

Data Input and Augmentation using Keras for Object Classification

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Problem:

- Complex Nilgai cattle-tick eradication project
- Trail cam data
- Primed for computer vision solution



Object Classification



cow



nilgai



pig

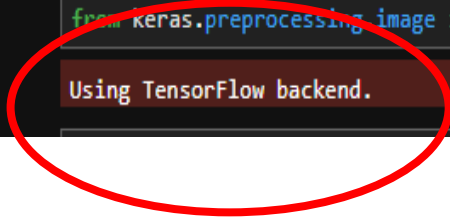


3 classes

Keras and Tensorflow 2 as a “backend”

- Keras is an open-sourced neural-network library written in Python
- Provides high-level building blocks for deep learning models
- Keras does not handle low-level operations
- It relies on a specialized tensor manipulation library to do so, serving as the "backend engine" of Keras
- Allows developers to focus on high-level development

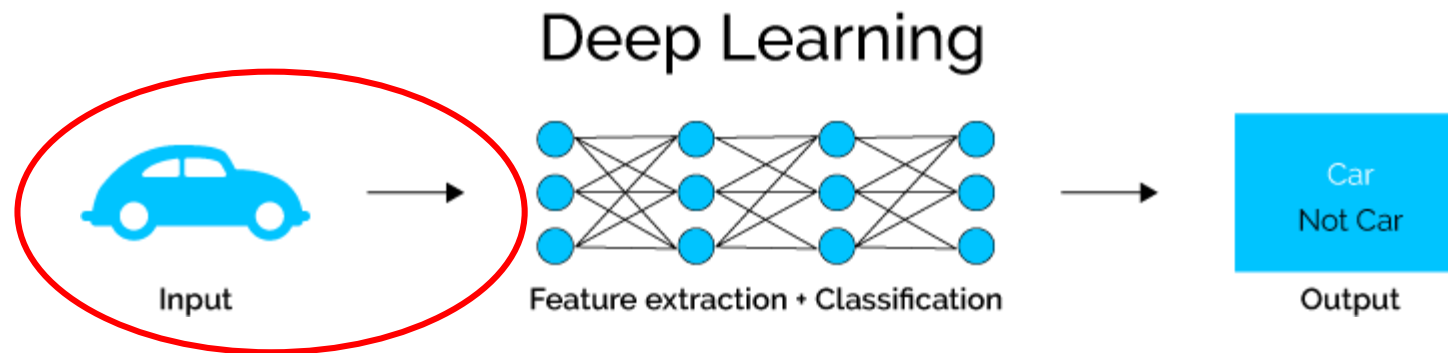
```
[1]: # example of normalizing a image dataset
      #import tensorflow as tf
      from keras.datasets import mnist
      from keras.preprocessing.image import ImageDataGenerator
```



Using TensorFlow backend.

Why is “Data input” a thing?

- Unstructured data (images) and labels
- 100 – 100,000 example dataset difficult to organize
- Automatic generation of train, validation, and test sets
- Progressive loading of image data that may not fit in system memory



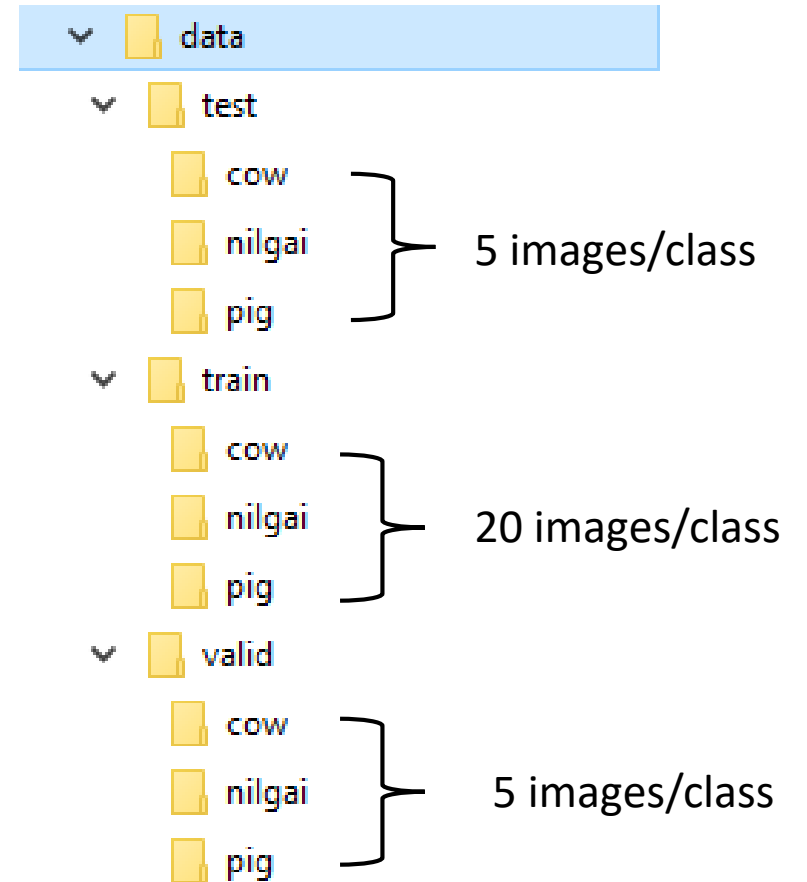
Organizing → Loading → Preprocessing

Organize train, validation, and test image datasets

Training set: examples used for learning - to fit the parameters of the classifier

Validation set: examples used to finetune the parameters of a classifier

Test set: examples used only to assess the performance of a fully-trained classifier (5 images)

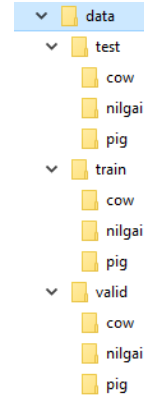


Loading data using ImageDataGenerator

```
# Import dependencies and TF for good measure
import tensorflow as tf
from keras.datasets import mnist
from keras.preprocessing.image import ImageDataGenerator

#check tf version
tf.__version__

'2.1.0-dev20191126'
```



ImageDataGenerator()

- Generate batches of tensor image data with on-the-fly data augmentation. The data will be looped over (in batches).

flow_from_directory

- Reads images from folders

```
train_path = 'C:\\Users\\ncu116\\Desktop\\data\\train'
val_path = 'C:\\Users\\ncu116\\Desktop\\data\\valid'
test_path = 'C:\\Users\\ncu116\\Desktop\\data\\test'
# create data generator
datagen = ImageDataGenerator()

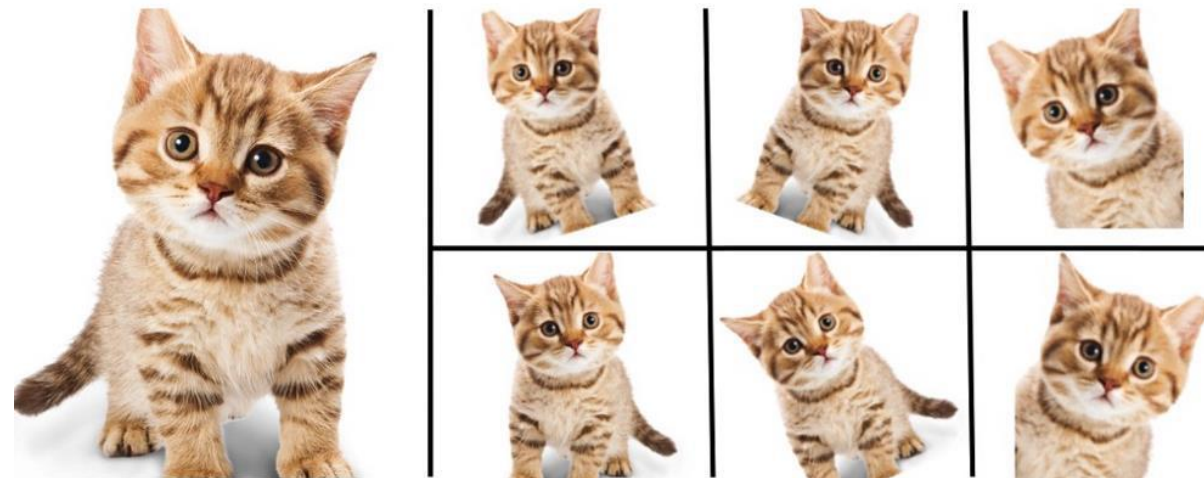
# prepare an iterators for each dataset arg = data directory and
#class_mode = classification type - either multiclass 'categorical' or 'binary'
train_it = datagen.flow_from_directory(train_path, class_mode='categorical')
val_it = datagen.flow_from_directory(val_path, class_mode='categorical')
test_it = datagen.flow_from_directory(test_path, class_mode='categorical')

# confirm iterator works
batchX, batchy = train_it.next()
print('Batch shape=%s, min=%.3f, max=%.3f' % (batchX.shape, batchX.min(), batchX.max()))
```

```
Found 60 images belonging to 3 classes.
Found 15 images belonging to 3 classes.
Found 15 images belonging to 4 classes.
Batch shape=(32, 256, 256, 3), min=0.000, max=255.000
```


What is and Why use Data Augmentation?

- Technique used to artificially expand size of training set
- Training on more data results in more accurate models
- Augmentation creates variations of images that can improve the ability of a model to generalize on new images
- Image augmentation in Keras is performed using `ImageDataGenerator()` class



Data Augmentation Types

1. Horizontal/Vertical shift
2. Horizontal/Vertical flip
3. Random rotation
4. Random brightness
5. Random zoom

Data Augmentation Types

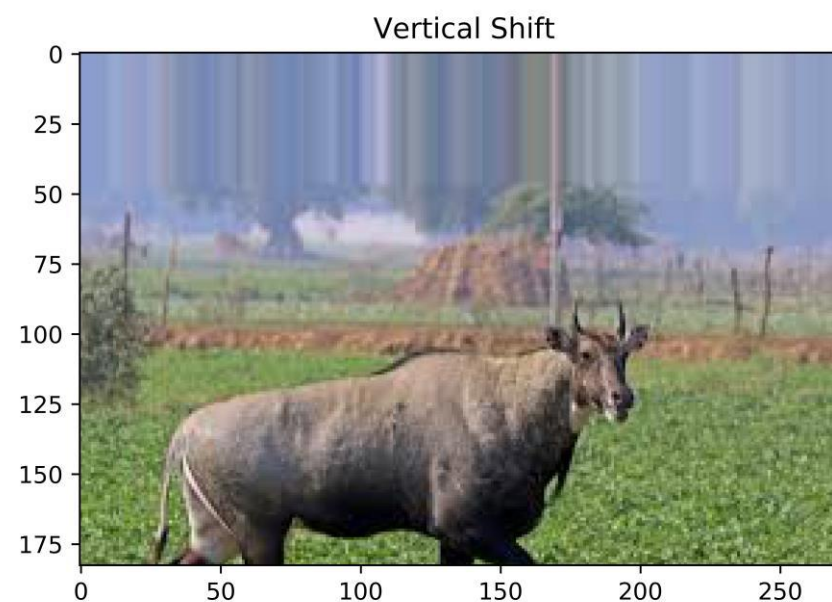
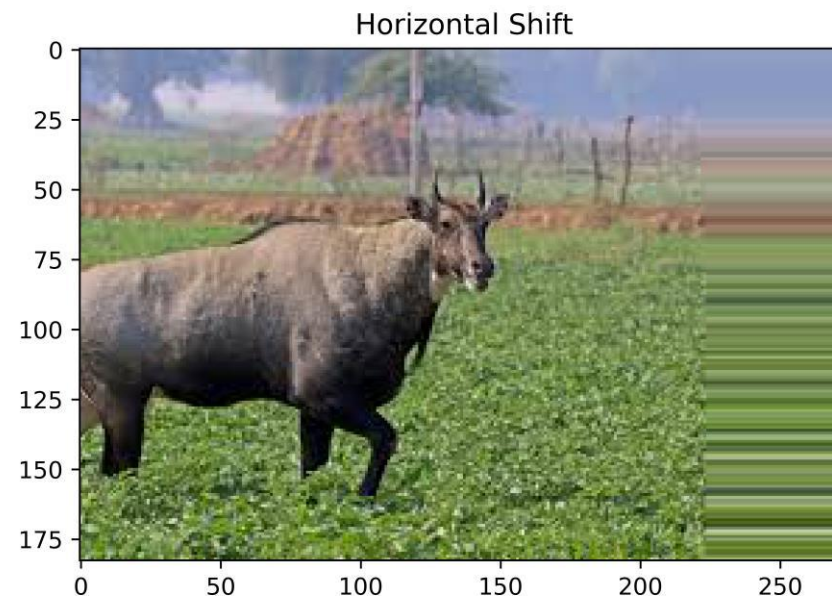
1. Horizontal/Vertical shift

datagen =
ImageDataGenerator(width_shift_range=[-50,50])

```
# Load the image
single_nilgai_path = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai\\nilgai1.png'
#img = load_img('C:\\Users\\ncu116\\Desktop\\data\\chapter_09\\bird.jpg')
save_here = 'C:\\Users\\ncu116\\Desktop\\data\\test\\test_augmented'
img = load_img(single_nilgai_path)
# convert to numpy array
data = img_to_array(img)
# expand dimension to one sample
samples = expand_dims(data, 0)
# create image data augmentation generator
datagen = ImageDataGenerator(width_shift_range=[-50,50])

# prepare iterator and save new examples to file "save here"
it = datagen.flow(samples, batch_size=1) #additional "save" argument can be used
pyplot.figure(figsize=(12, 12))
for i in range(1):
    pyplot.subplot(330 + 1 + i)
    batch = it.next()
    image = batch[0].astype('uint8')
    #fig = pyplot.figure()
    pyplot.imshow(image)
    pyplot.title('Horizontal Shift')

# show the figure
pyplot.show()
fig.savefig('hhhoz_shift.png', dpi=800)
```



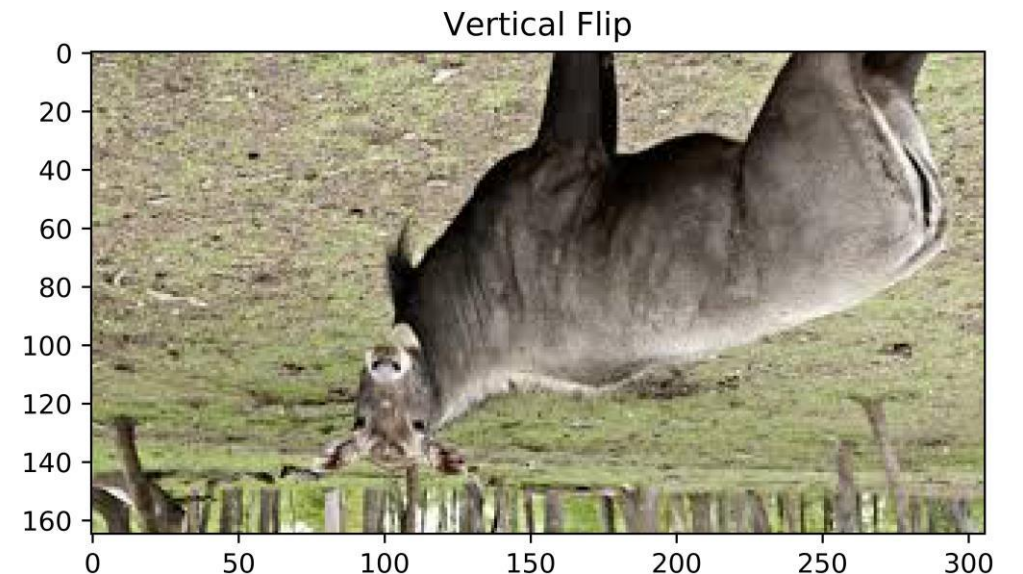
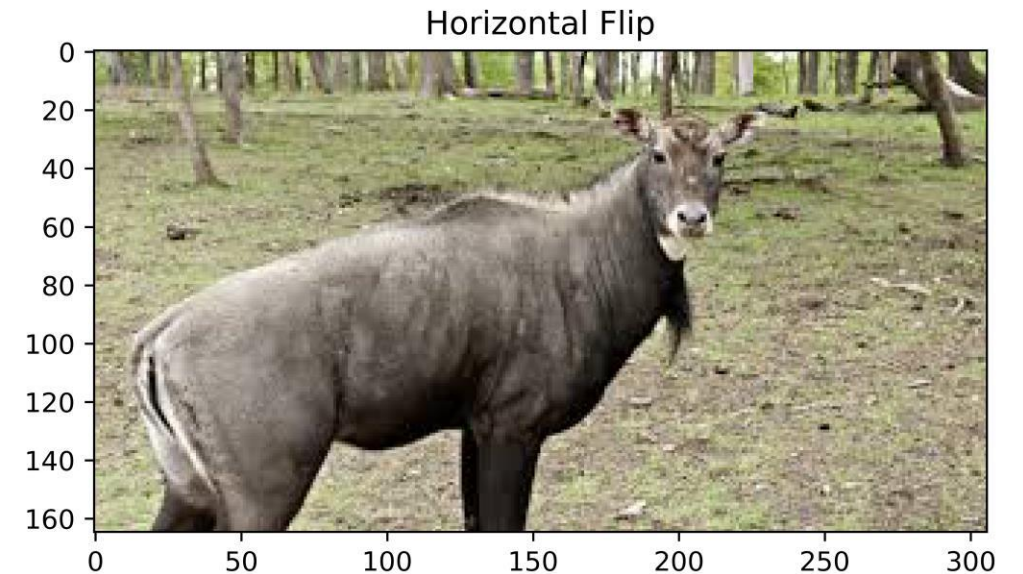
Data Augmentation Types

2. Horizontal/Vertical flip

`datagen = ImageDataGenerator(vertical_flip=True)`

```
single_nilgai_path = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai\\nilgai2.png'
#img = load_img('C:\\Users\\ncu116\\Desktop\\data\\chapter_09\\bird.jpg')
#save_here = 'C:\\Users\\ncu116\\Desktop\\data\\test\\test_augmented'
img = load_img(single_nilgai_path)
# convert to numpy array
data = img_to_array(img)
# expand dimension to one sample
samples = expand_dims(data, 0)
# create image data augmentation generator
datagen = ImageDataGenerator(vertical_flip=True)
it = datagen.flow(samples, batch_size=1)
pyplot.figure(figsize=(19, 12))

#generate samples and plot
for i in range(1):
    #define subplot
    pyplot.subplot(330 + 1 + i)
    # generate batch of images
    batch = it.next()
    # convert to unsigned integers for viewing
    image = batch[0].astype('uint8')
    fig = pyplot.figure()
    pyplot.imshow(image)
    pyplot.title('Vertical Flip')
    fig.savefig('vert_flip.png', dpi=800)
```



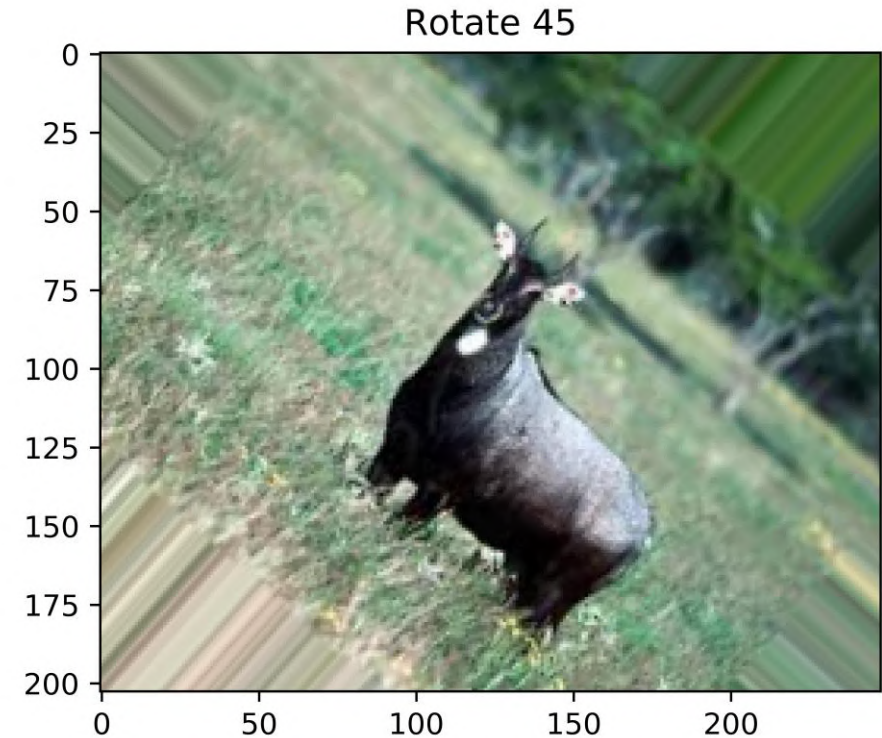
Data Augmentation Types

3. Random Rotation

```
datagen = ImageDataGenerator(rotation_range=45)
```

```
single_nilgai_path = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai\\nilgai3.png'
#save_here = 'C:\\Users\\ncu116\\Desktop\\data\\test\\test_augmented'
img = load_img(single_nilgai_path)
# convert to numpy array
data = img_to_array(img)
# expand dimension to one sample
samples = expand_dims(data, 0)
# create image data augmentation generator
datagen = ImageDataGenerator(rotation_range=45)
it = datagen.flow(samples, batch_size=1)
pyplot.figure(figsize=(19, 12))

#generate samples and plot
for i in range(1):
    #define subplot
    pyplot.subplot(330 + 1 + i)
    # generate batch of images
    batch = it.next()
    # convert to unsigned integers for viewing
    image = batch[0].astype('uint8')
    fig = pyplot.figure()
    pyplot.imshow(image)
    pyplot.title('Rotate 45')
    fig.savefig('Rotate_flip.png', dpi=800)
```

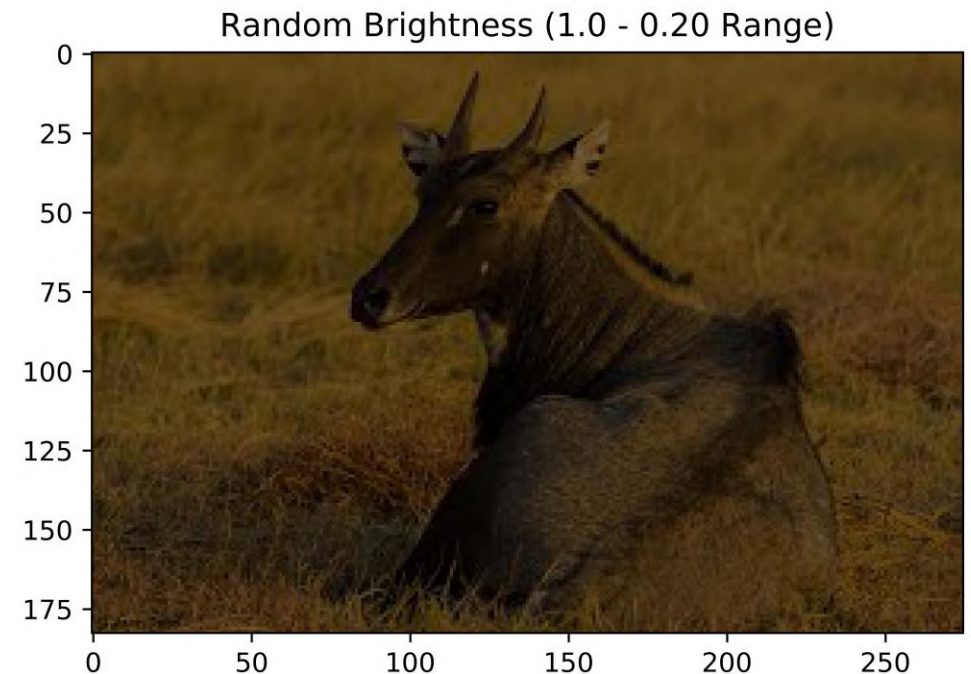


Data Augmentation Types

4. Random Brightness

```
datagen =  
ImageDataGenerator(brightness_range[0.2, 1.0])
```

```
single_nilgai_path = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai\\nilgai4.png'  
#save_here = 'C:\\Users\\ncu116\\Desktop\\data\\test\\test_augmented'  
img = load_img(single_nilgai_path)  
# convert to numpy array  
data = img_to_array(img)  
# expand dimension to one sample  
samples = expand_dims(data, 0)  
# create image data augmentation generator  
datagen = ImageDataGenerator(brightness_range=[0.2, 1.0])  
it = datagen.flow(samples, batch_size=1)  
pyplot.figure(figsize=(19, 12))  
  
#generate samples and plot  
for i in range(1):  
    #define subplot  
    pyplot.subplot(330 + 1 + i)  
    # generate batch of images  
    batch = it.next()  
    # convert to unsigned integers for viewing  
    image = batch[0].astype('uint8')  
    fig = pyplot.figure()  
    pyplot.imshow(image)  
    pyplot.title('Random Brightness (1 - 20% Change)')  
    fig.savefig('Rando_Bright.png', dpi=800)
```

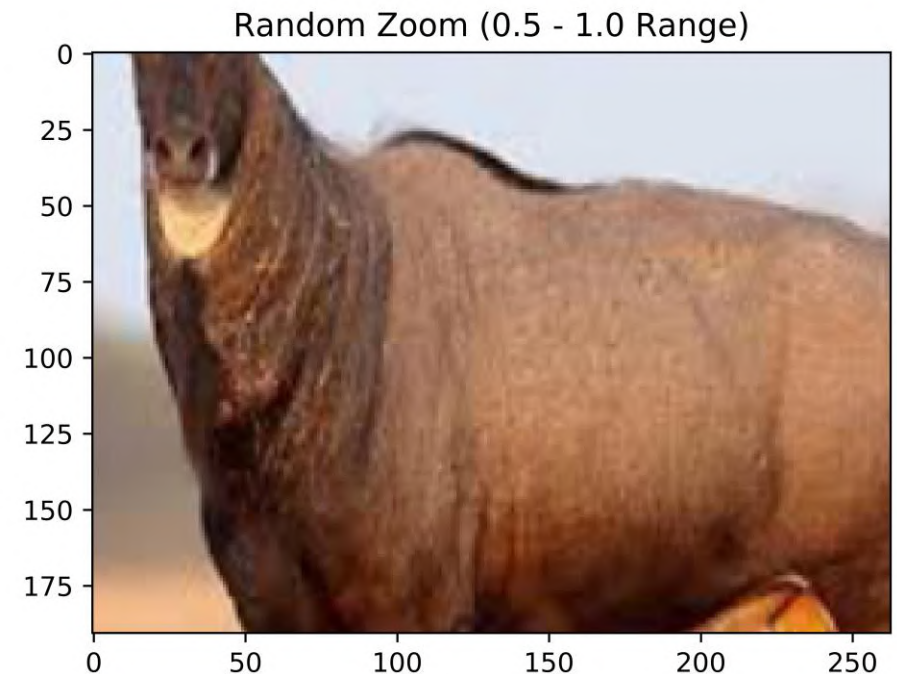


Data Augmentation Types

5. Random Zoom

```
datagen =  
ImageDataGenerator(zoom_range=[0.5, 1.0])
```

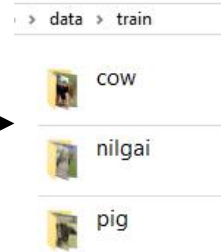
```
single_nilgai_path = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai\\nilgai5.png'  
#save_here = 'C:\\Users\\ncu116\\Desktop\\data\\test\\test_augmented'  
img = load_img(single_nilgai_path)  
# convert to numpy array  
data = img_to_array(img)  
# expand dimension to one sample  
samples = expand_dims(data, 0)  
# create image data augmentation generator  
datagen = ImageDataGenerator(zoom_range=[0.5, 1.0])  
it = datagen.flow(samples, batch_size=1)  
pyplot.figure(figsize=(19, 12))  
  
#generate samples and plot  
for i in range(1):  
    #define subplot  
    pyplot.subplot(330 + 1 + i)  
    # generate batch of images  
    batch = it.next()  
    # convert to unsigned integers for viewing  
    image = batch[0].astype('uint8')  
    fig = pyplot.figure()  
    pyplot.imshow(image)  
    pyplot.title('Random Zoom (0.5 - 1.0 Range)')  
    fig.savefig('rando_zoom.png', dpi=800)
```



Results: Data augmentation



Save location
'train'



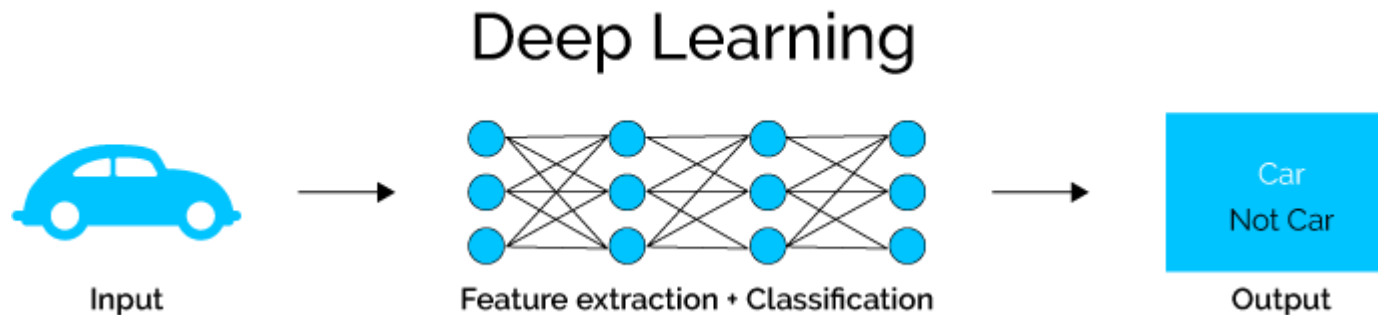
```
single_nilgai_path = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai\\nilgai1.  
#save location  
save_here = 'C:\\Users\\ncu116\\Desktop\\data\\train\\nilgai'  
img = load_img(single_nilgai_path)  
# convert to numpy array  
data = img_to_array(img)  
# expand dimension to one sample  
samples = expand_dims(data, 0)  
# create image data augmentation generator  
datagen = ImageDataGenerator(zoom_range=[0.5, 1.0])  
#batch flow from directory  
it = datagen.flow(samples, batch_size=1,  
                  save_to_dir=save_here,  
                  save_prefix='aug',  
                  save_format='png')  
  
pyplot.figure(figsize=(19, 12))  
  
#generate samples and plot  
for i in range(1):  
    #define subplot  
    pyplot.subplot(330 + 1 + i)  
    # generate batch of images  
    batch = it.next()  
    # convert to unsigned integers for viewing  
    image = batch[0].astype('uint8')  
    fig = pyplot.figure()  
    pyplot.imshow(image)  
    pyplot.title('Random Zoom (0.5 - 1.0 Range)')  
    fig.savefig('1rando_zoom.png', dpi=800)
```

'save' argument
in .flow method

$$1 + 5 \times 60 = 360 \text{ image training data saved to directory}$$

Results and Next Step:

- Generate batches of tensor image data with real-time data augmentation. The data will be looped over (in batches)
- Image now tensor object iterable by model
- Can now be applied to train classifier



Questions?

What is a validation set?

- Training set: A set of examples used for learning, that is to fit the parameters of the classifier.*
- Validation set: A set of examples used to tune the parameters of a classifier, for example to choose the number of hidden units in a neural network.*
- Test set: A set of examples used only to assess the performance of a fully-specified classifier.*

— Brian Ripley, page 354, [Pattern Recognition and Neural Networks](#), 1996

What is a tensor?

- A **tensor** is a generalization of vectors and matrices to potentially higher dimensions

“Tensors are matrices of many dimensions”

— Emmanuel Caradec, Understanding Keras Tensors, 2018

- Deep learning usually involves hundreds, if not thousands, of dimensions and fields.
- Best represented by tensors since they can represent anything ranging from *zero* to N dimensions.
- Images can easily be broken down into a few hundred features. Therefore, tensors can be best viewed as containers that wrap and store data features in the context of machine learning and deep learning.

— CE Kan, Quick ML Concepts: Tensors, 2018