not a number)
df.fillna(value=1) # fill entries with specific value
```

Accessing data

```
df["A"] # returns Series (column A)
```

```
df[0:2] # returns the first 2 rows
```

```
# boolean indexing
```

```
mask = df["A"] > 1
```

```
filtered_df = df[mask]
```

Scipy basics

Non-linear least squares

```
from scipy.optimize import curve_fit

def myfunc(x, a, b, c):
    return a * np.exp(-b * x) + c

xdata = np.load("x.dat")
ydata = np.load("y.dat")

popt, pcov = curve_fit(myfunc, xdata, ydata)
# popopt: optimal values for the parameters
# pcov: estimated covariance of popopt
```

Hessian approach

```
from scipy.optimize import minimize

xdata = np.load("x.dat")
ydata = np.load("y.dat")

def myfunc(x, a, b, c):
    return a * np.exp(-b * x) + c

def myloss(x):
    return np.sum(np.square(myfunc(xdata, x[0], x[1], x[2]) - ydata))

res = minimize(myloss, x0, method="BFGS", options={"disp": True})
# res.x: the solution of the optimization
```

TensorFlow basics

TensorFlow is a library for high performance numerical computation.

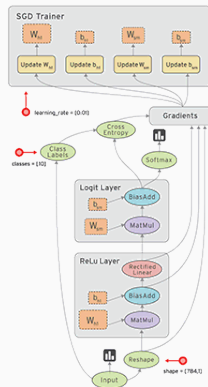
TensorFlow is a library for high performance numerical computation.

Pros:

- solves optimization problems with automatic differentiation.
- can be extended in python and c/c++.
- runs seamlessly on CPU and GPU, and can uses JIT technology.

Cons:

- do not provides builtin models from the core framework
- less automation for cross-validation and hyperparameter tune



Tensors

```
import tensorflow as tf

x = tf.constant([1, 2, 3]) # immutable object

x.shape
x.dtype
x ** 2
tf.reduce_sum(x)

# similarly to numpy
a = tf.zeros(5, dtype=tf.float32)
b = tf.ones(5, dtype=tf.float32)

# convert to numpy
c = b.numpy()
```

Variables

```
import tensorflow as tf

var = tf.Variable([0, 0, 0]) # mutable object
var.assign([1, 1, 1])
```

Automatic differentiation

```
import tensorflow as tf

var = tf.Variable(1.0) # mutable object

def myfunc(x):
    return x ** 2 + 2 * x - 5

f(x) # <tf.Tensor: shape=(), dtype=float32, numpy=-2.0>

with tf.GradientTape() as tape:    # gradient calculation
    y = f(x)

g_x = tape.gradient(y, x) #  $g(x) = dy/dx = (2 * x + 2)$ 
g_x # <tf.Tensor: shape=(), dtype=float32, numpy=4.0>
```

Graphs - performance optimization

```
import tensorflow as tf
```

```
@tf.function
```

```
def myfunc(x):
```

```
    print("Tracing")
```

```
    return tf.reduce_sum(x)
```

```
myfunc(tf.constant([1, 2, 3]))
```

```
# Tracing
```

```
# <tf.Tensor: shape=(), dtype=int32, numpy=6>
```

```
myfunc(tf.constant([-1, 2, -1]))
```

```
# <tf.Tensor: shape=(), dtype=int32, numpy=0>
```

```
myfunc(tf.constant([10.0, 9.1, 8.2], dtype=tf.float32))
```

```
# Tracing -> <tf.Tensor: shape=(), dtype=float32, numpy=27.3>
```