

BE/BAT 485/585

Remote Sensing Data and Methods Lab - 3

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vip.arizona.edu
vegetation index & phenology Lab.
...Understanding a piece of the Earth system

 Institute of the
Environment
 USA npn
National Phenology Network
 USGS
LP DAAC
LAND PROCESS DATA ACTIVE ARCHIVE CENTER

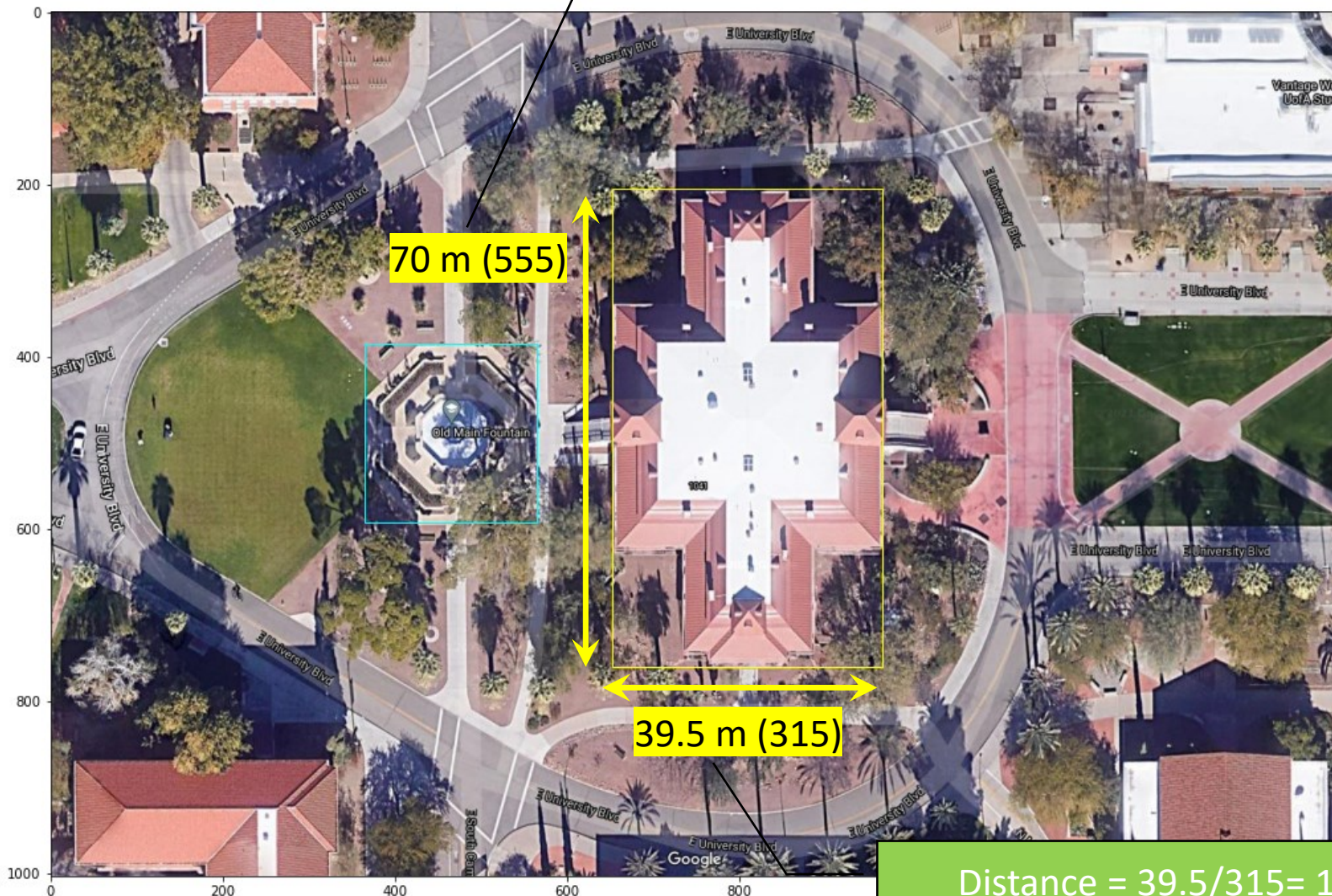

Revisit Lab #2

Finish your work, get help, or ensure accuracy.

Exercise #1: Pixel size

- Here is one way

$$\text{Distance} = 70/555 = 12.61 \text{ cm}$$



$$\text{Distance} = 39.5/315 = 12.54$$

New Lab #3 Work

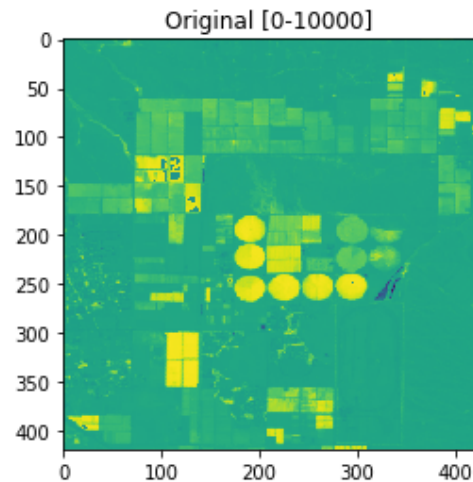
Today you will learn

- Degrading and Super sampling
 - There are two user defined functions that do that
 - They are straightforward and use simple averaging and/or repetition
 - Can you improve them? To use other sampling techniques/methods?
- You will learn some basic matplotlib functions
 - Load an image
 - Apply colors
 - Label the images
 - Add ticks
 - etc.
 - Create line plot
 - Graph Y vs. X
 - Superimpose more than one line graph
 - Label them
 - Add Title/X-Axis/Y-Axis

Basic Matplotlib operations

```
[10]: # Display Input band
plt.figure()
plt.title('Original [0-10000]')
plt.imshow(DataNDVI)
```

```
[10]: <matplotlib.image.AxesImage at 0x12950350d30>
```



Todo: Look at presentation

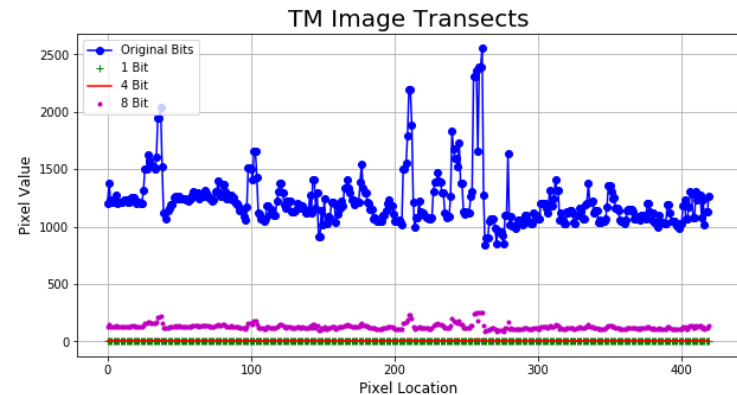
```
[20]: ### Here we print the row of ddat to show what happened to the original data
# Import plotting libraries
import matplotlib.pyplot as plt
import matplotlib.colors

# Create data holders for the data with the correct size
Xvalues=np.zeros(ncols)
for i in range(0,420):
    Xvalues[i]=i

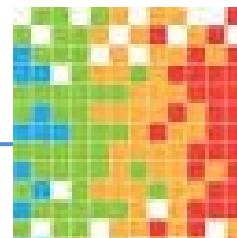
# And we plot as usual
plt.figure(figsize=(10,5))
plt.title('TM Image Transects',fontsize=20)
plt.xlabel('Pixel Location',fontsize=12)
plt.ylabel('Pixel Value',fontsize=12)

plt.plot(Xvalues,DataNDVI[15,:], "b-o", label="Original Bits")
plt.plot(Xvalues,Data_1bit[15,:], "g+", label="1 Bit")
plt.plot(Xvalues,Data_4bit[15,:], "r-", label="4 Bit")
plt.plot(Xvalues,Data_8bit[15,:], "m.", label="8 Bit")
# 'b', 'g', 'r', 'c', 'm', 'y', 'k', 'w';
plt.grid()
plt.legend(loc='upper left')
```

```
[20]: <matplotlib.legend.Legend at 0x26a525489e8>
```



Exercise #1: Resampling

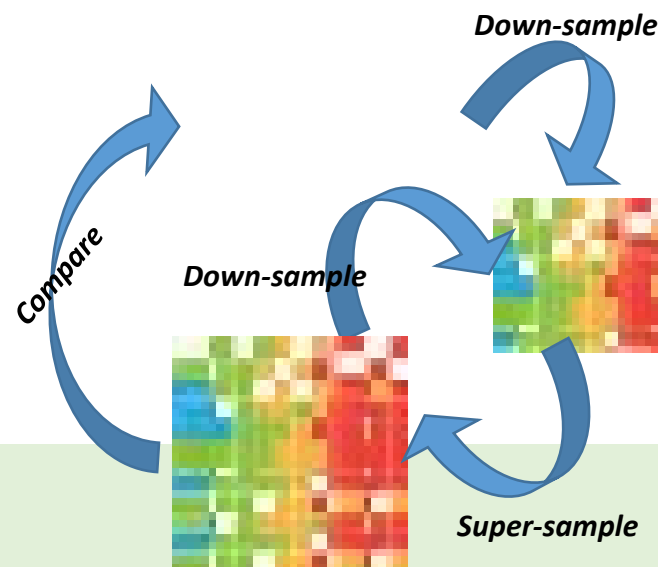


- **Resampling**

- Down sample (previous lab)
- Super-sample (increase resolution)
 - Why? Sometimes we need to super sample, so data sets match each other
- Downsample the resolution of an input image, then reverse it with super-sampling then compare the resulting data
 - **Is it a reversible process?**

- **Homework:**

- Down-sample then Super-sample the original image
 - Sample and graph a row of values from the original input image and compare to the super-sampled Image (same size).
 - What happened at each stage and how the pixel values changed
 - What does this tell you?
- Now repeat the “down-sampling then super sampling” multiple times and watch the impact by plotting some data

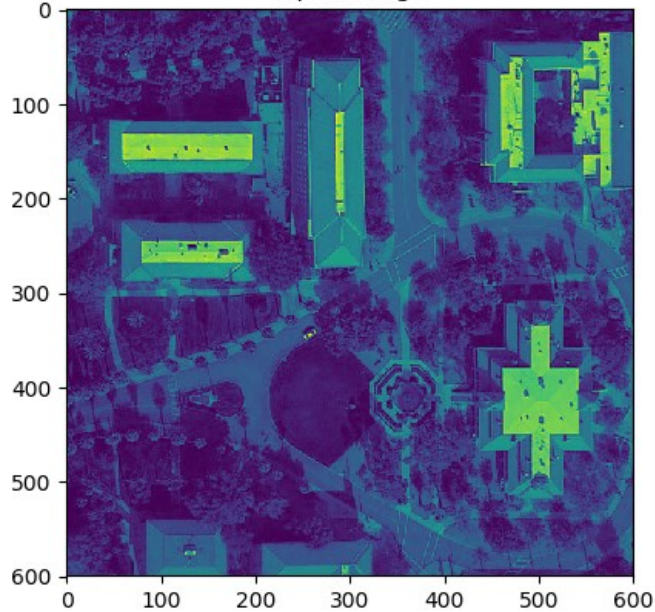


Instructions:

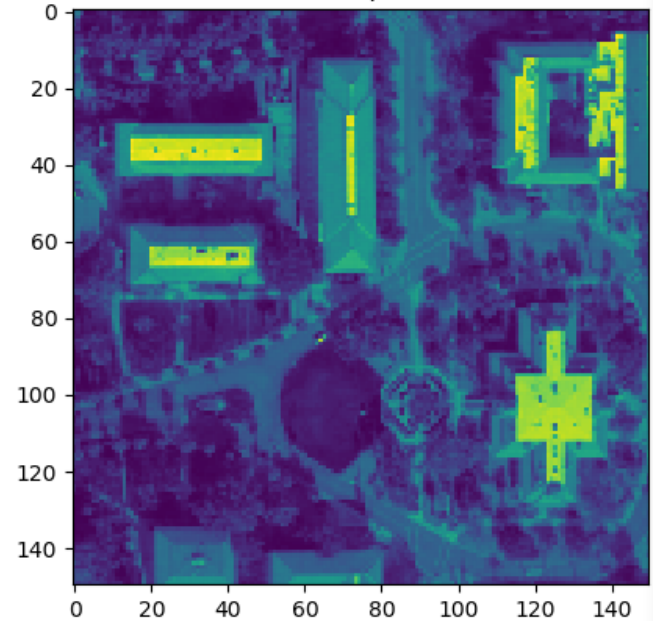
- Download from D2L files:
 - UA_Old_Main_New.jpg
 - viplab_lib.py
 - BE485_Lab_3_Ex1.ipynb

Resampling

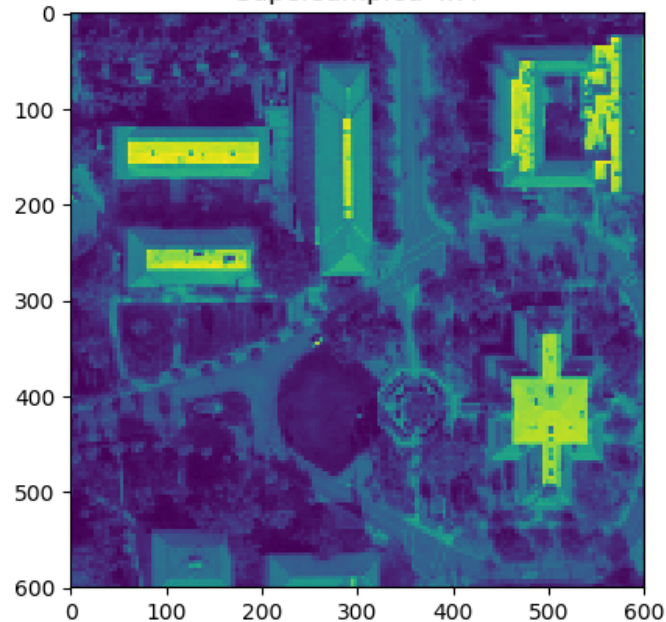
Input Image



Downsampled 4x4



Supersampled 4x4

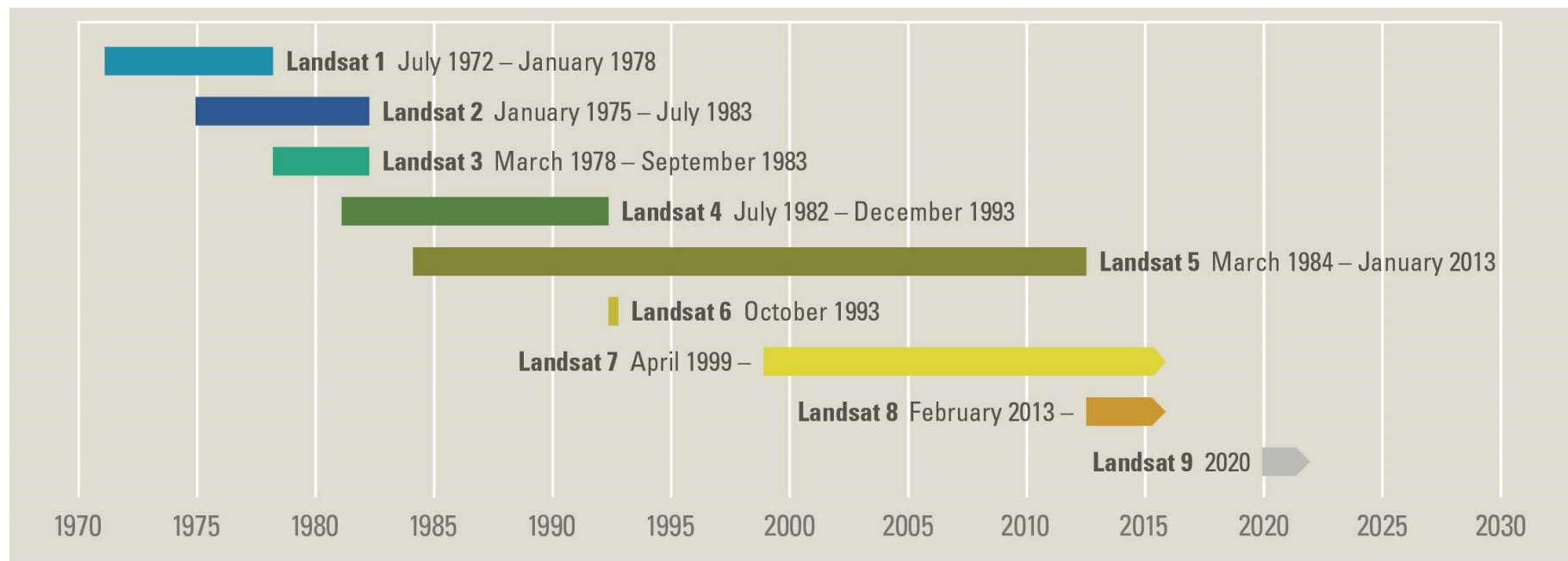


Notes:

- Look at the number of rows and columns
- Visually inspect the pixels
- Compare Images
- Was the process reversed?
 - Could this be a lossless process?
 - Recall the mixing from the lecture?

Landsat

- The Landsat series of satellites provides the longest temporal record of moderate resolution multispectral data of the Earth's surface on a global basis for over 40 years so far.
- The Landsat record has remained remarkably unbroken and is a unique resource to assist a broad range of specialists in managing the world's food, water, forests, and other natural resources for a growing world population. It is a record unmatched in quality, detail, coverage, and value (Source: [USGS](#))



Landsat Data naming convention

- When working with landsat, it is important to understand both the metadata and the file naming convention.
- The metadata tell you how the data were processed, where the data are from and how they are structured.
- The file names, tell you what sensor collected the data, the date the data were collected, and more.

Scene ID

LXSPPPRRRRYYYYDDDGSIIV

L = Landsat
X = Sensor
S = Satellite
PPP = WRS path
RRR = WRS row
YYYY = Year
DDD = Julian day of year
GSI = Ground station identifier
VV = Archive version number

Examples:

LC80290302015343LGN00
LE70160392004262EDC02
LT40170361982320XXX08
LM10170391976031AAA01

Landsat Product Identifier

LXSS_LLLL_PPPRRR_YYYYMMDD_yyyymmdd_CC_TX

L = Landsat
X = Sensor ("C" = OLI/TIRS Combined, "O" = OLI-only, "T" = TIRS-only, "E" = ETM+, "T" = TM, "M" = MSS)
SS = Satellite ("07" = Landsat 7, "08" = Landsat 8)
LLLL = Processing correction level ("L1TP": Precision Terrain, "L1GT": Systematic Terrain, "L1GS": Systematic)
PPP = WRS path
RRR = WRS row
YYYYMMDD = Acquisition year (YYYY) / Month (MM) / Day (DD)
yyymmdd = Processing year (yyyy) / Month (mm) / Day (dd)
CC = Collection number ("01", "02")
TX = Collection category ("RT" for Real-Time, "T1" for Tier 1, or "T2" for Tier 2)

Examples:

LC08_L1GT_029030_20151209_20160131_01_RT
LE07_L1TP_016039_20040918_20160211_01_T1
LT04_L1GS_017036_19821115_20160315_01_T2
LM01_L1GS_017039_19760131_20160225_01_T2

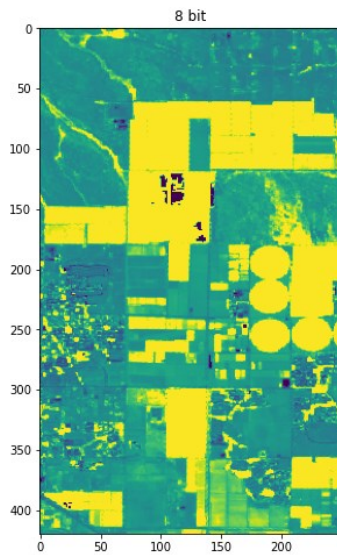
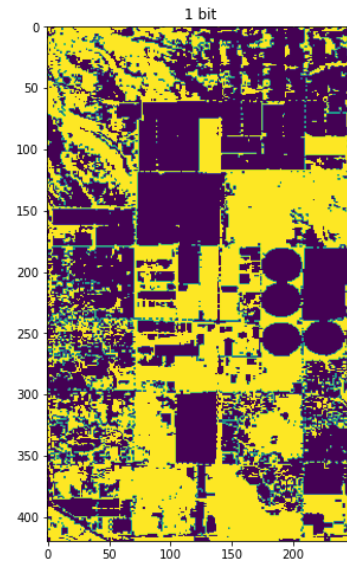
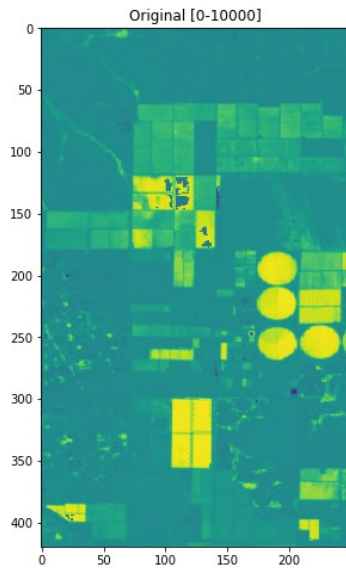
Exercise #2: Radiometric resolution

- Radiometric resolution
 - Read an image file/input (from Excel – we will eventually work directly with images)
 - Display the original image
 - Apply different Radiometric resolution
 - 1 or 8 bits from original (16 bits)
- **Homework**
 - Generate a dataset at 6 and 12 bits radiometric resolution
 - Display these images
 - Extract a row of values at each resolution and plot the values in an X-Y graph
 - Comment on what is happening to the data/values

Instructions:

- Download from D2L files:
 - LANDSAT8.A2017164.xlsx
 - BE485_Lab_3_Ex2.ipynb

Radiometric resolution



Recall this is like
scaling but in the data
domain

