

# Image Processing in Python

This is a tutorial

Make sure you have your laptop at hand!

# Logistics for Cloud Computing

- We will use Google Cloud Platform
- Everyone will get \$50 credits (see Canvas)
- Use your credits wisely!  
(They should be reserved for GPUs)

# Python Tutorial (Not Covered)

- We will support both Python 2.X & Python 3.X
- We recommend Python 3
- Python tutorial from Google (a good starting point)  
<https://developers.google.com/edu/python/>
- The official tutorial (TL;DR)  
<https://docs.python.org/3/tutorial/index.html>

# NumPy: Scientific Computing in Python

- Documentation

<https://docs.scipy.org/doc/numpy/reference/index.html>

- A quick tutorial

<https://docs.scipy.org/doc/numpy/user/quickstart.html>

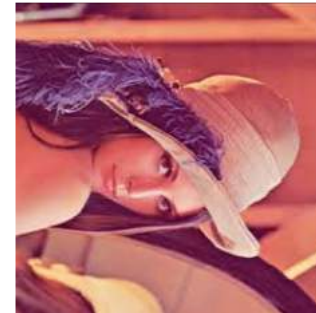
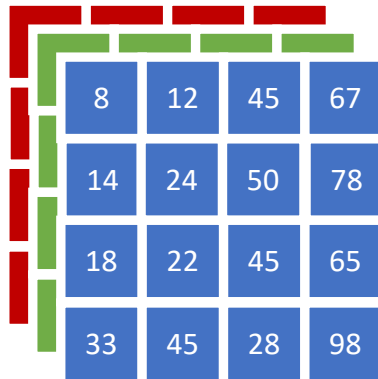
- If you are switching from Matlab to Numpy

<http://mathesaurus.sourceforge.net/matlab-numpy.html>

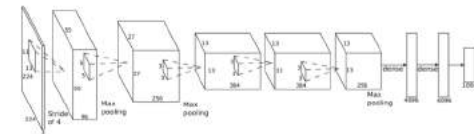
# Image Processing Packages in Python

- Python Image Library – PIL (latest dev **Pillow**)
  - Light weighted library for basic image processing
  - Well polished functions
  - Limited functionality, not very efficient
- Open Source Computer Vision (**OpenCV**)
  - Advanced library for image processing and computer vision
  - Good coverage of functionality, highly optimized
  - Sometimes can be buggy 😞

# Putting them together ...



Learning  
Based  
Methods



## Data IO

- Load images / videos
- Decode data
- Pillow/OpenCV

## Image Transforms

- Pre-processing
- Data augmentation
- NumPy + Pillow/OpenCV

And everything is done on the cloud ...

\* Thanks for the support from Google Cloud Platform



# Goals

20 min: Set up the cloud computing environment

When you need to use the GPUs

10 min: Basic image IO and image manipulations

How to use OpenCV and Pillow

20 min: Image Processing in Python

Resize, color transforms, filtering, rotation, ...

# Setup the Cloud

- Sign up for Google Cloud (use your wisc edu account)
- Cloud Console <https://console.cloud.google.com/>
- Create projects to better manage your computing resources
- Use the navigation bar on the left to
  - Create new instance – a virtual remote server for computing
  - Check your bill, monitor your instances, etc

# Create a Cloud Instance

- Go to “Compute Engine” -> “VM Instances” -> Create
- **Recommended:** Use this link to launch an pre-configured VM  
<https://cloud.google.com/deep-learning-vm/>
- You can choose CPUs / memory / GPUs

# Create a Cloud Instance (cont)

- You can specify CPUs and memory (“customize” in Machine type)
- **Make sure you choose your Deep Learning package**
- You will get an estimated monthly cost (**which is NOT accurate**)
- **When using GPUs, make sure you choose to install the driver**
- We recommend the Nvidia K80 to get started

# Examples

## Cloud Deep Learning VM Image

Preconfigured VMs for deep learning applications.

[VIEW DOCUMENTATION](#)[VIEW CONSOLE](#)

## Build your deep learning project fast on Google Cloud

Provision a VM quickly and effortlessly, with everything you need to get your deep learning project started on Google Cloud. Cloud Deep Learning VM Image makes it easy and fast to instantiate a VM image containing the most popular deep learning and machine learning frameworks on a Google Compute Engine instance. You can launch Compute Engine instances pre-installed with popular ML frameworks like TensorFlow, PyTorch, or scikit-learn. You can also add Cloud TPU and GPU support with a single click. You can either instantiate the image using the Google Cloud Platform (GCP) Cloud Marketplace UI or through Cloud SDK from the command line.

Deployment name

pytorch1

Zone ?

GPU availability is limited to certain zones. [Learn more](#)

us-central1-a

Machine type ?

2 vCPUs

13 GB memory

[Customize](#)

GPUs

The number of GPU dies is linked to the number of CPU cores and memory selected for this instance. For this machine type, you can select no fewer than 1 GPU die.

[Learn more](#)

Number of GPUs

1

GPU type

NVIDIA Tesla K80

**Machines with GPUs can't migrate on host maintenance**

Framework

Choose the primary machine learning framework you will be using. If the library you would like to use is not listed, choose the base image, which provides core packages.

PyTorch 1.0 + fastai 1.0 (CUDA 10.0)

Access to the Jupyter Lab

☐ **Beta. Enable access via URL instead of SSH ?**

Enabling this Beta feature allows you to access your JupyterLab instance using a URL. Anyone who is in the Editor or Owner role in your GCP project can access this URL. This feature is available only in the US, EU and Asia.

GPU

☒ **Install NVIDIA GPU driver automatically on first startup? ?**

I want to use NVIDIA GPUs with this image. Please fetch NVIDIA GPU drivers from a third-party location and install them on my behalf (requires internet access on the VM).

Boot Disk

Boot disk type ?

SSD Persistent Disk

K80

✔ pytoch-vm1 has been deployed

Overview - pytoch-vm1

tensorflow tensorflow.jinja

tensorflow-vm-tmpl vm\_instance.py

pytoch-vm1-vm vm instance

software-status software\_status.py

pytoch-vm1-config config

pytoch-vm1-software config waiter

# Access the Cloud Instance

- You can check your launched instance under “Compute Engine” -> “VM Instances”
- You can ssh into the instance using Google Console
- But we recommend using SSH keys to get access

# Access the Cloud Instance (cont)

- You can check your launched instance under “Compute Engine” -> “VM Instances”
- You can ssh into the instance using Google Console
- We recommend using SSH keys to get access the instance
  - Allow multiple users (your team members) to access the instance

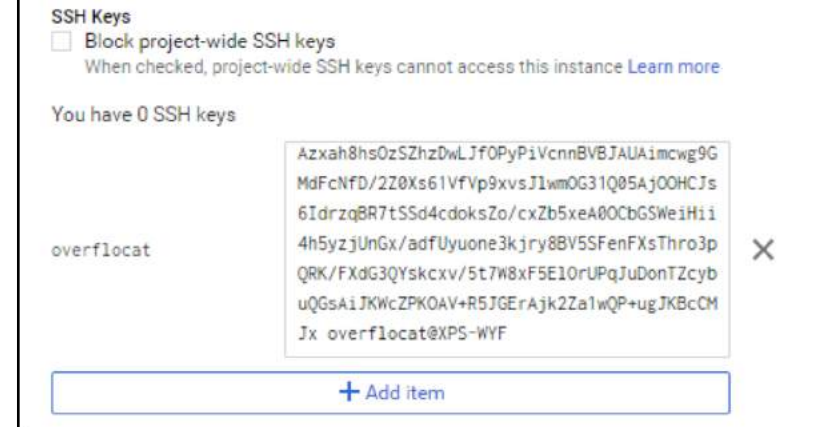
# Access the Cloud Instance (recommended)

- You will need the following tools
  - ssh-keygen: create your public and private key pairs
  - ssh client: to access the server
  - They are easy to setup in Linux/macOS
  - We recommend Putty for Windows users (<http://www.putty.org/>)
- Step 1: Create the SSH key pairs
  - *ssh-keygen -t rsa*
- Step 2: Copy your public key to Google Cloud
  - “Compute Engine” -> “VM Instances” -> Your instance -> edit -> SSH keys -> Add item -> Save
- Step 3: *ssh -i your\_private\_key your\_account@server\_ip*



# Examples (cont)

```
overflocat@XPS-WYF:~$ ssh-keygen -t rsa
Generating public/private rsa key pair.
Enter file in which to save the key (/home/overflocat/.ssh/id_rsa):
/home/overflocat/.ssh/id_rsa already exists.
Overwrite (y/n)? y
Enter passphrase (empty for no passphrase):
Enter same passphrase again:
Your identification has been saved in /home/overflocat/.ssh/id_rsa.
Your public key has been saved in /home/overflocat/.ssh/id_rsa.pub.
The key fingerprint is:
SHA256:mbSXzyhsqOCmkYKuvBiRWY/e79+jlFX1YXFKJPYqw64 overflocat@XPS-WYF
The key's randomart image is:
+---[RSA 2048]-----+
|
| + o
| + . . .
| . o . . . + o
| . + . . S o + . o +
| + o . . o . * . o =
| * . . o + . + o = .
| . = . o . ++ .
| . o . . E + + + .
+---[SHA256]-----+
overflocat@XPS-WYF:~$ ls
chinadns.txt      google-cloud-sdk-203.0.0-linux-x86_64.tar.gz  movie
download          hw3_grader                                     stop.sh
google-cloud-sdk  img                                             styletransfer
overflocat@XPS-WYF:~$ cd .ssh
overflocat@XPS-WYF:~/.ssh$ ls
config            google_compute_engine.pub  id_rsa            known_hosts
google_compute_engine  google_compute_known_hosts  id_rsa.pub
overflocat@XPS-WYF:~/.ssh$ cat id_rsa.pub
ssh-rsa AAAAB3NzaC1yc2EAAAADAQABAAQDQJLR3rHIYQ2o0p5bOCmokWldjsx8XzMk3IWppTyi9brWaQA8CnH
yxBtzCzL12s+x36gSgd8zKewSmrfof45cRQB9Azhah8sOzSZhzDwLJf0PyPiVcnnBVBJAUAimcw9GMdFcNfD/2Z
0Xs61VFvp9xvsJlwmOG31Q05Aj00HCJs6IdrzqBR7tSSd4cdoksZo/cxZb5xeA00CbGSWeiHii4h5yzjUnGx/adfU
yuone3kjry8BV5SFenFXsThro3pQRK/FXdG3QYskcxv/5t7W8xF5E10rUPqJuDonTZcybuQGsAiJKWcZPKOAV+R5J
GErAjk2Za1wQP+ugJKBCmJx overflocat@XPS-WYF
overflocat@XPS-WYF:~/.ssh$
```



```
overflocat@XPS-WYF:~/.ssh$ ssh -i id_rsa overflocat@35.225.247.205
The authenticity of host '35.225.247.205 (35.225.247.205)' can't be established.
ECDSA key fingerprint is SHA256:Gk+odxhY90yln2+SE/s8/IU1JmfisL49JqqPeGwP1Uc...
Are you sure you want to continue connecting (yes/no)? yes
Warning: Permanently added '35.225.247.205' (ECDSA) to the list of known hosts.
=====
Welcome to the Google Deep Learning VM
=====

Version: m19
Based on: Debian GNU/Linux 9.7 (stretch) (GNU/Linux 4.9.0-8-amd64 x86_64)

Resources:
* Google Deep Learning Platform StackOverflow: https://stackoverflow.com/questions/tagge
d/google-dl-platform
* Google Cloud Documentation: https://cloud.google.com/deep-learning-vm
* Google Group: https://groups.google.com/forum/#!forum/google-dl-platform

To reinstall Nvidia driver (if needed) run:
sudo /opt/deeplearning/install-driver.sh
This image uses python 3.7 from the Anaconda. Anaconda is installed to:
/opt/anaconda3/

Linux pytoch-vm1-vm 4.9.0-8-amd64 #1 SMP Debian 4.9.130-2 (2018-10-27) x86_64

The programs included with the Debian GNU/Linux system are free software;
the exact distribution terms for each program are described in the
individual files in /usr/share/doc/*/copyright.

Debian GNU/Linux comes with ABSOLUTELY NO WARRANTY, to the extent
permitted by applicable law.
overflocat@pytoch-vm1-vm:~$ python
Python 3.7.1 (default, Dec 14 2018, 19:28:38)
[GCC 7.3.0] :: Anaconda, Inc. on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> import torch
>>>
```

# Now that you have an instance ...

- **The server has to be on Linux**, but you can still ssh from a window box
- Make sure you are using the right version of Python!
- PIL and OpenCV are pre-installed if you are using the pre-configured instance.
- Otherwise, you need to pip install them

# Do remember to terminate your instance!

“Compute Engine” -> “VM Instances” -> “Delete”

# Image IO

- Load / Save an image using PIL / OpenCV
- Convert the loaded image into NumPy array
- PIL and OpenCV loads a DIFFERENT channel ordering!
- Image resolution (W, H) & Data Type (uint8)

# Images in Python

- Suppose we have a  $N \times M$  RGB image called “im”
  - $\text{im}(0,0,0)$  = top-left pixel value in 1st channel
  - $\text{im}(y, x, b)$  =  $y$  pixels down,  $x$  pixels to right in the 2nd channel
  - $\text{im}(N-1, M-1, 2)$  = bottom-right pixel in the 3rd channel

**column** →

**row** ↓

**R or B**

**G**

**B or R**

Slide Cr

Slide Credit: James Hays

# Image Manipulations

- Read a pixel value
- Modify pixel values within an image (careful: data type!)
- Swap the color channels
- Flip the image (vertical / horizontal)
- Crop a region from the image

# Basic IO & Manipulation (PIL)

```
from PIL import Image

im = Image.open("dog.jpg")
w, h = im.size
print('Original image size: %sx%s' % (w, h))
print('Image mode is: %s' % im.mode)
```

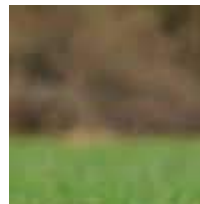
Original image size: 313x161  
Image mode is: RGB

```
out_hf = im.transpose(Image.FLIP_LEFT_RIGHT)
out_vf = im.transpose(Image.FLIP_TOP_BOTTOM)
out_hf.save("dog_hf.jpg")
out_vf.save("dog_vf.jpg")
box = (0, 0, 100, 100)
region = im.crop(box)
region.save("dog_crop.jpg")
```

```
import numpy as np

region_np = np.array(region)
print('Data type is:', region_np.dtype)
w, h, c = region_np.shape
print('Original image size: %sx%sx%s' % (w, h, c))
region_np[0:20, 0:20] = np.array([0, 0, 255])
region_new = Image.fromarray(np.uint8(region_np))
region_new.save("dog_crop_new.jpg")
```

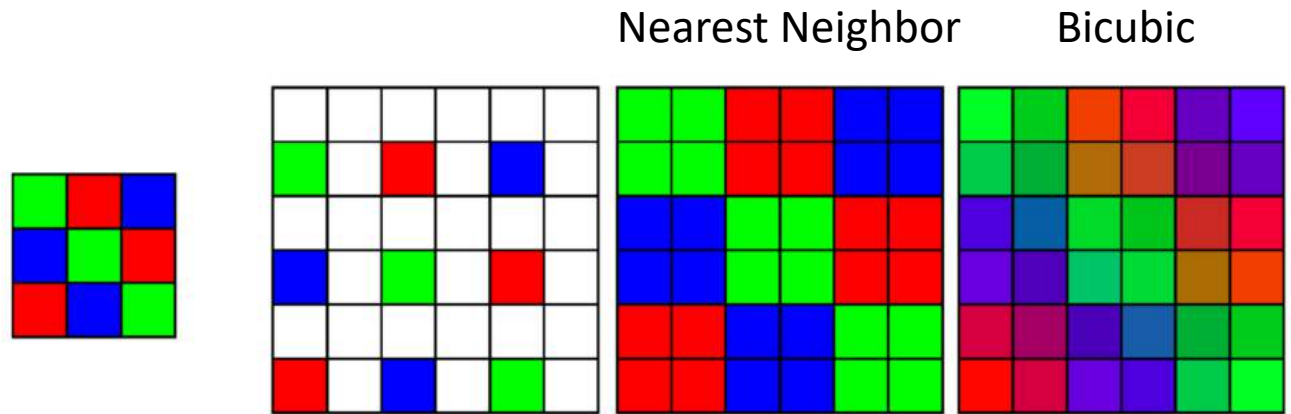
Data type is: uint8  
Original image size: 100x100x3



The interactive OpenCV version is also uploaded on Canvas.

# Image Resizing (Zoom in / out)

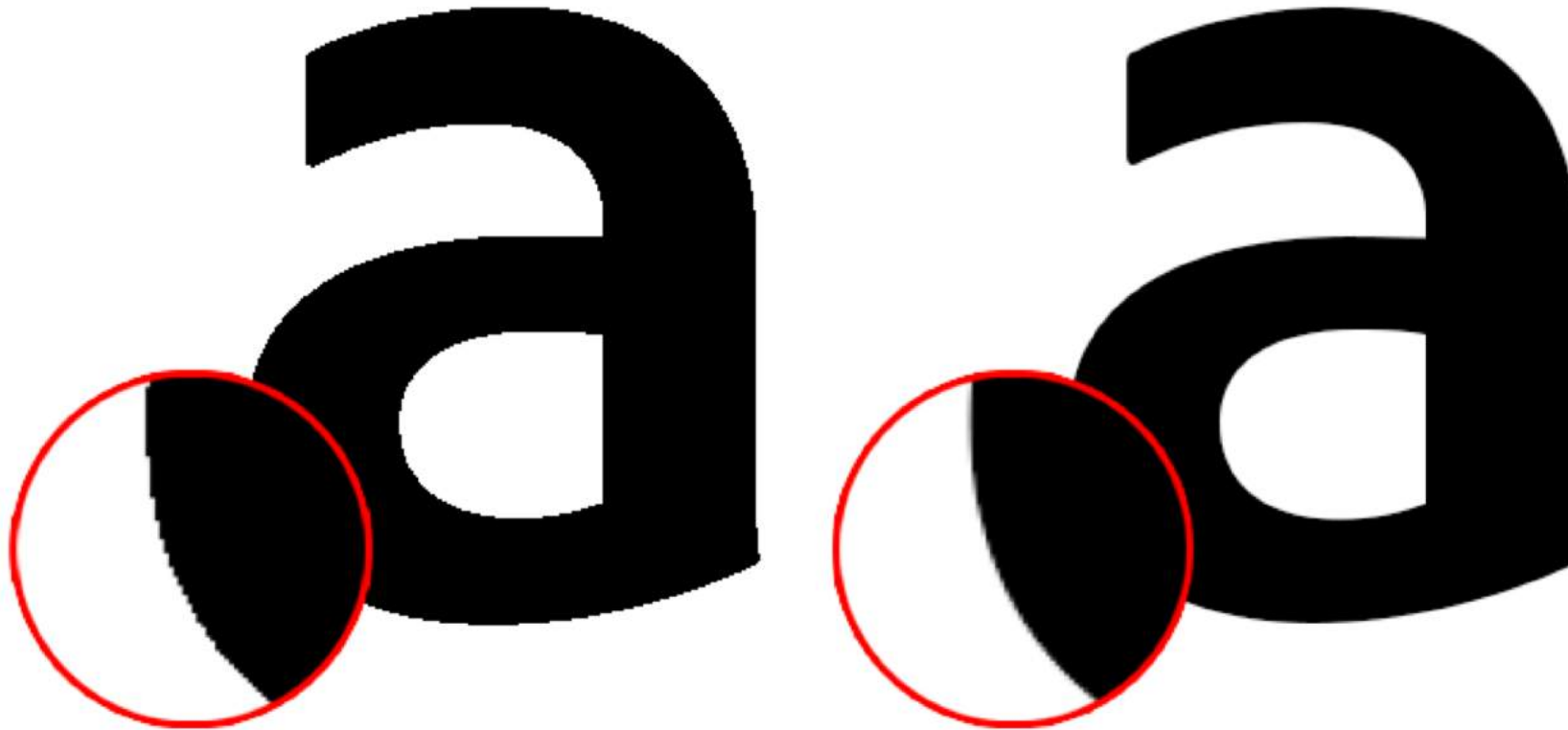
- Increase / reduce the resolution of the image
- Different interpolation schemes (how to fill in the pixels)
- Sometimes can be tricky





# Image Resizing – The Tricks

- Anti-aliasing (blur a bit before resizing)

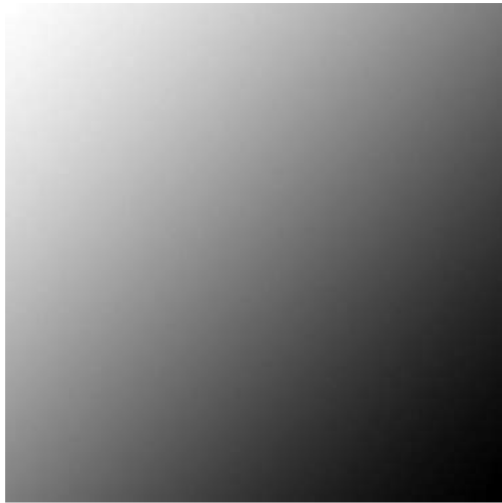


# What if you want to resize a mask?



# Even nearest neighbor interpolation can be buggy!

Input Image



OpenCV



Pillow



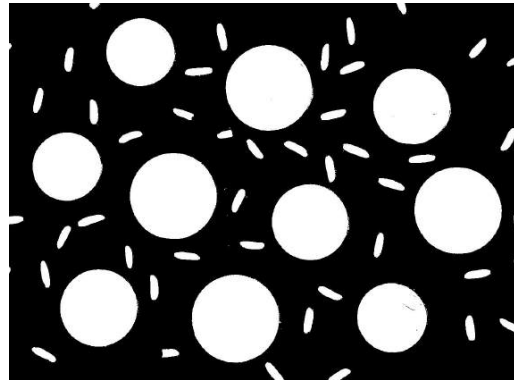
# Image Upsampling (PIL)

```
im = Image.open('lenna.jpg')
w, h = im.size
print('Original image size: %sx%s' % (w, h))
im_up_nn = im.resize((w*10, h*10), resample=Image.NEAREST)
im_up_bl = im.resize((w*10, h*10), resample=Image.BILINEAR)
im_up_bc = im.resize((w*10, h*10), resample=Image.BICUBIC)
w, h = im_up_nn.size
print('Upsampled image size: %sx%s' % (w, h))
im_up_nn.save('lenna_up_nn.jpg')
im_up_bl.save('lenna_up_bl.jpg')
im_up_bc.save('lenna_up_bc.jpg')
```

Original image size: 326x326  
Upsampled image size: 3260x3260



Original Image 1



Original Image 2  
(From CS534)

The interactive OpenCV version is also uploaded on Canvas.



Nearest Neighbor



Nearest Neighbor



Bilinear



Bilinear



Bicubic



Bicubic

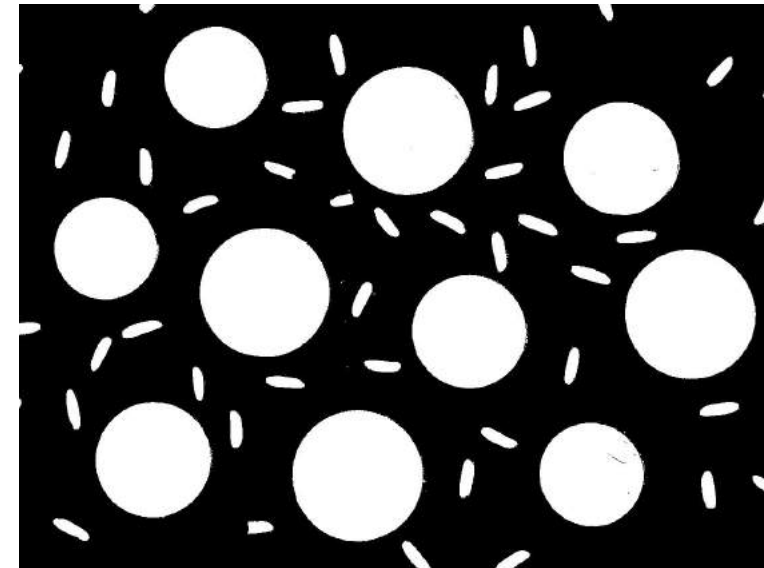
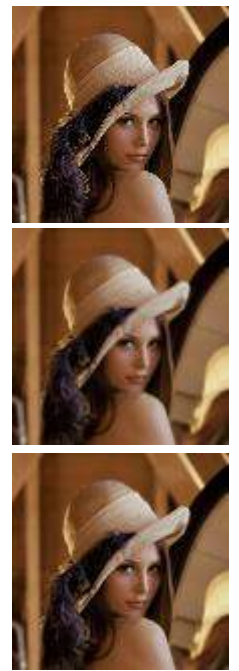
# Image Downsampling (PIL)

```
im = Image.open('lenna.jpg')
w, h = im.size
print('Original image size: %sx%s' % (w, h))
im_up_nn = im.resize((w//3, h//3), resample=Image.NEAREST)
im_up_bl = im.resize((w//3, h//3), resample=Image.BILINEAR)
im_up_bc = im.resize((w//3, h//3), resample=Image.BICUBIC)
w, h = im_up_nn.size
print('Downsampled image size: %sx%s' % (w, h))
im_up_nn.save('lenna_down_nn.jpg')
im_up_bl.save('lenna_down_bl.jpg')
im_up_bc.save('lenna_down_bc.jpg')
```

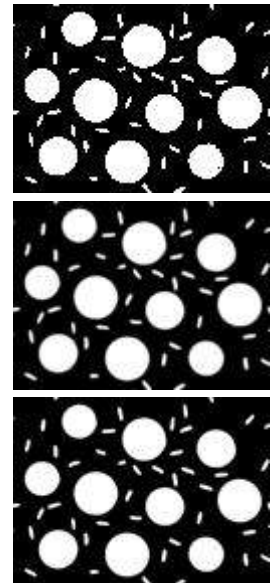
Original image size: 326x326  
Downsampled image size: 108x108



Original Image 1



Original Image 2  
(From CS534)

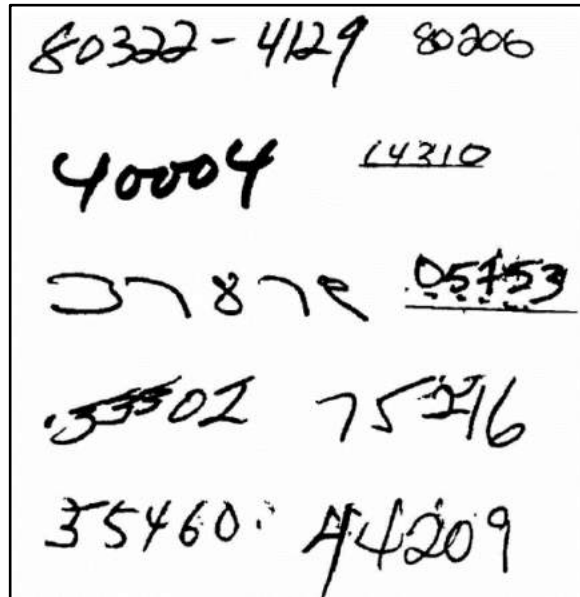


The interactive OpenCV version is also uploaded on Canvas.



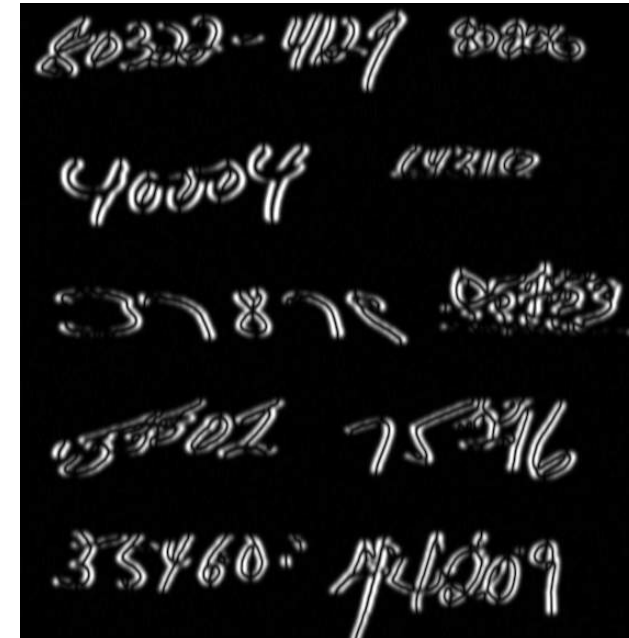
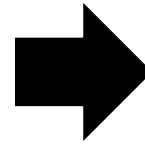
# Image Filtering

Input



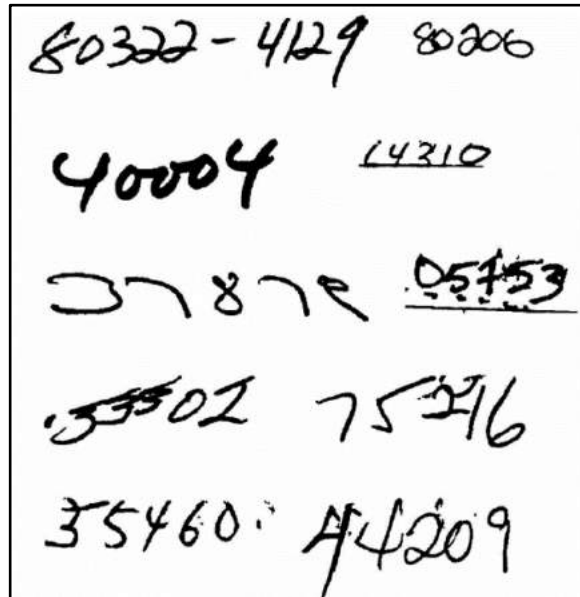
-1	0	+1
-2	0	+2
-1	0	+1

x filter



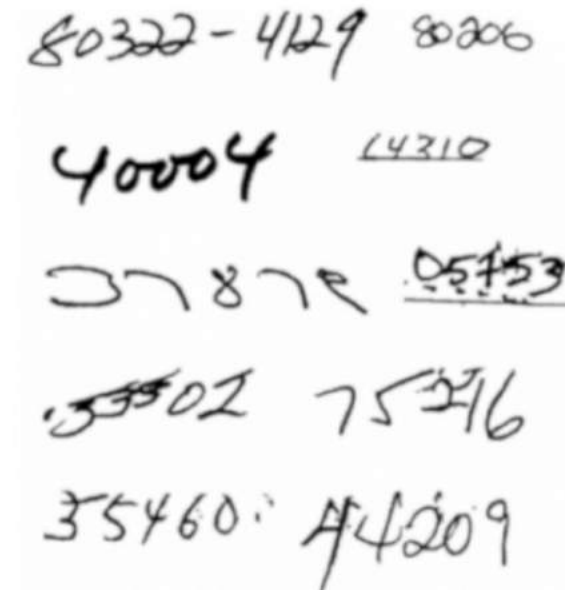
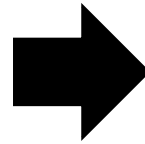
# Image Filtering

Input

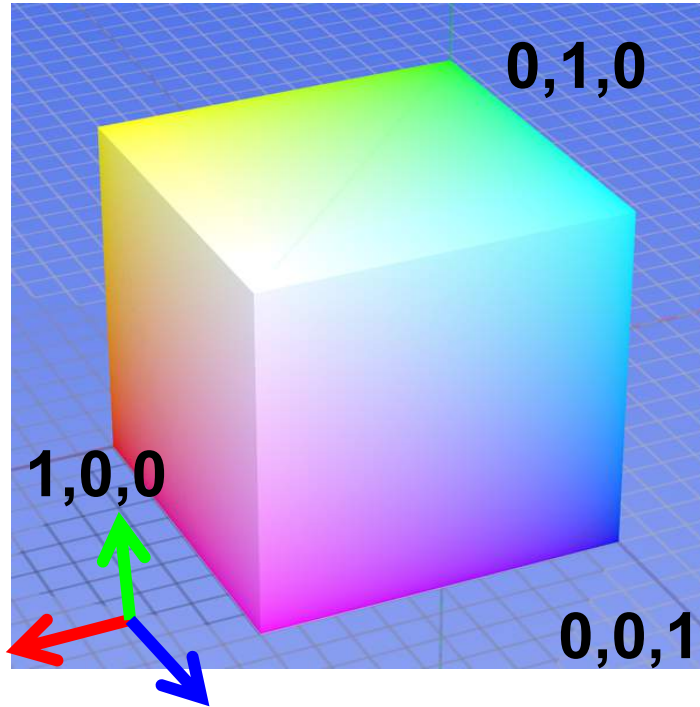


$\frac{1}{16} \times$

1	2	1
2	4	2
1	2	1



# Image Color Space: RGB



Some drawbacks

- Strongly correlated channels
- Non-perceptual



**R**

(G=0,B=0)



**G**

(R=0,B=0)



**B**

(R=0,G=0)





# Image Color Perturbation

- Rescale the color channels a bit (color jittering)
- Look quite different yet both can be realistic



# Image Rotation & Warping



translation



rotation



aspect



affine



perspective

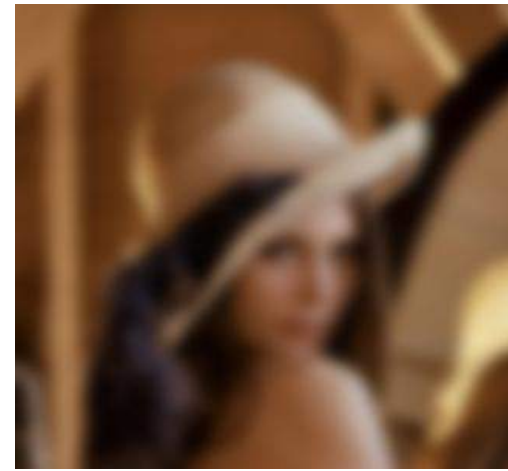


cylindrical

# Image Processing Examples (PIL)

```
import numpy as np
from PIL import Image
from PIL import ImageFilter
im = Image.open('lenna.jpg')
w, h = im.size
print('Original image size: %sx%s' % (w, h))
im_gaussian = im.filter(ImageFilter.GaussianBlur(radius=5))
im_rotated_0 = im.rotate(30, expand=0)
im_rotated_1 = im.rotate(30, expand=1)
im_np = np.array(im).astype(float)
im_np[:, :, 0] = im_np[:, :, 0] / 2
im_jittered = Image.fromarray(np.uint8(im_np))
im_gaussian.save('lenna_gaussian.jpg')
im_rotated_0.save('lenna_rotated_0.jpg')
im_rotated_1.save('lenna_rotated_1.jpg')
im_jittered.save('lenna_jittered.jpg')
w, h = im_rotated_0.size
print('Size without expansion: %sx%s' % (w, h))
w, h = im_rotated_1.size
print('Size with expansion: %sx%s' % (w, h))
```

Original image size: 326x326  
Size without expansion: 326x326  
Size with expansion: 446x446



Gaussian Filtering



Color Perturbation



Rotation w/ expansion



Rotation w/o expansion

The interactive OpenCV version is also uploaded on Canvas.




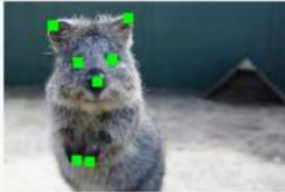









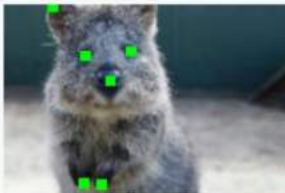
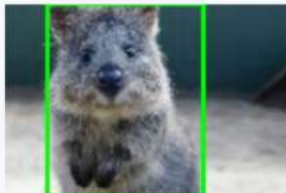




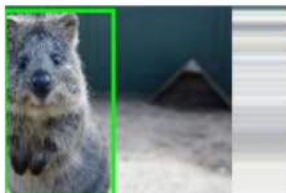


# Multiple (Random) Image Transforms

- Flip horizontally
- Color jittering
- Rotation



# Multiple Image Transforms + Label Transforms

	Image	Heatmaps	Seg. Maps	Keypoints	Bounding Boxes
<i>Original Input</i>					
Gauss. Noise + Contrast + Sharpen					
Affine					
Crop + Pad					

<https://github.com/aleju/imgaug>

# Multiple Image Transforms as Data Augmentation

- Image transforms can drastically change the pixel values
- But the semantics often remain the same
- Get multiple different versions of the same data sample
- Help the learning, e.g., via preventing overfitting

- Create and access an instance on the Cloud
- Basic image IO and manipulations
- Simple image transforms

*The end*