

Numpy Arrays

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NumPy Fundamentals

- Numpy is a module for faster vector processing with numerous other routines
- Scipy is a more extensive module that also includes many other functionalities such as machine learning and statistics

NumPy Fundamentals

- Why Numpy?
 - Remember that Python does not limit lists to just elements of a single class
 - If we have a large list $[a_1, a_2, a_3, \dots, a_n]$ and we want to add a number to all of the elements, then Python will ask for each element:
 - What is the type of the element
 - Does the type support the + operation
 - Look up the code for the + and execute
 - This is slow

NumPy Fundamentals

- Why Numpy?
 - Primary feature of Numpy are arrays:
 - List like structure where all the elements have the same type
 - Usually a floating point type
 - Can calculate with arrays much faster than with list
 - Implemented in C / Java for Cython or Jython

NumPy Arrays

- NumPy Arrays are containers for numerical values
- Numpy arrays have dimensions
 - Vectors: one-dimensional
 - Matrices: two-dimensional
 - Tensors: more dimensions, but much more rarely used
- Nota bene: A matrix can have a single row and a single column, but has still two dimensions

NumPy Arrays

- After installing, try out `import numpy as np`
- Making arrays:
 - Can use lists, though they better be of the same type

```
import numpy as np
my_list = [1, 5, 4, 2]
my_vec = np.array(my_list)
my_list = [[1, 2], [4, 3]]
my_mat = np.array(my_list)
```

Array Creation

- Numpy can generate arrays:
 - From disks or the net,
 - using various libraries
 - using loadtxt and similar functions
 - From lists and similar data structures
 - Generate them natively

Array Creation

- Numpy has a number of ways to create an array
 - Import numpy as np
 - `np.zeros((2,3))`
 - `array([[0., 0., 0.], [0., 0., 0.]])`
 - `np.ones(5)`
 - `array([1., 1., 1., 1., 1.])`
 - `np.eye(3)` generates the identity matrix
 - `array([[1., 0., 0.], [0., 1., 0.], [0., 0., 1.]])`

Array Creation

- Numpy has a number of ways to create arrays
 - `np.linspace(1., 4., 6)` creates an array of 6 elements between 1.0 and 4.0 evenly spaced out
 - `array([1. , 1.6, 2.2, 2.8, 3.4, 4.])`
 - `np.arange(2, 3, 0.1)` a more generalized version of Python's range function (with float step)
 - `array([2. , 2.1, 2.2, 2.3, 2.4, 2.5, 2.6, 2.7, 2.8, 2.9])`
 - `np.arange(2, 5)`
 - `array([2, 3, 4])`

Array Creation

- Can generate using lists, tuples, etc. even with a mix of types
 - `np.array([[1, 2, 3], (1, 0, 0.5)])`
 - `array([[1. , 2. , 3.], [1. , 0. , 0.5]])`

Array Creation

- Creating arrays:
 - `np.full` to fill in with a given value

```
np.full(5, 3.141)
```

```
array([3.141, 3.141, 3.141, 3.141, 3.141])
```

Array Creation

- Can also use random values.
 - Uniform distribution between 0 and 1

```
>>> np.random.random( (3, 2) )
array([[0.39211415, 0.50264835],
       [0.95824337, 0.58949256],
       [0.59318281, 0.05752833]])
```

Array Creation

- Or random integers

```
>>> np.random.randint(0,20,(2,4))
```

```
array([[ 5,  7,  2, 10],  
       [19,  7,  1, 10]])
```

Array Creation

- Or other distributions, e.g. normal distribution with mean 2 and standard deviation 0.5

```
>>> np.random.normal(2, 0.5, (2, 3))  
array([[1.34857621, 1.34419178, 1.977698   ],  
       [1.31054068, 2.35126538, 3.25903903]])
```

Array Creation

- fromfunction

```
>>> x = np.fromfunction(lambda i,j: (i**2+j**2)//2, (4,5) )
```

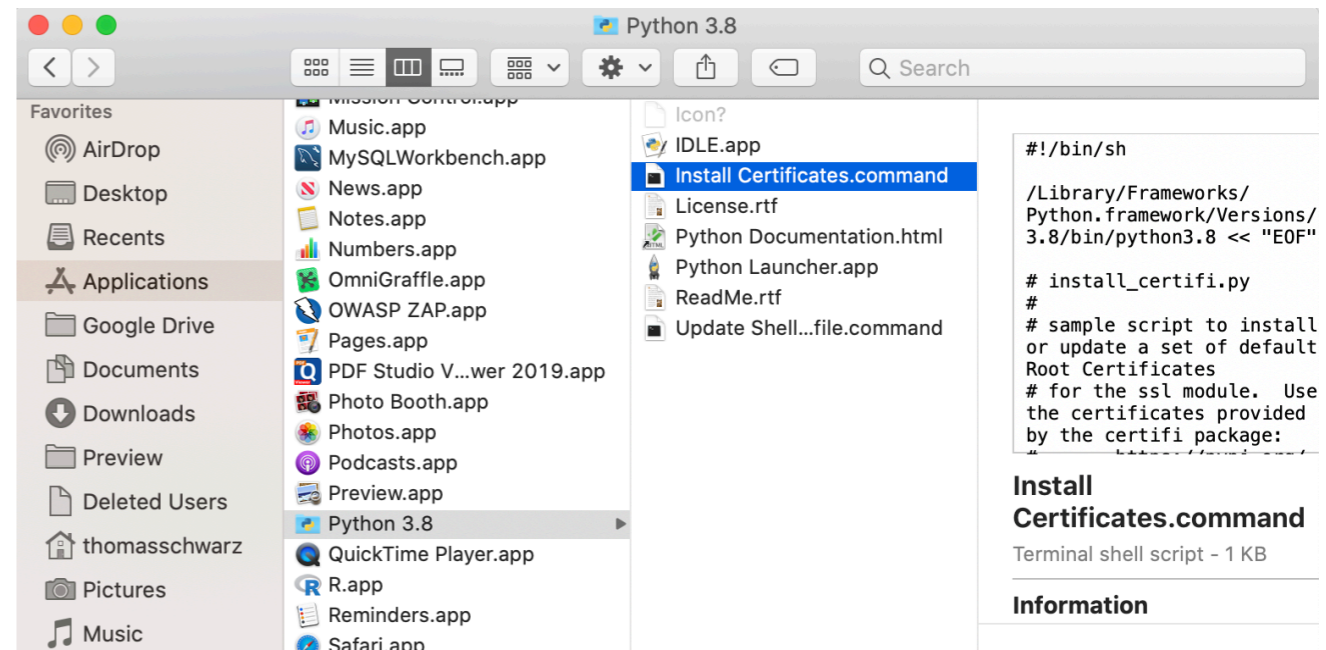
```
>>> x.astype(int)
```

```
array([[ 0,  0,  2,  4,  8],
       [ 0,  1,  2,  5,  8],
       [ 2,  2,  4,  6, 10],
       [ 4,  5,  6,  9, 12]])
```

```
>>> x.shape
(4, 5)
```

Array Creation

- Creating from download / file
 - We use urllib.request module
 - If you are on Mac, you need to have Python certificates installed
 - Go to your Python installation in Applications and click on "Install Certificates command"



Array Creation

- Use `urllib.request.urlretrieve` with website and file name
 - Remember: A file will be created, but the directory needs to exist

```
urllib.request.urlretrieve(  
    url = "https://ndownloader.figshare.com/files/12565616",  
    filename = "avg-monthly-precip.txt"  
)
```

- This is a text file, with one numerical value per line
- Then create the numpy array using

```
avgmp = np.loadtxt(fname = 'avg-monthly-precip.txt')  
print(avgmp)
```

Array Creation

- Example: Get an account at openweathermap.org/appid
- Install requests and import json
 - Use the [openweathermap.org](https://openweathermap.org/api) api to request data on a certain town
 - Result is send as a JSON dictionary

Array Creation

```
import numpy as np
import requests
import json

mumbai=json.loads(requests.get('http://api.openweathermap.org/data/
2.5/weather?
q=mumbai,india&APPID=4561e0cd15ec2ee307bdcfe19ec22ab9').text)
vasai = json.loads(requests.get('http://api.openweathermap.org/data/
2.5/weather?
q=vasai,india&APPID=4561e0cd15ec2ee307bdcfe19ec22ab9').text)
navi_mumbai = json.loads(requests.get('http://api.openweathermap.org
data/2.5/weather?
q=navi%20mumbai,india&APPID=4561e0cd15ec2ee307bdcfe19ec22ab9').text)
```

Array Creation

- Can use `np.genfromtext`
 - Very powerful and complicated function with many different options
 -

Array Creation

- Example

```
converters = {5: lambda x: int(0 if 'Iris-setosa' else 1 if 'Iris-  
virginica' else 2) }  
my_array = np.genfromtxt('../Classes2/Iris.csv',  
                          usecols=(1,2,3,4,5),  
                          dtype=[float, float, float, float, float],  
                          delimiter = ',',  
                          converters = converters,  
                          skip_header=1)
```

- Need a source (the iris file)
- Can specify the columns we need
- Give the dtype U20-unicode string, S20-byte string
- Delimiter
- Skipheader, skipfooter
- converters to deal with encoding

Array Creation

- This is an array of 150 tuples
- Use comprehension to convert to a two-dimensional array

```
m = np.array( [ [row[0], row[1], row[2], row[3], row[4]]  
               for row in my_array ] )
```

NumPy Array Attributes

- The number of dimensions: `ndim`
- The values of the dimensions as a tuple: `shape`
- The size (number of elements)

```
>>> tensor
array([[ [2.11208424, 2.01510638, 2.03126777, 1.89670846],
        [1.94359036, 2.02299445, 2.08515919, 2.05402626],
        [1.8853457 , 2.01236192, 2.07019962, 1.93713157]]],

       [[ [1.84275427, 1.99537922, 1.96060154, 1.90020305],
        [2.00270166, 2.11286224, 2.03144254, 2.06924855],
        [1.95375653, 2.0612986 , 1.82571628, 1.86067971]]]])
>>> tensor.ndim
3
>>> tensor.shape
(2, 3, 4)
>>> tensor.size
24
```

NumPy Array Attributes

- The data type: dtype
 - can be bool, int, int64, uint, uint64, float, float64, complex ...
- The size of a single element in bytes: itemsize
- The size of the total array: nbytes

NumPy Array Indexing

- Single elements
 - Use the bracket notation []
 - Single array: Same as in standard python

```
>>> vector = np.random.normal(10,1,(5))
>>> print(vector)
[10.25056641 11.37079651 10.44719557 10.54447875 10.43634562]
>>> vector[4]
10.436345621654919
>>> vector[-2]
10.544478746079845
```

NumPy Arrays Indexing

- Matrix and tensor elements: Use a single bracket and a comma separated tuple

```
>>> tensor
array([[[[2.11208424, 2.01510638, 2.03126777, 1.89670846],
         [1.94359036, 2.02299445, 2.08515919, 2.05402626],
         [1.8853457 , 2.01236192, 2.07019962, 1.93713157]]],

       [[1.84275427, 1.99537922, 1.96060154, 1.90020305],
        [2.00270166, 2.11286224, 2.03144254, 2.06924855],
        [1.95375653, 2.0612986 , 1.82571628, 1.86067971]]]])
>>> tensor[0,0,1]
2.015106376191313
```

NumPy Arrays Indexing

- Multiple bracket notation
 - We can also use the Python indexing of multi-dimensional lists using several brackets

```
>>> tensor[0][1][2]  
2.085159191502853
```

- It is more writing and more error prone than the single bracket version

NumPy Arrays Indexing

- We can also define slices

```
>>> vector = np.random.normal(10,1,(3))
>>> vector
array([10.61948855,  7.99635252,  9.05538706])
>>> vector[1:3]
array([7.99635252,  9.05538706])
```

NumPy Arrays Indexing

- In Python, slices are new lists
- In NumPy, slices are **not** copies
 - Changing a slice changes the original

NumPy Arrays Indexing

- Example:

- Create an array

```
>>> vector = np.random.normal(10, 1, (3))
```

```
>>> vector
```

```
array([10.61948855,  7.99635252,  9.05538706])
```

- Define a slice

```
>>> x = vector[1:3]
```

NumPy Arrays Indexing

- Example (cont.)
 - Change the first element in the slice

```
>>> x[0] = 5.0
```

- Verify that the change has happened

```
>>> x  
array([5.          , 9.05538706])
```

- But the original has also changed:

```
>>> vector  
array([10.61948855, 5.          , 9.05538706])
```

NumPy Arrays Indexing

- Slicing does **not** makes copies
 - This is done in order to be efficient
 - Numerical calculations with a large amount of data get slowed down by unnecessary copies

NumPy Arrays Indexing

- If we want a copy, we need to make one with the copy method
- Example:

- Make an array

```
>>> vector = np.random.randint(0,10,5)
>>> vector
array([0, 9, 5, 7, 8])
```

- Make a copy of the array

```
>>> my_vector_copy = vector.copy()
```

NumPy Arrays Indexing

- Example (continued)

- Change the middle elements in the copy

```
>>> my_vector_copy[1:-2]=100
```

- Check the change

```
>>> my_vector_copy  
array([  0, 100, 100,   7,   8])
```

- Check the original

```
>>> vector  
array([0, 9, 5, 7, 8])
```

- No change!

NumPy Arrays Indexing

- Multi-dimensional slicing
 - Combines the slicing operation for each dimension

```
>>> slice = tensor[1:, :2, :1]
>>> slice
array([[ [1.84275427],
        [2.00270166]]])
```

NumPy Arrays

Conditional Selection

- We can create an array of Boolean values using comparisons on the array

```
>>> array = np.random.randint(0,10,8)
>>> array
array([2, 4, 4, 0, 0, 4, 8, 4])
>>> bool_array = array > 5
>>> bool_array
array([False, False, False, False, False,
       False,  True, False])
```

NumPy Arrays

Conditional Selection

- We can then use the Boolean array to create a selection from the original array

```
>>> selection=array[bool_array]
>>> selection
array([8])
```

- The new array only has one element!

Selftest

- Can you do this in one step?
 - Create a random array of 10 elements between 0 and 10
 - Then select the ones larger than 5

Selftest Solution

- Solution:
 - Looks a bit cryptic
 - First, we create an array

```
>>> arr = np.random.randint(0,10,10)
>>> arr
array([3, 2, 7, 8, 7, 2, 1, 0, 4, 8])
```

- Then we select in a single step

```
>>> sel = arr[arr>5]
>>> sel
array([7, 8, 7, 8])
```

NumPy Arrays

Conditional Selection

- Let's try this out with a matrix
 - We create a vector, then use **reshape** to make the array into a vector
 - Recall: the number of elements needs to be the same

```
>>> mat = np.arange(1,13).reshape(3,4)
>>> mat
array([[ 1,  2,  3,  4],
       [ 5,  6,  7,  8],
       [ 9, 10, 11, 12]])
```


NumPy Arrays

Conditional Selection

- Now let's select:

```
>>> mat1 = mat[mat>6]
>>> mat1
array([ 7,  8,  9, 10, 11, 12])
```

- This is no longer a matrix, which makes sense

Slicing

- Photo Manipulation
 - Need to install imageio and matplotlib

```
import imageio
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
```

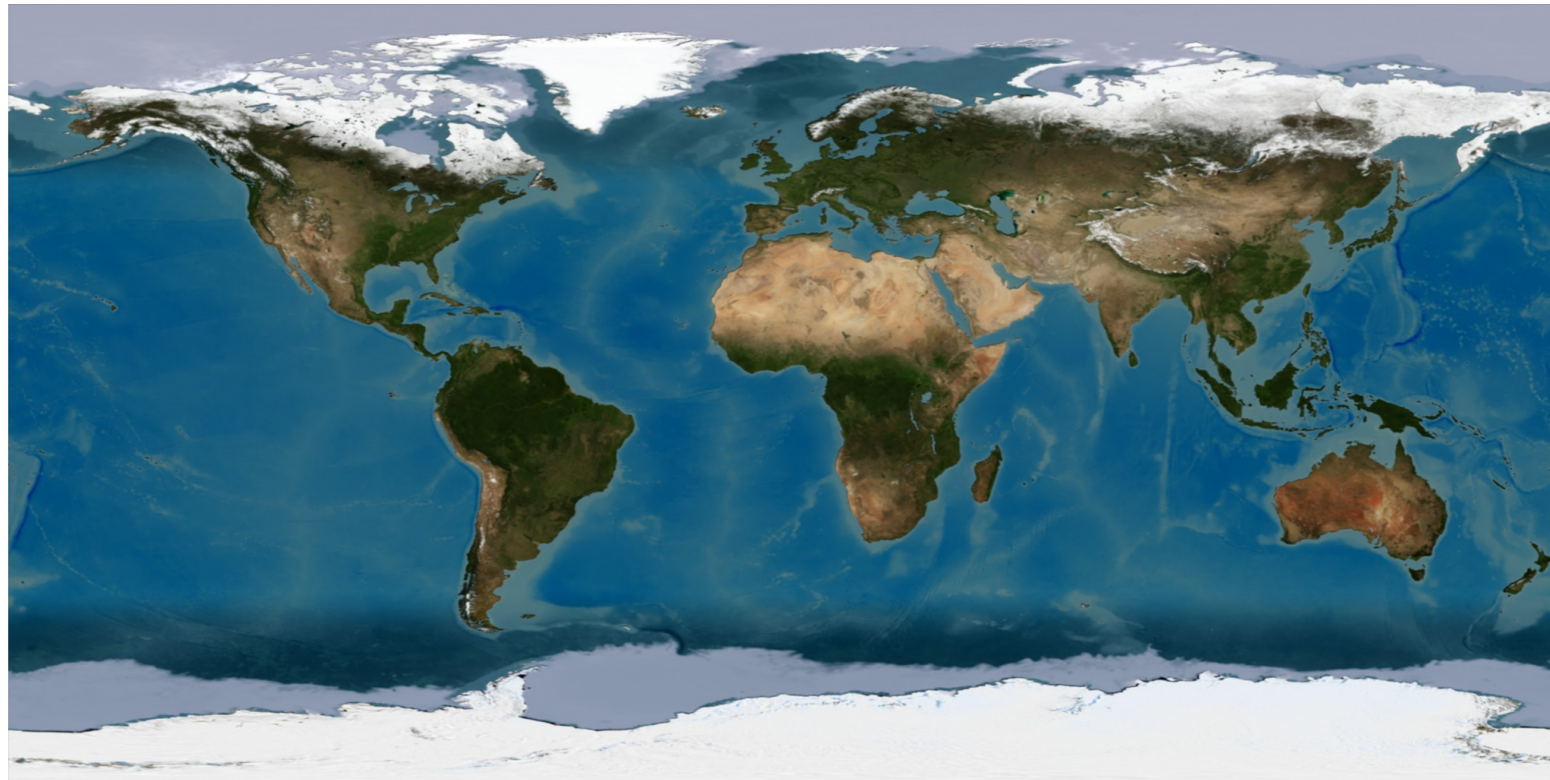
- Get a photo as a 3-dimensional array

-

```
im = mpimg.imread('earth.jpg')
print(im.shape)
```

Slicing

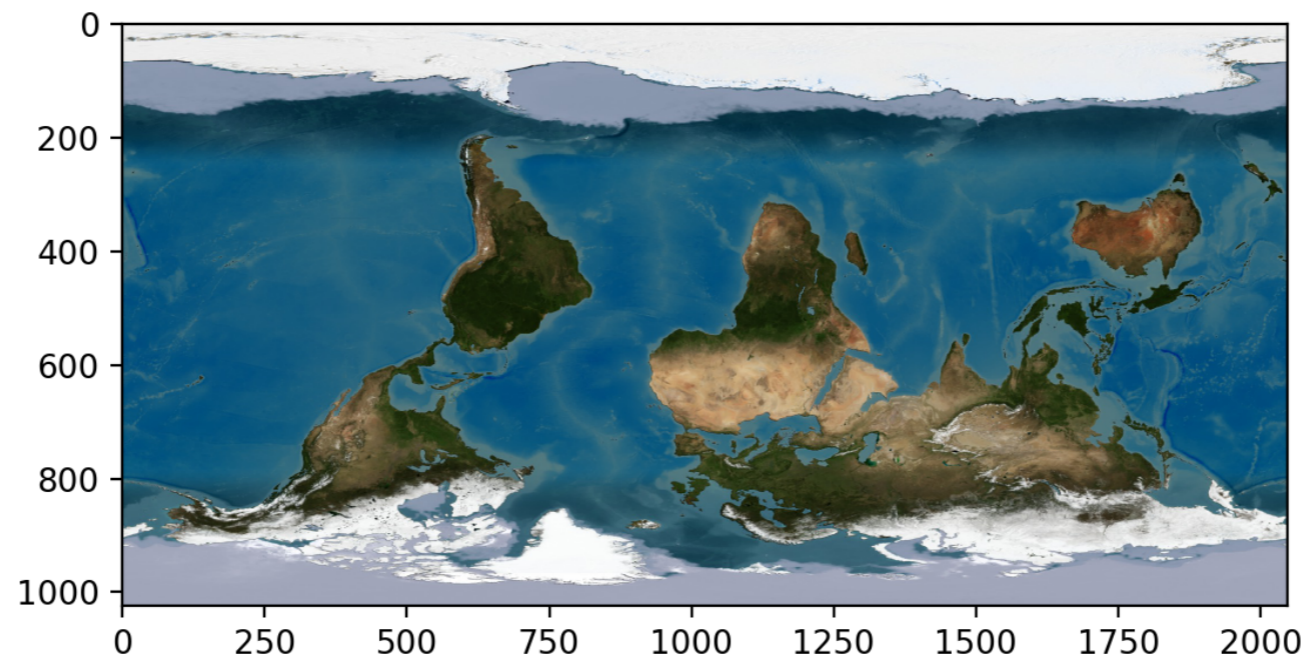
- Display the photo
- ```
plt.imshow(im)
plt.show()
```



# Slicing

- Swap first coordinate

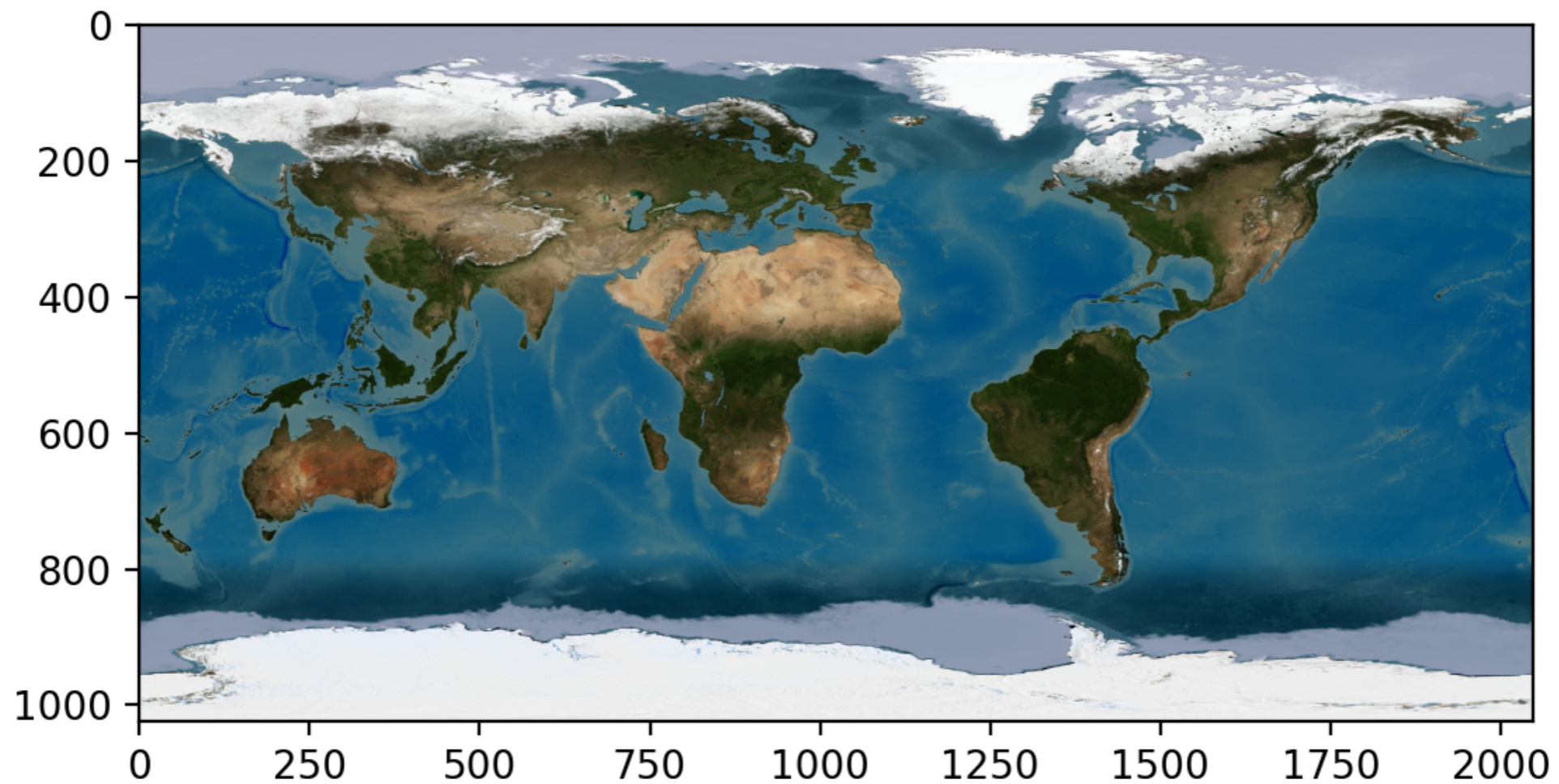
```
plt.imshow(im[::-1,])
plt.show()
```



# Slicing

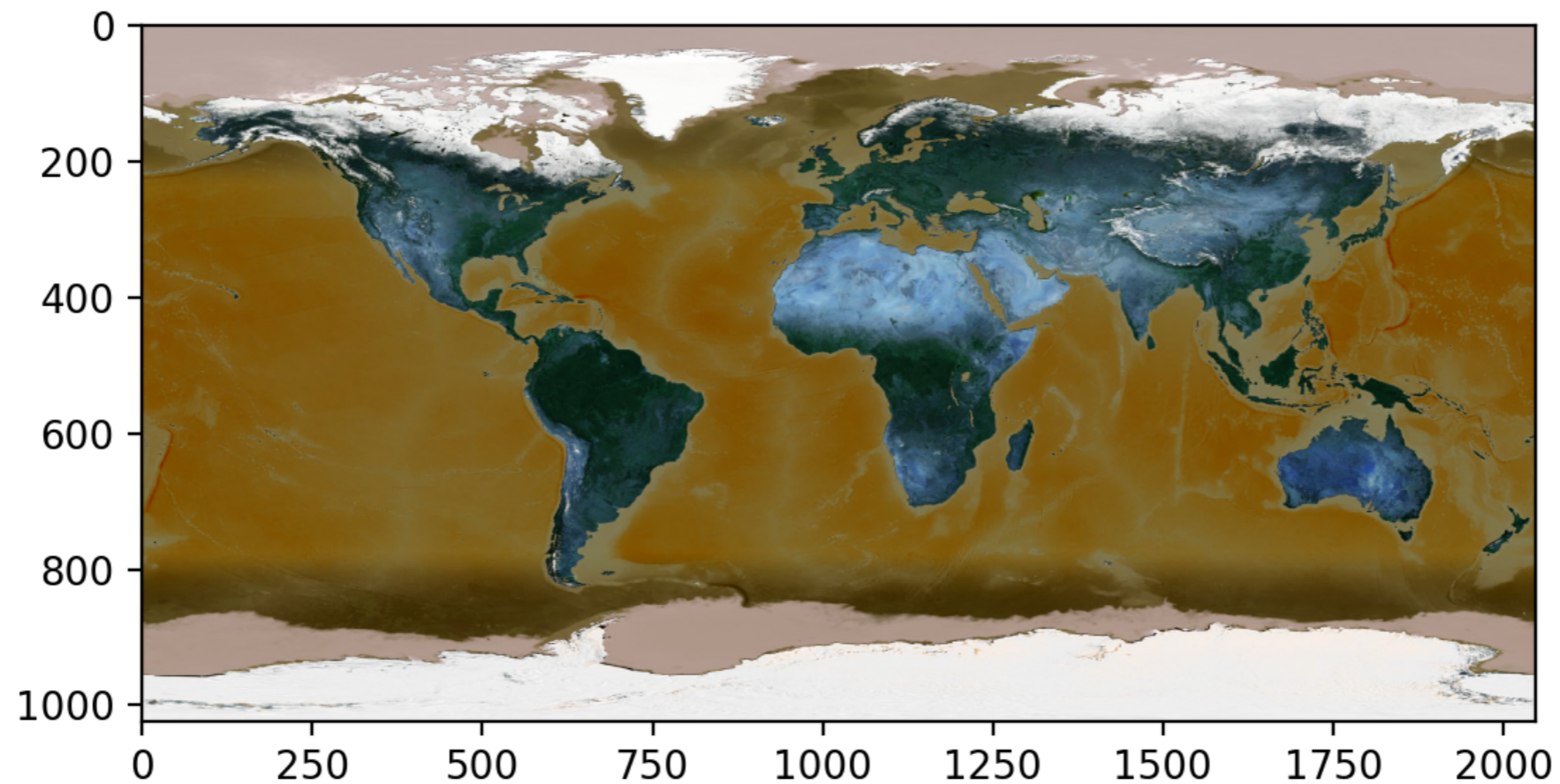
- Swap second coordinate

```
plt.imshow(im[:,::-1,])
plt.show()
```



# Slicing

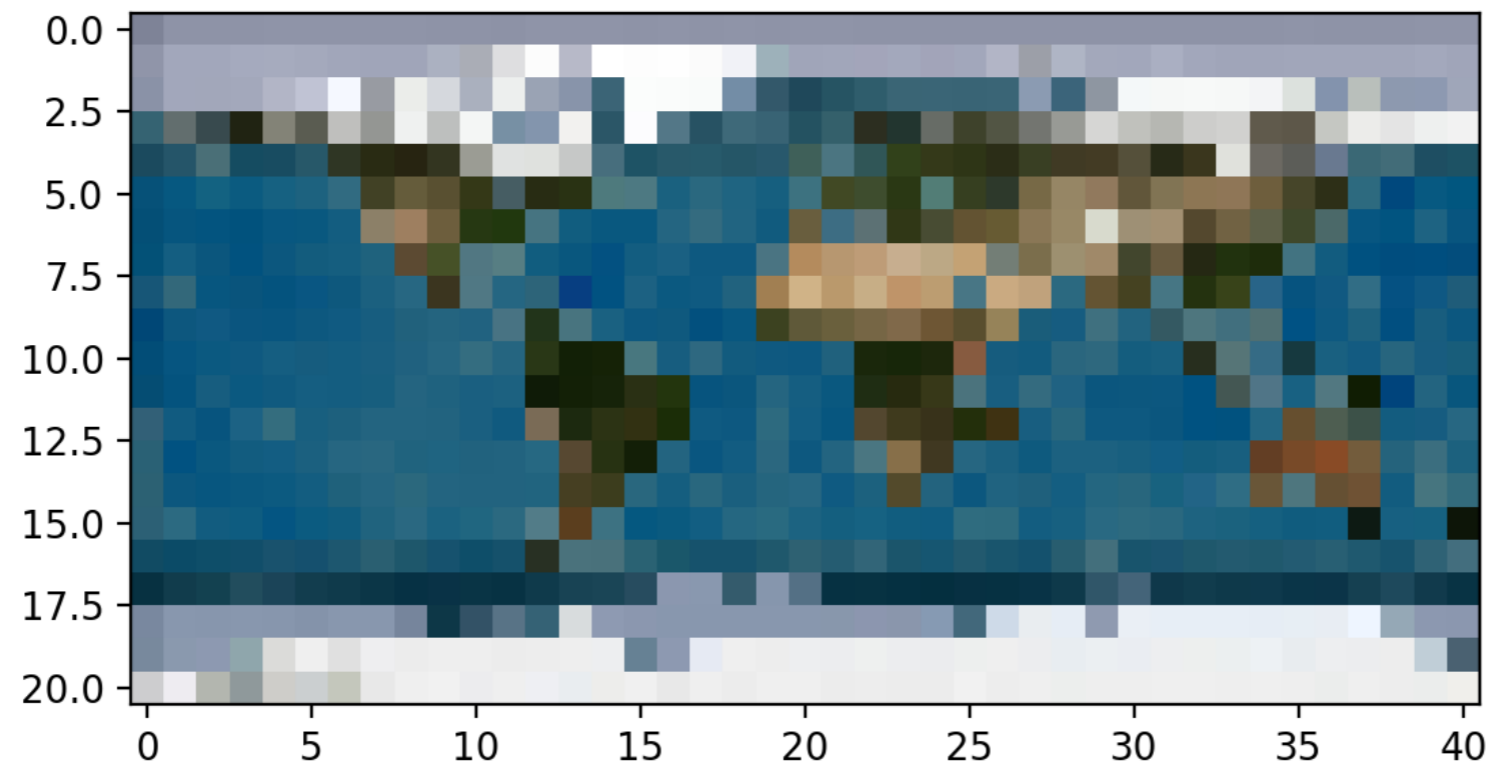
- Swap third coordinate (color coordinate)
  - `plt.imshow(im[:, :, ::-1])`  
`plt.show()`



# Slicing

- Take every 50th line and column

```
plt.imshow(im[::50, ::50,])
plt.show()
```

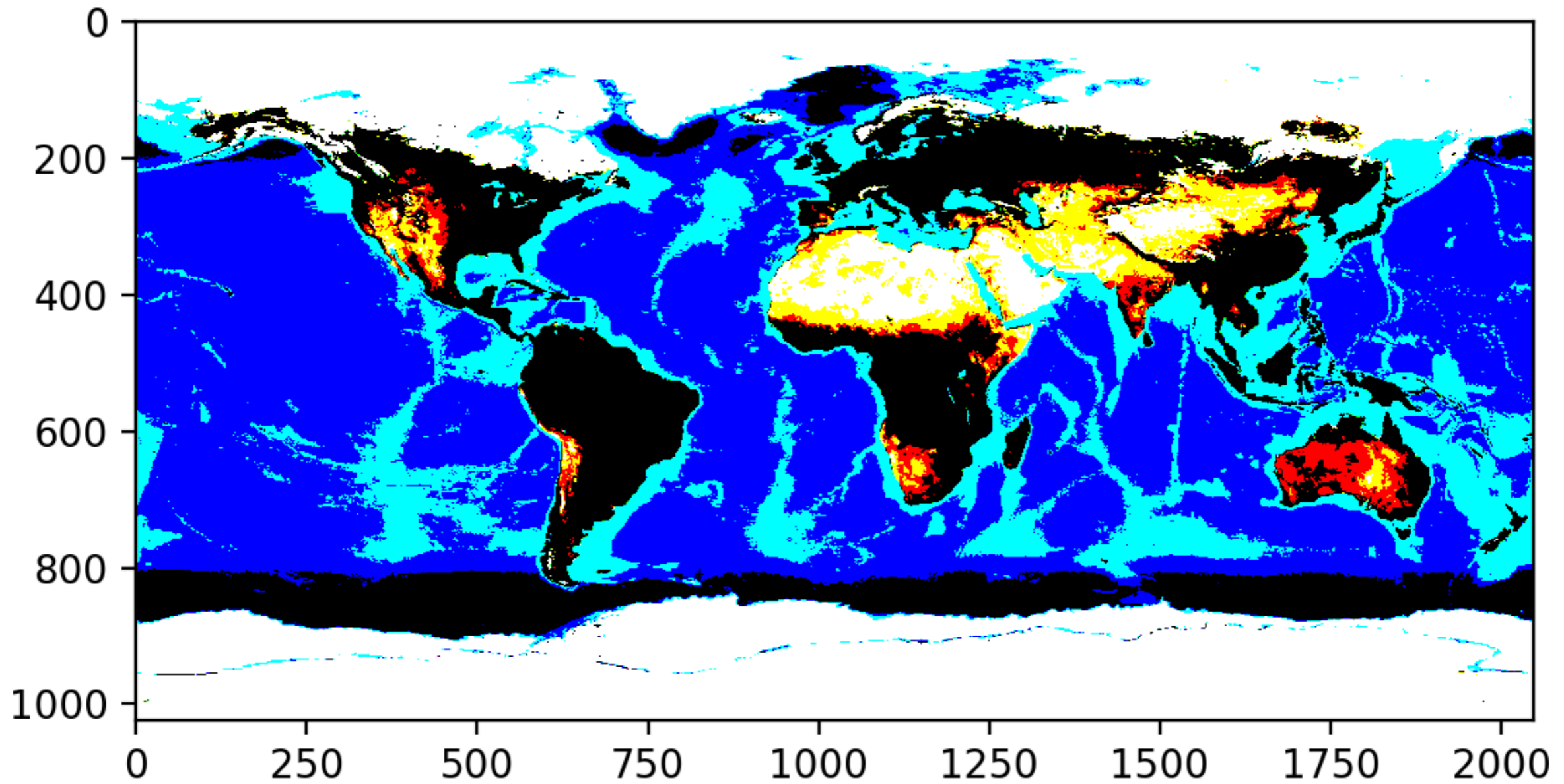


# Slicing

- We can also apply functions
  - `np.where` allows us to replace values
    - `image = np.where(im>100, 255, 0)`
      - Where-ever the value of the image is less than 100, replace it with 0
      - Otherwise, replace it with 255



# Slicing



# Slicing

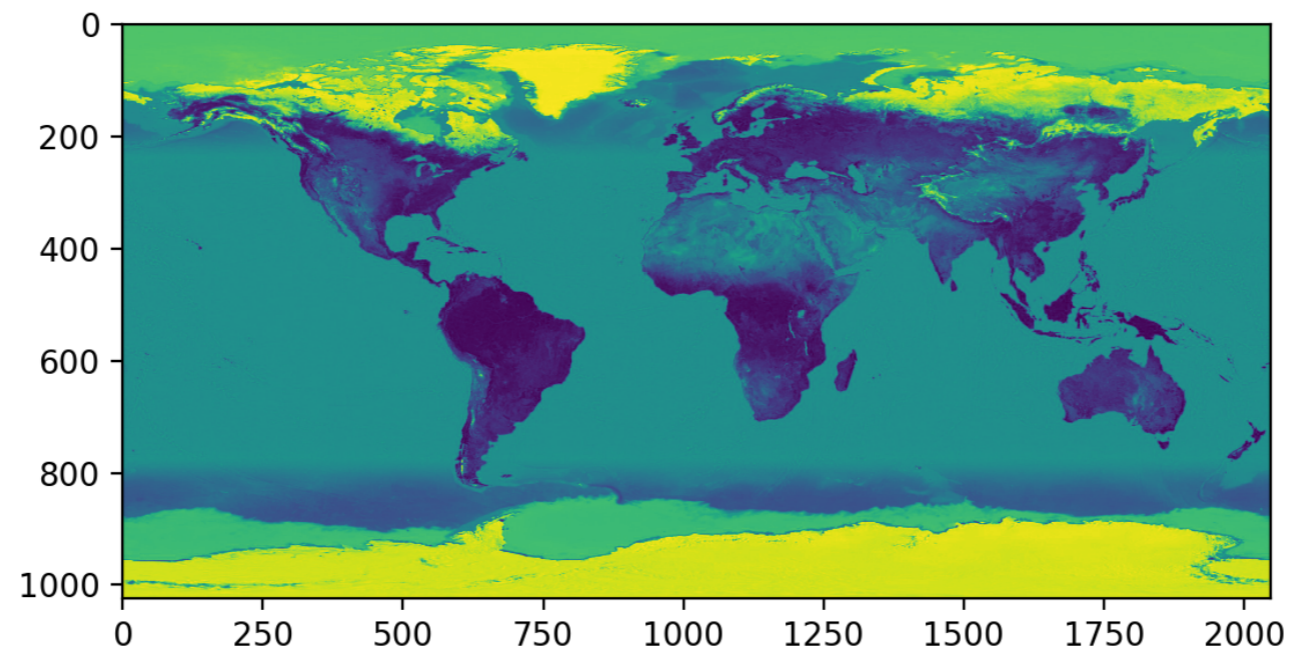
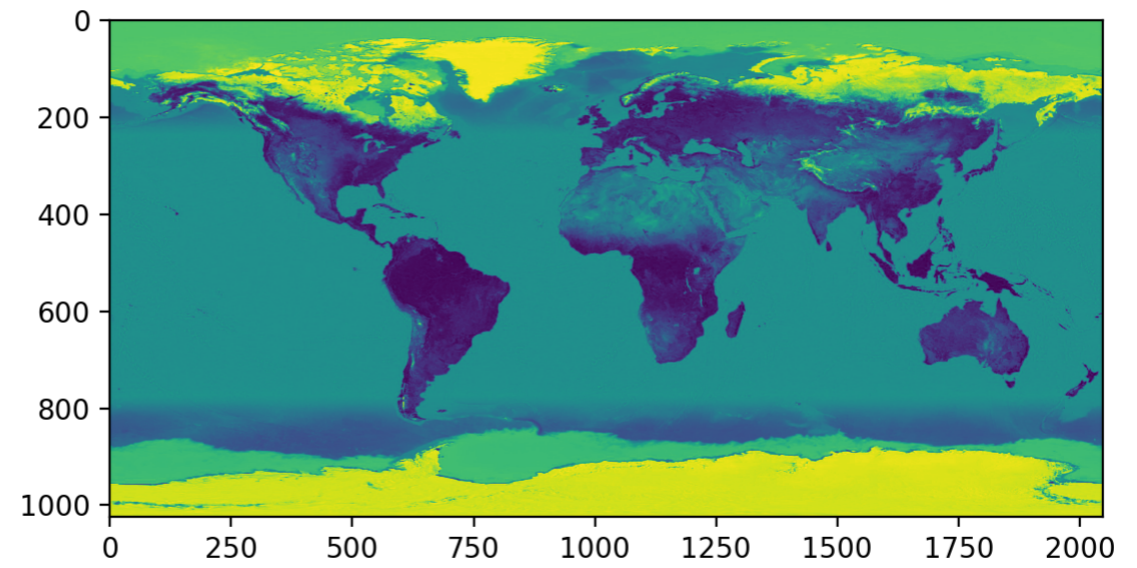
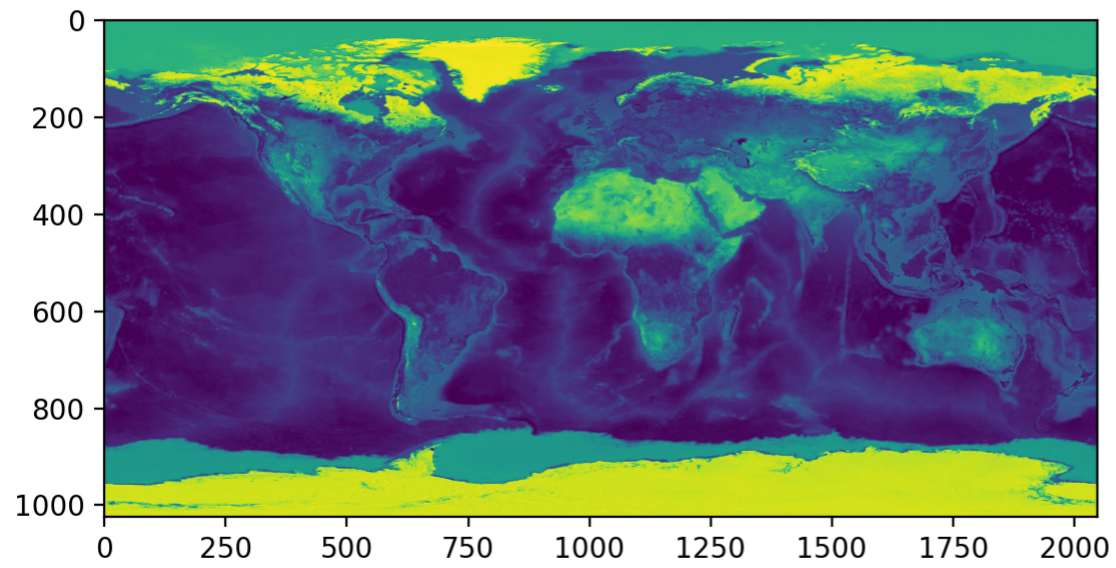
- Can use a sub-image

```
plt.imshow(im[:, :, 0])
plt.show()
```

```
plt.imshow(im[:, :, 1])
plt.show()
```

```
plt.imshow(im[:, :, 2])
plt.show()
```

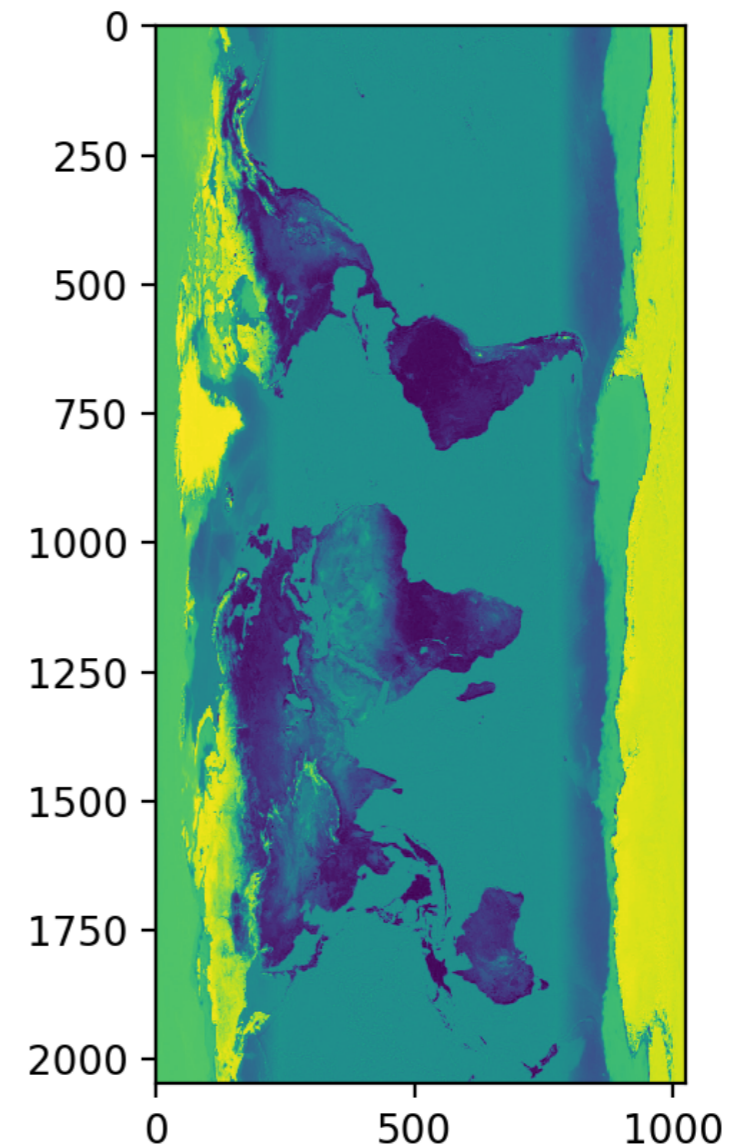
# Slicing



# Slicing

- Can use the transpose

```
plt.imshow(im[:, :, 2].T)
plt.show()
```



# Slicing

- Or just concentrate on the essential

```
plt.imshow(im[250:500, 1350:1600, :])
```

# Slicina

