# In [1]:

```
import os
os.environ["CUDA_DEVICE_ORDER"]="PCI_BUS_ID"
os.environ["CUDA_VISIBLE_DEVICES"]="{}".format(2) # gpu idx
```

#### In [2]:

```
import warnings
warnings.filterwarnings(action='ignore')
import tensorflow as tf
from keras.models import Sequential, Model
from keras.layers import ZeroPadding3D, Input, Dense, Flatten, BatchNormalization,MaxPool3D, Activa
from keras.layers.convolutional import Conv3D
from keras import optimizers
from keras.optimizers import Adam, SGD
import keras
from keras import backend as K
import keras_video.utils
```

Using TensorFlow backend.

# In [3]:

```
import keras_video
import glob
classes = [i.split(os.path.sep)[1] for i in glob.glob('train_five/*')]
glob_pattern='train_five/{classname}/*.webm'
```

# In [4]:

```
train = keras_video.VideoFrameGenerator(classes=classes, glob_pattern=glob_pattern, nb_frames=16, ba
```

```
class attach, validation count: 100, train count: 400 class close, validation count: 100, train count: 400 class cover, validation count: 100, train count: 400 class drop, validation count: 100, train count: 400 class fold, validation count: 100, train count: 400 class hold, validation count: 100, train count: 400 class move, validation count: 100, train count: 400 class put, validation count: 100, train count: 400 class take, validation count: 100, train count: 400 class throw, validation count: 400 class throw, validation count: 400 class throw count: 400 class thro
```

#### In [5]:

```
valid = train.get_validation_generator()
```

Total data: 10 classes for 1000 files for validation

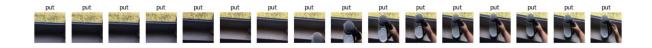
### In [6]:

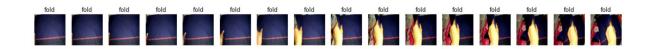
keras\_video.utils.show\_sample(train, random=True)

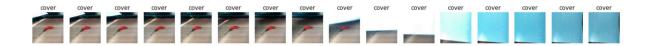


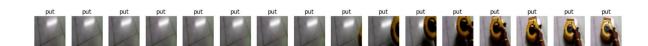




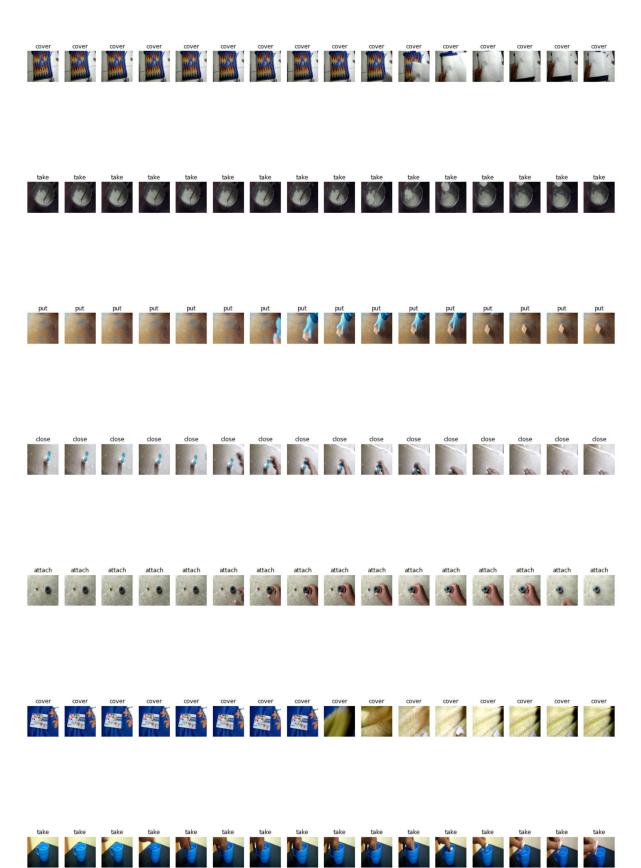




















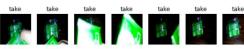


















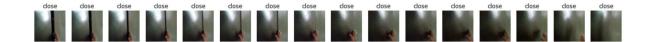












# In [ ]:

```
## input layer
input_layer = Input((16, 120, 120, 3))
## convolutional layers
conv_layer1 = Conv3D(filters=8, kernel_size=(3, 3, 3), activation='relu')(input_layer)
conv_layer2 = Conv3D(filters=16, kernel_size=(3, 3, 3), activation='relu')(conv_layer1)
## add max pooling to obtain the most imformatic features
pooling_layer1 = MaxPool3D(pool_size=(2, 2, 2))(conv_layer2)
conv_layer3 = Conv3D(filters=32, kernel_size=(3, 3, 3), activation='relu')(pooling_layer1)
conv_layer4 = Conv3D(filters=64, kernel_size=(3, 3, 3), activation='relu')(conv_layer3)
pooling_layer2 = MaxPool3D(pool_size=(2, 2, 2))(conv_layer4)
## perform batch normalization on the convolution outputs before feeding it to MLP architecture
pooling layer2 = BatchNormalization()(pooling layer2)
flatten_layer = Flatten()(pooling_layer2)
## create an MLP architecture with dense layers : 4096 -> 512 -> 10
## add dropouts to avoid overfitting / perform regularization
dense_layer1 = Dense(units=2048, activation='relu')(flatten_layer)
dense_layer1 = Dropout(0.4)(dense_layer1)
dense_layer2 = Dense(units=512, activation='relu')(dense_layer1)
dense_layer2 = Dropout(0.4)(dense_layer2)
output_layer = Dense(units=10, activation='softmax')(dense_layer2)
## define the model with input layer and output layer
model = Model(inputs=input_layer, outputs=output_layer)
```

#### In [8]:

```
## input layer
input_layer = Input((16, 112, 112, 3))
## convolutional layers
conv_layer1 = Conv3D(filters=64, kernel_size=(3, 3, 3), activation='relu', padding='same', strides=
pooling_layer1 = MaxPool3D(pool_size=(1, 2, 2), strides=(1, 2, 2), padding='valid')(conv_layer1)
conv_layer2 = Conv3D(filters=128, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
pooling_layer2 = MaxPool3D(pool_size=(2, 2, 2), strides=(2, 2, 2), padding='valid')(conv_layer2)
## add max pooling to obtain the most imformatic features
conv_layer3 = Conv3D(filters=256, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
conv_layer4 = Conv3D(filters=256, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
pooling_layer3 = MaxPool3D(pool_size=(2, 2, 2), strides=(2, 2, 2), border_mode='valid')(conv_layer4|
conv_layer5 = Conv3D(filters=512, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
conv_layer6 = Conv3D(filters=512, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
pooling_layer4 = MaxPool3D(pool_size=(2, 2, 2), strides=(2, 2, 2), padding='valid')(conv_layer6)
conv_layer7 = Conv3D(filters=512, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
conv_layer8 = Conv3D(filters=512, kernel_size=(3, 3, 3), activation='relu', padding='same', strides
pooling_layer5 = ZeroPadding3D(padding=(0, 1, 1))(conv_layer8)
## perform batch normalization on the convolution outputs before feeding it to MLP architecture
pooling_layer6 = MaxPool3D(pool_size=(2, 2, 2), strides=(2, 2, 2), padding='valid')(pooling_layer5)
flatten_layer = Flatten()(pooling_layer6)
## create an MLP architecture with dense layers : 4096 -> 512 -> 10
## add dropouts to avoid overfitting / perform regularization
dense_layer1 = Dense(units=4096, activation='relu')(flatten_layer)
dense_layer1 = Dropout(0.5)(dense_layer1)
dense_layer2 = Dense(units=4096, activation='relu')(dense_layer1)
dense_layer2 = Dropout(0.5)(dense_layer2)
output_layer = Dense(units=10, activation='softmax')(dense_layer2)
## define the model with input layer and output layer
model = Model(inputs=input_layer, outputs=output_layer)
```

# In [9]:

```
optimizer = Adam(learning_rate=1e-5)
#optimizer = SGD(learning_rate=1e-6)
model.compile(optimizer=optimizer, loss='categorical_crossentropy', metrics=['accuracy'])
print (model.summary())
```

Model: "model\_1"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	(None, 16, 112, 112, 3)	0
conv3d_1 (Conv3D)	(None, 16, 112, 112, 64)	5248
max_pooling3d_1 (MaxPooling3	(None, 16, 56, 56, 64)	0
conv3d_2 (Conv3D)	(None, 16, 56, 56, 128)	221312
max_pooling3d_2 (MaxPooling3	(None, 8, 28, 28, 128)	0
conv3d_3 (Conv3D)	(None, 8, 28, 28, 256)	884992
conv3d_4 (Conv3D)	(None, 8, 28, 28, 256)	1769728
max_pooling3d_3 (MaxPooling3	(None, 4, 14, 14, 256)	0
conv3d_5 (Conv3D)	(None, 4, 14, 14, 512)	3539456
conv3d_6 (Conv3D)	(None, 4, 14, 14, 512)	7078400
max_pooling3d_4 (MaxPooling3	(None, 2, 7, 7, 512)	0
conv3d_7 (Conv3D)	(None, 2, 7, 7, 512)	7078400
conv3d_8 (Conv3D)	(None, 2, 7, 7, 512)	7078400
zero_padding3d_1 (ZeroPaddin	(None, 2, 9, 9, 512)	0
max_pooling3d_5 (MaxPooling3	(None, 1, 4, 4, 512)	0
flatten_1 (Flatten)	(None, 8192)	0
dense_1 (Dense)	(None, 4096)	33558528
dropout_1 (Dropout)	(None, 4096)	0
dense_2 (Dense)	(None, 4096)	16781312
dropout_2 (Dropout)	(None, 4096)	0
dense_3 (Dense)	(None, 10)	40970
T-1-1 70 000 740	<del>_</del>	

Total params: 78,036,746 Trainable params: 78,036,746 Non-trainable params: 0

None

### In [10]:

```
EPOCHS=30
callbacks = [
   #keras.callbacks.ReduceLROnPlateau(verbose=1),#측정 항목이 향상되지 않는 경우 learning rate 줄임
   keras.callbacks.ModelCheckpoint(
       '/home/jiho/work/repos/somth2smth_conv3d/model/weights.{epoch:02d}-{val_loss:.2f}.hdf5',
       verbose=1),
   keras.callbacks.TensorBoard(log_dir='/home/jiho/work/repos/somth2smth_conv3d')#그래프로 보기위해
model.fit_generator(
   train,
   validation_data=valid,
   verbose=1,
   epochs=EPOCHS,
   callbacks=callbacks
)
200/200 [============= ] - 149s 743ms/step - loss: 1.1376 - accur
acy: 0.6143 - val_loss: 1.0841 - val_accuracy: 0.4880
Epoch 00022: saving model to /home/jiho/work/repos/somth2smth_conv3d/model/weight
s.22-1.08.hdf5
Epoch 23/30
                         =======] - 148s 740ms/step - loss: 1.0392 - accur
200/200 [========
acy: 0.6435 - val_loss: 1.9013 - val_accuracy: 0.4930
Epoch 00023: saving model to /home/jiho/work/repos/somth2smth_conv3d/model/weight
s.23-1.90.hdf5
Epoch 24/30
200/200 [============ ] - 148s 741ms/step - loss: 0.9965 - accur
acy: 0.6597 - val_loss: 1.4739 - val_accuracy: 0.5090
Epoch 00024: saving model to /home/jiho/work/repos/somth2smth_conv3d/model/weight
s.24-1.47.hdf5
Epoch 25/30
200/200 [=============] - 148s 742ms/step - loss: 0.9279 - accur
acv: 0 6845 - val loss: 1 1828 - val accuracy: 0 5050
In [ ]:
```