

Advanced Techniques for CNNs and Keras

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- One practical obstacle to building image classifiers is obtaining labeled training data.
- Finding images is difficult.
- Labeling images is time consuming and costly.
- How can me make the most out of the labelled data we have?



• If this is a chair:



• If this is a chair...



• Then so is this!



• If this is a chair...



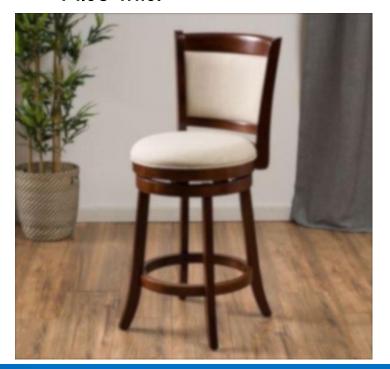
Also this:



• If this is a chair...



Also this:



- By slightly altering images, we can increase our effective data size.
- Also allows the neural network to learn invariance to certain transformations.
- But we need to be careful this can have unintended consequences.

Would not want a self-driving car to think these mean the same thing!





Data Flows in Keras

- Keras has a convenient mechanism for Data Augmentation.
- Requires use of "Data Generators"
- To date, we have used the standard model.fit mechanism
- Holds entire dataset in memory
- Reads the batches one by one out of memory

Data Flows in Keras

- Alternative mechanism is to use a "data generator"
- Idea: define a generator object which "serves" the batches of data.
- Then use model.fit with datagen.flow
- Generators can be used to serve images from disk to conserve working memory

ImageDataGenerator

- Keras has an ImageDataGenerator class which permits "realtime" data-augmentation.
- When a batch of images is served, you can specify random changes to be made to the image.
- These include shifting, rotating, flipping, and various normalizations of the pixel values.

ImageDataGenerator

```
keras.preprocessing.image.ImageDataGenerator(
   featurewise center=False, samplewise center=False,
   featurewise std normalization=False,
   samplewise std normalization=False,
   zca whitening=False,
   rotation range=0.,
   width shift range=0.,
   height shift range=0.,
   shear range=0., zoom range=0., channel shift range=0.,
   fill mode='nearest', cval=0.,
   horizontal flip=False, vertical flip=False,
   rescale=None, preprocessing function=None,
   data format=K.image data format())
```

Shifting Images

```
keras.preprocessing.image.ImageDataGenerator(
   width_shift_range=0.,
   height_shift_range=0.,
   ...)
width_shift_range=0.1
```

Shifting Images (how to fill in)

```
keras.preprocessing.image.ImageDataGenerator(
   . . . ,
  fill mode='nearest', cval=0.,
  ...)
cval used with fill mode='constant'
fill_mode: {"constant", "nearest", "reflect" or "wrap"}
   'constant': kkkkkkkk|abcd|kkkkkkkk (cval=k)
   'nearest': aaaaaaaa|abcd|dddddddd
   'reflect': abcddcba|abcd|dcbaabcd
   'wrap': abcdabcd|abcd|abcdabcd
```

Rotating Images

Flipping Images

Example

```
datagen = keras.preprocessing.image.ImageDataGenerator(
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True,
    zoom_range=0.2)

train_generator = datagen.flow(X, y, batch_size=32)

model.fit(train_generator, epochs=10)
```

