



# Data Structures

Aula 2

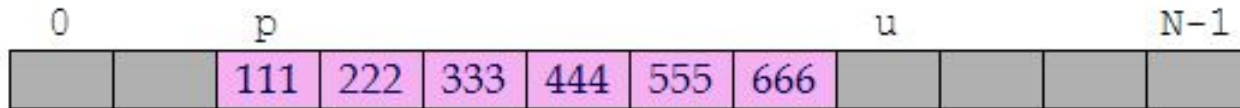
# Queues

Dynamic structure that uses the strategy “First-In-First-Out” (FIFO)

The first inserted object in queue is also the first removed

Insertion and remotion  $O(1)$

Some applications: Transport simulation, message, buffer ...





# Queues (in python)

```
from collections import deque
```

```
fila = deque(["Eric", "John", "Michael"])
```

```
fila.append("Terry")
```

```
fila.append("Graham")
```

```
fila.popleft()
```

```
fila.popleft()
```

```
fila
```

# Stacks

Last-In-First-Out (LIFO)

The first inserted item is the last to be removed

Insertion and removal  $O(1)$

Applications: Expression evaluation and syntax parsing, recursivity ...





# Stacks (in python)

```
pilha = [3, 4, 5]
```

```
pilha.append(6)
```

```
pilha.append(7)
```

```
pilha.pop()
```

```
print(pilha)
```

```
pilha.pop()
```

```
pilha.pop()
```

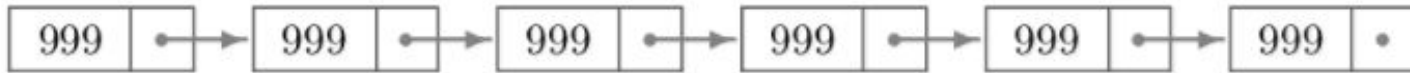
```
pilha
```

# Lists

It's a sequence of cells: Each cell contains an object of some type and the address to the next cell

Can be used to store a list of elements and form the basis for other abstract data types including the queue, the stack, and their variations

Applications: Storage (hard drive, file systems, memory ...), trees ...





# Lists (in python)

```
a = [66.25, 333, 333, 1, 1234.5]
```

```
a.insert(2, -1)
```

```
a.append(333)
```

```
print(a)
```

```
a.index(333)
```

```
a.remove(333)
```

```
print(a)
```

```
a.reverse()
```

```
print(a)
```

```
a.sort()
```

```
a
```



# Filter

```
def f(x):
```

```
    return x % 2 != 0 and x % 3 != 0
```

```
res = filter(f, range(2, 25))
```

```
for i in res:
```

```
    print(i)
```





# Map

```
def f(x):
```

```
    return x % 2 != 0 and x % 3 != 0
```

```
res = map(f, range(2, 25))
```

```
for i in res:
```

```
    print(i)
```



# Reduce

```
from functools import reduce
```

```
def f(x, y):
```

```
    return x + y
```

```
print(reduce(f, range(1, 5)))
```

# Set

Do not admit repetitions

It is not ordered

Allows mathematical expressions like union, intersection, difference ...

Applications: Verifies if exists an element, eliminate duplicated elements:

```
cesta = ['uva', 'laranja', 'uva', 'abacaxi', 'laranja', 'banana']
```

```
frutas = set(cesta)
```

```
print(frutas)
```

```
'capim' in frutas
```



# Set (math operations)

```
a = set('abracadabra')
```

```
b = set('alacazam')
```

```
print(a)
```

```
print(a - b)
```

```
print(a | b)
```

```
print(a & b)
```

```
print(a ^ b)
```



# Iterating through lists

```
knight = {'gallahad': 'the pure', 'robin': 'the brave'}
```

```
for k, v in knight.items():
```

```
    print(k, v)
```

```
# two or more lists
```

```
questions = ['name', 'quest', 'favorite color']
```

```
answers = ['lancelot', 'the holy grail', 'blue']
```

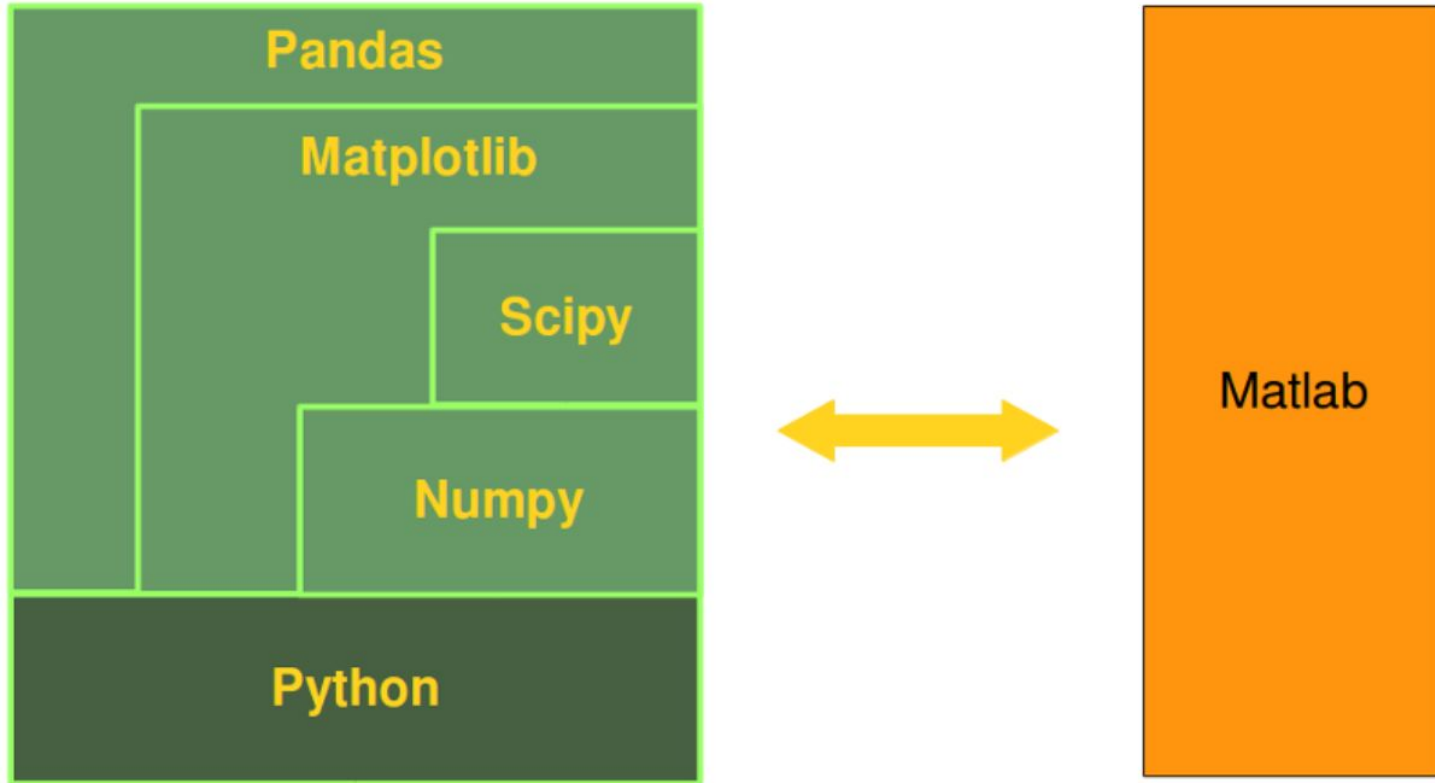
```
for q, a in zip(questions, answers):
```

```
    print('What is your {0}? It is {1}'.format(q, a))
```



# Numpy

# Matlab alternative





# Comparison between Core Python and Numpy

The advantages of Core Python:

- high-level number objects: integers, floating point
- containers: lists with cheap insertion and append methods, dictionaries with fast lookup

Advantages of using Numpy with Python:

- array oriented computing
- efficiently implemented multi-dimensional arrays
- designed for scientific computation





# Numpy array

The most important attributes of a ndarray object are:

- `Ndarray.ndim`
- `Ndarray.shape`
- `Ndarray.size`
- `Ndarray.dtype`
- `ndarray.itemsize`



# Numpy array (some useful functions)

```
import numpy as np
```

```
a = np.arange(15)
```

```
print(a)
```

```
print(a.reshape(3, 5))
```

```
print(a.shape)
```

Test the previous commands !



# Exercises

# Print the same result of `np.arange(15)` without use numpy

# Print the same result of `a.reshape(3, 5)` without use numpy



# Numpy array (some useful functions)

```
Z = np.zeros( (3,4) )
```

```
print(Z)
```

```
O = np.ones( (2,3,4) )
```

```
print(O)
```

```
print(np.empty( (2,3) ) )
```



# Exercises

# print the same result of `np.zeros( (3,4) )` without use numpy

# print the same result of `np.ones( (2,3,4) )` without use numpy

# print the same result of `np.empty( (2,3) )` without use numpy



# Numpy array (some useful functions)

```
L = np.linspace( 0, 2, 9 )           # 9 numbers from 0 to 2
```

```
print(L)
```

```
x = np.linspace( 0, 2*pi, 100 )     # useful to evaluate function at lots of points
```

```
print(x)
```



# Exercises

# print the same result of `np.linspace( 0, 2, 9 )` without use numpy

# print the same result of `np.linspace( 0, 2*pi, 100 )` without use numpy



# Numpy operations

```
A = np.array( [[1,1], [0,1]] )
```

```
B = np.array( [[2,0], [3,4]] )
```

```
print(A*B)                # elementwise product
```

```
print(A.dot(B))           # matrix product
```

```
print(np.dot(A, B))
```





# Exercises

```
A = np.array( [[1,1], [0,1]] )
```

```
B = np.array( [[2,0], [3,4]] )
```

```
# Do elementwise product "A*B" without numpy
```

```
# Do matrix product "A.dot(B)" without numpy
```



# Numpy shape manipulation

```
import numpy as np
```

```
a = np.random.random((2,3))
```

```
print(a)
```

```
print(a.ravel())
```

```
print(a.reshape(3,2))
```

```
print(a)
```

```
a.resize((3,2))
```

```
print(a)
```



# Exercises

# Do the same list modification performed by `np.resize((3,2))` without numpy

# Do the `numpy.ravel()` method without numpy



# Matplotlib



# Plotting

```
import matplotlib.pyplot as plt
```

```
C = np.array([20.1, 20.8, 21.9, 22.5, 22.7, 22.3, 21.8, 21.2, 20.9, 20.1])
```

```
plt.plot(C)
```

```
plt.show()
```



# Plotting (labels)

```
import matplotlib.pyplot as plt
```

```
plt.plot([1,2,3,4])
```

```
plt.ylabel('some numbers')
```

```
plt.show()
```

```
``plt.xlabel('label of x')
```

```
plt.title('title of figure')
```

```
...
```



# Plotting (axis)

```
import matplotlib.pyplot as plt
```

```
plt.plot([1,2,3,4], [1,4,9,16], 'ro')
```

```
plt.axis([0, 6, 0, 20])
```

```
plt.show()
```



# Plotting (line styles)

```
import matplotlib.pyplot as plt
```

```
# evenly sampled time at 200ms intervals
```

```
t = np.arange(0., 5., 0.2)
```

```
# red dashes, blue squares and green triangles
```

```
plt.plot(t, t, 'r--', t, t**2, 'bs', t, t**3, 'g^')
```

```
plt.show()
```



# Plotting (multiple figures and axes)

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
def f(t):
```

```
    return np.exp(-t) * np.cos(2*np.pi*t)
```

```
t1 = np.arange(0.0, 5.0, 0.1)
```

```
t2 = np.arange(0.0, 5.0, 0.02)
```

```
plt.figure(1)
```

```
plt.subplot(211)
```

```
plt.plot(t1, f(t1), 'bo', t2, f(t2), 'k')
```

```
plt.subplot(212)
```

```
plt.plot(t2, np.cos(2*np.pi*t2), 'r--')
```

```
plt.show()
```



# Plotting (annotate)

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
ax = plt.subplot(111)
```

```
t = np.arange(0.0, 5.0, 0.01)
```

```
s = np.cos(2*np.pi*t)
```

```
line, = plt.plot(t, s, lw=2)
```

```
plt.annotate('local max', xy=(2, 1),  
            xytext=(3, 1.5),  
            arrowprops=dict(facecolor='black',  
                             shrink=0.05),)
```

```
plt.ylim(-2,2)
```

```
plt.show()
```



# Histograms

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
mu, sigma = 2, 0.5
```

```
v = np.random.normal(mu,sigma,10000)
```

```
plt.hist(v, bins=50, normed=1)    # matplotlib version (plot)
```

```
plt.show()
```



# Histograms (numpy)

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
(n, bins) = np.histogram(v, bins=50, normed=True) # NumPy version (no plot)
```

```
plt.plot(.5*(bins[1:]+bins[:-1]), n)
```

```
plt.show()
```