## Lab 01

### Lab 01

- Numpy
- Image manipulation

# Using numpy

import numpy as np

### Arrays

Arrays represent matrices in numpy

```
a = np.array([1, 2, 3]) # Create a rank 1 array
print(type(a), a.shape, a[0], a[1], a[2])
a[0] = 5 # Change an element of the array
print(a)

b = np.array([[1,2,3],[4,5,6]]) # Create a rank 2 array
print(type(b), b.shape)
print(b)
print(b)
print(b[0, 0], b[0, 1], b[1, 0])
```

```
(class 'numpy.ndarray') (3,) 1 2 3
[5 2 3]
(class 'numpy.ndarray') (2, 3)
[[1 2 3]
[4 5 6]]
1 2 4
```

# initialized arrays

## initialized arrays

```
c = np.full((2,2), 7) # Create a constant array
        print(c)
[777]
 [7 7]]
        d = np.eye(2) # Create a 2x2 identity matrix
        print(d)
[[ 1. 0.]
[ 0. 1.]]
        e = np.random.random((2,2)) # Create an array filled with random values
        print(e)
[[ 0.72534879  0.18028294]
 [ 0.72548017  0.2593257 ]]
```

# Array indexing: slicing

Equivalent to list slicing

```
import numpy as np
# Create the following rank 2 array with shape (3, 4)
# [[ 1 2 3 4]
# [5 6 7 8]
# [ 9 10 11 12]]
a = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
# Use slicing to pull out the subarray consisting of the first 2 rows
# and columns 1 and 2; b is the following array of shape (2, 2):
# [[2 3]
# [6 7]]
b = a[:2, 1:3]
print(b)
```

# Array indexing: slicing

Similar to list slicing

```
import numpy as np
# Create the following rank 2 array with shape (3, 4)
# [[ 1 2 3 4]
# [5 6 7 8]
# [ 9 10 11 12]]
a = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
# Use slicing to pull out the subarray consisting of the first 2 rows
# and columns 1 and 2; b is the following array of shape (2, 2):
# [[2 3]
# [6 7]]
b = a[:2, 1:3]
print(b)
```

# Array indexing: slicing

• Slice = view

```
# [[ 1 2 3 4]

# [ 5 6 7 8]

# [ 9 10 11 12]]

print(a[0, 1])

b[0, 0] = 77 # b[0, 0] is the same piece of data as a[0, 1]

print(a[0, 1])
```



### Array indexing: integer indexing

```
>>> a
array([[1, 2],
a = np.array([[1,2], [3, 4], [5, 6]])

# An example of integer array indexing.
# The returned array will have shape (3,) and
print(a[[0, 1, 2], [0, 1, 0]])

# The above example of integer array indexing is equivalent to this:
print(np.array([a[0, 0], a[1, 1], a[2, 0]]))
```

```
[1 4 5]
[1 4 5]
```

### Array indexing: integer indexing

Integer indexing creates a new array

```
>>> a = np.array([[1,2], [3, 4], [5, 6]])
>>> b = a[[0,0],[1,1]]
>>> a
array([[1, 2],
       [3, 4],
       [5, 6]])
>>> b
array([2, 2])
>>> b[0] = 55
>>> b
array([55, 2])
>>> a
array([[1, 2],
       [3, 4],
       [5, 6]])
```

### Array indexing: mixed indexing

```
row_r1 = a[1, :]  # Rank 1 view of the second row of a
row_r2 = a[1:2, :]  # Rank 2 view of the second row of a
row_r3 = a[[1], :]  # Rank 2 view of the second row of a
print(row_r1, row_r1.shape)  # [[ 1  2  3  4]
print(row_r2, row_r2.shape)  # [ 5  6  7  8]
print(row_r3, row_r3.shape)  # [ 9 10 11 12]]
```



```
[5 6 7 8] (4,)
[[5 6 7 8]] (1, 4)
[[5 6 7 8]] (1, 4)
```

### Array indexing: mixed indexing

Mixed indexing = view

## Mutating elements

```
# Create a new array from which we will select elements
a = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
print(a)
[[1 2 3]
 [4 5 6]
 [7 8 9]
 [10 11 12]]
# Create an array of indices
b = np.array([0, 2, 0, 1])
# Select one element from each row of a using the indices in b
print(a[np.arange(4), b]) # Prints "[ 1 6 7 11]"
[1 6 7 11]
```

# Mutating elements

```
>>> a = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
>>> print(a)
[[ 1  2  3]
  [ 4  5  6]
  [ 7  8  9]
  [10 11 12]]
>>> bb = list(range(4))
>>> bb
[0, 1, 2, 3]
>>> b = [0,2,0,1]
>>> a[bb,b]
array([ 1, 6, 7, 11])
```

# Mutating elements

```
# Mutate one element from each row of a using the indices in b
a[np.arange(4), b] += 10
print(a)

[[11  2  3]
  [ 4  5  16]
  [17  8  9]
  [10  21  12]]
```

## Boolean indexing

# Boolean indexing

```
# We use boolean array indexing to construct a rank 1 array
# consisting of the elements of a corresponding to the True values
# of bool_idx
print(a[bool_idx])

# We can do all of the above in a single concise statement:
print(a[a > 2])

[3 4 5 6]
[3 4 5 6]
```

## Numpy datatypes

```
x = np.array([1, 2]) # Let numpy choose the datatype
y = np.array([1.0, 2.0]) # Let numpy choose the datatype
z = np.array([1, 2], dtype=np.int64) # Force a particular datatype
print(x.dtype, y.dtype, z.dtype)
int64 float64 int64
```

#### Math: add

```
x = np.array([[1,2],[3,4]], dtype=np.float64)
y = np.array([[5,6],[7,8]], dtype=np.float64)

# Elementwise sum; both produce the array
print(x + y)
print(np.add(x, y))

[[ 6.  8.]
[ 10.  12.]]
[[ 6.  8.]
[ 10.  12.]]
```

#### Math: subtract

```
# Elementwise difference; both produce the array
print(x - y)
print(np.subtract(x, y))

[[-4. -4.]
[-4. -4.]]
[[-4. -4.]]
```

## Math: multiply

Note that \* is elem-wise multiplication

```
# Elementwise product; both produce the array
print(x * y)
print(np.multiply(x, y))

[[ 5. 12.]
  [ 21. 32.]]
[[ 5. 12.]
  [ 21. 32.]]
```

#### Math: divide

# Math: inner product

```
v = np.array([9,10])
w = np.array([11, 12])

# Inner product of vectors; both produce 219
print(v.dot(w))
print(np.dot(v, w))
```

### Math: matrix multiplication

```
x = np.array([[1,2],[3,4]])
          y = np.array([[5,6],[7,8]])
          v = np.array([9,10])
          w = np.array([11, 12])
# Matrix / vector product; both produce the rank 1 array [29 67]
print(x.dot(v))
print(np.dot(x, v))
# Matrix / matrix product; both produce the rank 2 array
# [[19 22]
# [43 50]]
print(x.dot(y))
print(np.dot(x, y))
```

#### Useful functions

```
x = np.array([[1,2],[3,4]])
print(np.sum(x)) # Compute sum of all elements; prints "10"
print(np.sum(x, axis=0)) # Compute sum of each column; prints "[4 6]"
print(np.sum(x, axis=1)) # Compute sum of each row; prints "[3 7]"
10
[4 6]
[3 7]
                v = np.array([[1,2,3]])
print(x)
                print(v)
print(x.T)
                print(v.T)
[[1 2]
                [[1 2 3]]
 [3 4]]
                \lceil \lceil 1 \rceil
[[1 3]
                 [2]
 [2 4]]
                 [3]]
```

### Broadcasting

Sometimes we work with matrices in different shapes

```
# We will add the vector v to each row of the matrix x,
# storing the result in the matrix y
x = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
v = np.array([1, 0, 1])
y = np.empty_like(x)  # Create an empty matrix with the same shape as x

# Add the vector v to each row of the matrix x with an explicit loop
for i in range(4):
    y[i, :] = x[i, :] + v
print(y)
```

### Broadcasting

One way to do the same

### Broadcasting

```
# We will add the vector v to each row of the matrix x,
# storing the result in the matrix y
x = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
v = np.array([1, 0, 1])
y = x + v # Add v to each row of x using broadcasting
print(y)
```

# Image manipulation

- scikit-image
  - from skimage import io
  - from skimage import color

### Load image

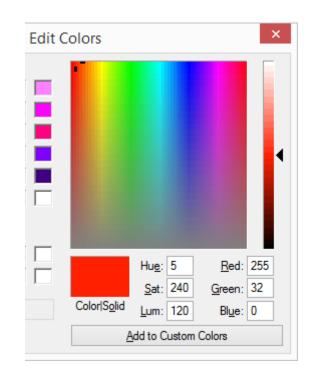
• io.imread returns numpy array of shape (height, width, 3)

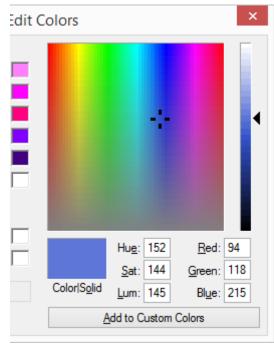
```
>>> from skimage import io
>>> asdf = io.imread('image1.jpg')
>>> asdf.shape
<300, 300, 3>
```

### Image representation

- numpy array of (height, width, channels)
  - -(300,300,3)
  - 3 channels are for R,G,B







# Save image

io.imsave saves numpy array into a file

```
io.imsave('image11.jpg',asdf)
```

#### homework

- Write a function to do the followings:
  - Load an image
  - Suppress red channel in the image
  - Save the image

### specification

- Write hw\_[student\_id].py containing a function hw1(fname\_input, fname\_output)
- zip to submit

# Pre-check your score

Use grader.py