Reusing trained models with TensorFlow

Alejandro Solano - PyCon PL 2017



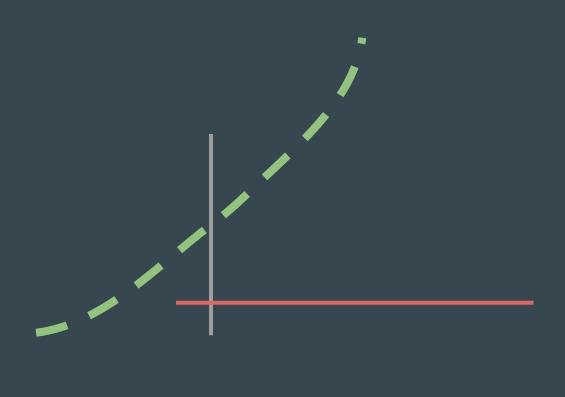


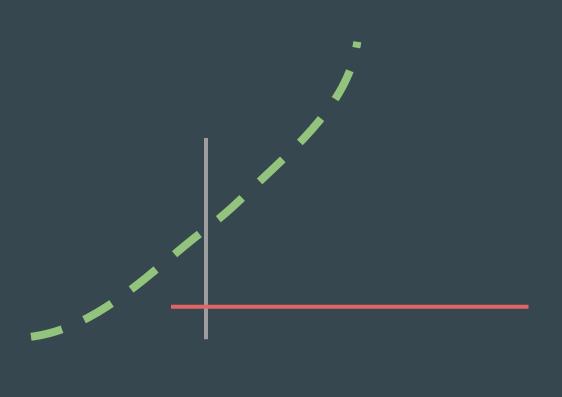
save

save load

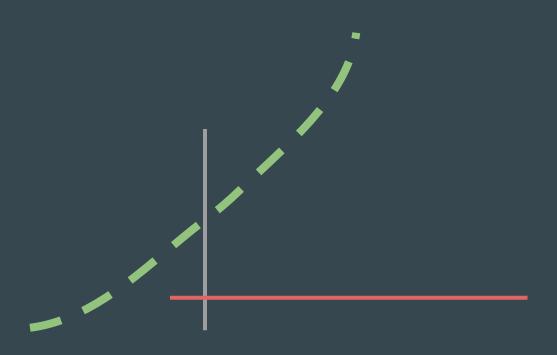
save load run



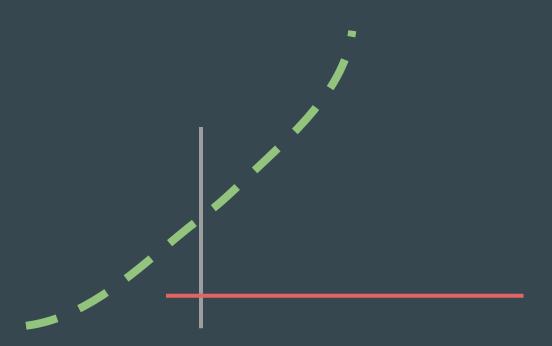




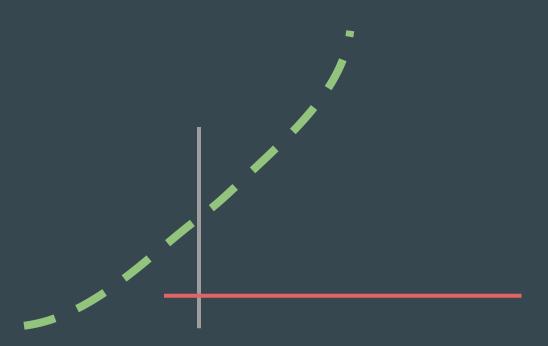




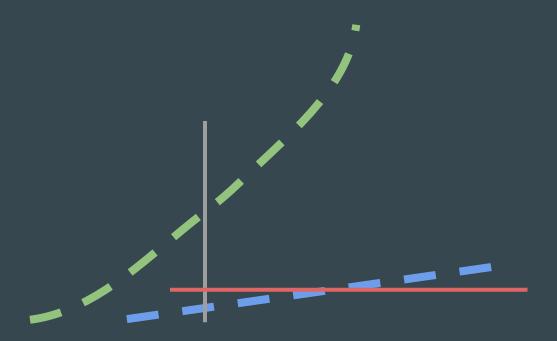




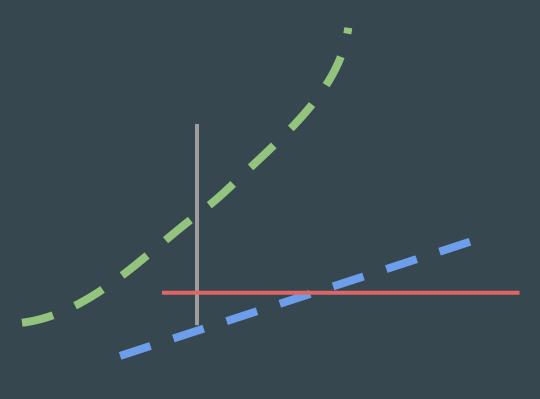
X·W +



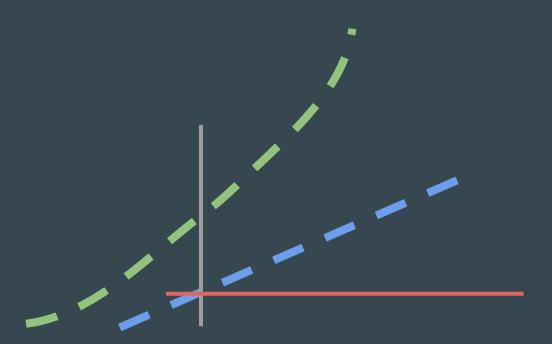
x·W + b



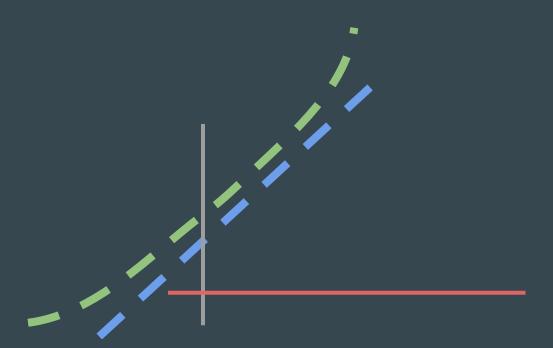
 $\times \cdot 0.1 + -0.1$



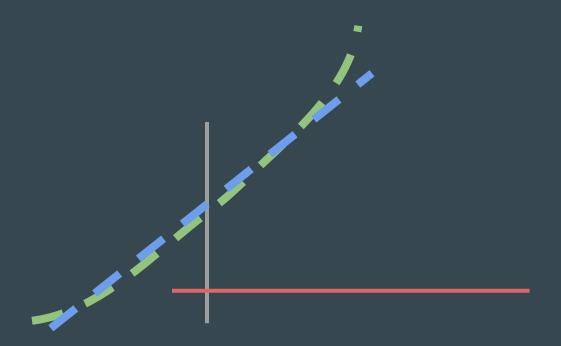
$\times \cdot 0.3 + -0.2$



 $\times \cdot 0.4 + 0.1$



$\times 1.2 + 0.8$

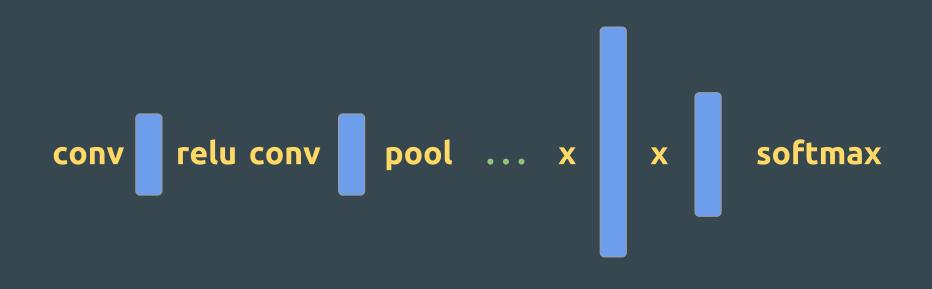


 $\times \cdot 0.95 + 1.05$

-0.95 + 1.05

static





dynamic

static





Checkpoints

Checkpoints - Save

```
##### GRAPH #####
saver = tf.train.Saver()
##### SESSION #####
saver.save(sess, 'model.ckpt')
```

Checkpoints - Save

```
##### GRAPH #####
          saver = tf.train.Saver()
          ##### SESSION #####
          saver.save(sess, 'model.ckpt')
                     .ckpt.index
                                           .ckpt.data
.ckpt.meta
```

Checkpoints - Save

```
##### GRAPH #####
          saver = tf.train.Saver()
          ##### SESSION #####
          saver.save(sess, 'model.ckpt')
.ckpt.meta
                                  .ckpt
```

Checkpoints - Load

.ckpt.meta

.ckpt

Checkpoints - Load

.ckpt.meta

.ckpt

```
##### GRAPH ####
saver = tf.train.import_meta_graph('model.ckpt.meta')
##### SESSION ####
saver.restore(sess, 'model.ckpt')
```

```
##### SESSION #####
with tf.Session() as sess:
    preds = sess.run(output_tensor, feed_dict={
        input_placeholder: image
    })
```

```
##### SESSION #####
with tf.Session() as sess:
    preds = sess.run(output_tensor, feed_dict={
        input_placeholder: image
    })
```

we need access to those tensors!

```
input_placeholder =
tf.get_default_graph().get_tensor_by_name('inputs:0')

output_tensor =
tf.get_default_graph().get_tensor_by_name('probs:0')
```

```
input_placeholder =
tf.get_default_graph().get_tensor_by_name('inputs:0')
output_tensor =
tf.get_default_graph().get_tensor_by_name('probs:0')
```

tensorboard

https://www.youtube.com/watch?v=eBbEDRsCmv4

Checkpoints - Summary

Save	Files	Load
tf.train.Saver -> .save()	.ckpt.meta	saver = tf.train.import_meta_graph()
	.ckpt	saver.restore()

Checkpoints

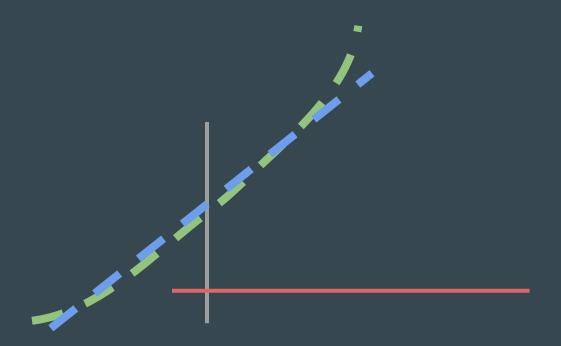
Pros:

- Checkpoints keep the variables dynamic, so we can **retrain** the model after loading it.
- Easy to save and load.

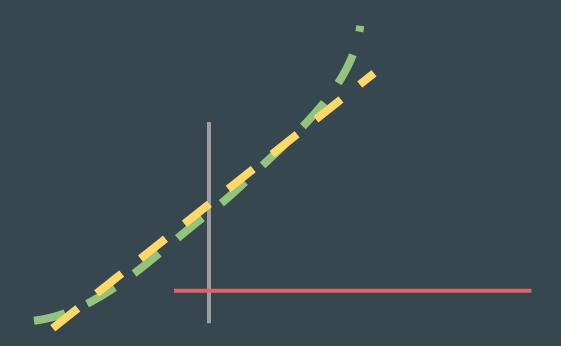
Cons:

Soooo slow.

Frozen graph



 $\times \cdot 0.95 + 1.05$



 $\times \cdot 0.95 + 1.05$









```
##### GRAPH #####
saver = tf.train.Saver()
##### SESSION #####
saver.save(...)
```

```
##### GRAPH #####
saver = tf.train.Saver()
##### SESSION #####
saver.save(...)
```

```
##### GRAPH ####
saver = tf.train.Saver()

##### SESSION #####
saver.save(...)

tf.train.write_graph(...)
```

```
##### GRAPH #####
saver = tf.train.Saver()
                                                 .ckpt
##### SESSION #####
saver.save(...)
tf.train.write_graph(...)
                                                  .pb
```

Frozen graph - Freeze

graph.pb

.ckpt

Frozen graph - Freeze

```
graph.pb

ckpt

from tensorflow.python.tools import freeze_graph

freeze_graph.freeze_graph('graph.pb', ..., 'model.ckpt', ...)
```

Frozen graph - Freeze

```
graph.pb
                                      .ckpt
from tensorflow.python.tools import freeze_graph
freeze_graph.freeze_graph('graph.pb', ..., 'model.ckpt', ...)
                frozen_graph.pb
```

frozen_graph.pb

frozen_graph.pb

```
tf.import_graph_def(graph_def, name='')
```

frozen_graph.pb

```
from tensorflow.core.framework import graph_pb2
graph_def = graph_pb2.GraphDef()
```

```
tf.import_graph_def(graph_def, name='')
```



```
from tensorflow.core.framework import graph_pb2
graph_def = graph_pb2.GraphDef()
with open('frozen_graph.pb', 'rb') as f:
    graph_def.ParseFromString(f.read())
tf.import_graph_def(graph_def, name='')
```



Frozen Graph - Summary

Save	Files	Freeze	File	Load
tf.train.write_graph()	.pb	freeze_ graph()	.pb	graph_def.ParseFromString() tf.import_graph_def(graph_def)
tf.train.Saver -> .save()	.ckpt			

Compiled

Compiled

Pros:

- Very fast.
- Just 60 Mb (executable) + 17 Mb (frozen model) with no dependencies.

Cons:

Need to be built in C++ with Bazel...

demo

Retraining using checkpoints

Fine-tuning

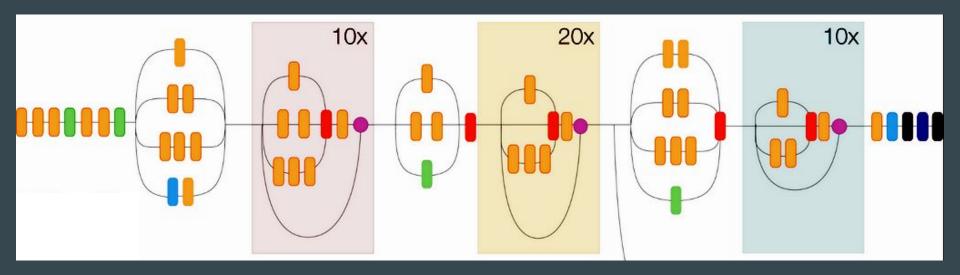
```
targets = tf.placeholder(tf.float32, shape=[None, n_classes])
# define a cost function and optimizer for these targets
with tf.Session() as sess:
    saver.restore(sess, 'model.ckpt')
    sess.run(optimizer, feed dict={
        inputs: inputs,
        targets: targets
```

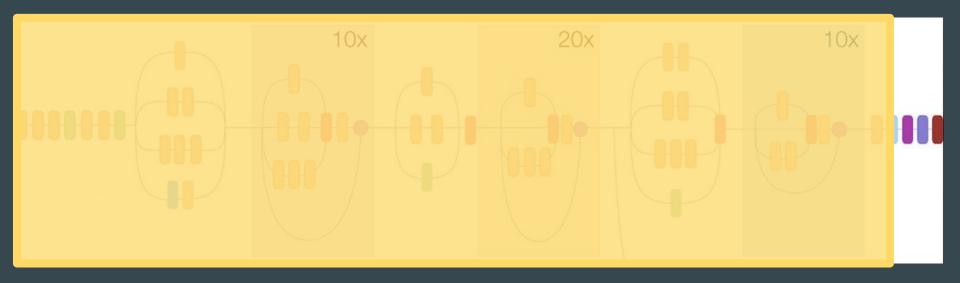
Continue training

```
optimizer =
tf.get default graph().get tensor by name('name:0')
# same for inputs and targets
with tf.Session() as sess:
    saver.restore(sess, 'model.ckpt')
    sess.run(optimizer, feed dict={
        inputs: inputs,
        targets: targets
```

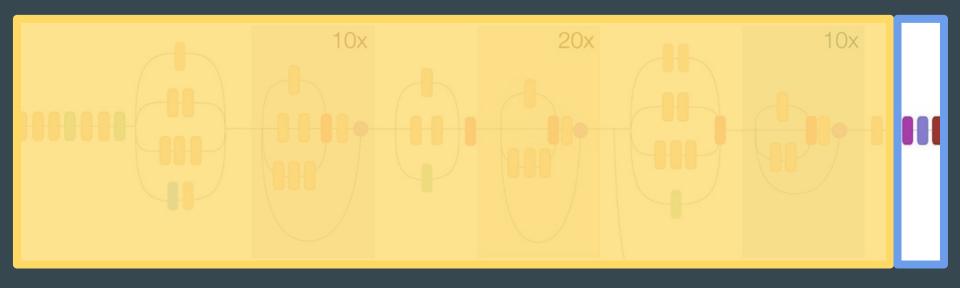






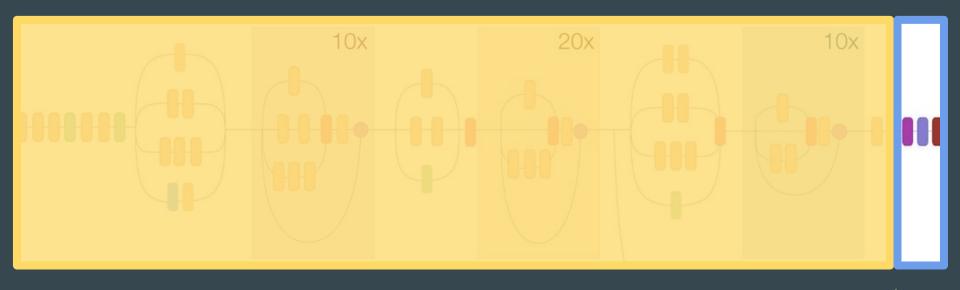














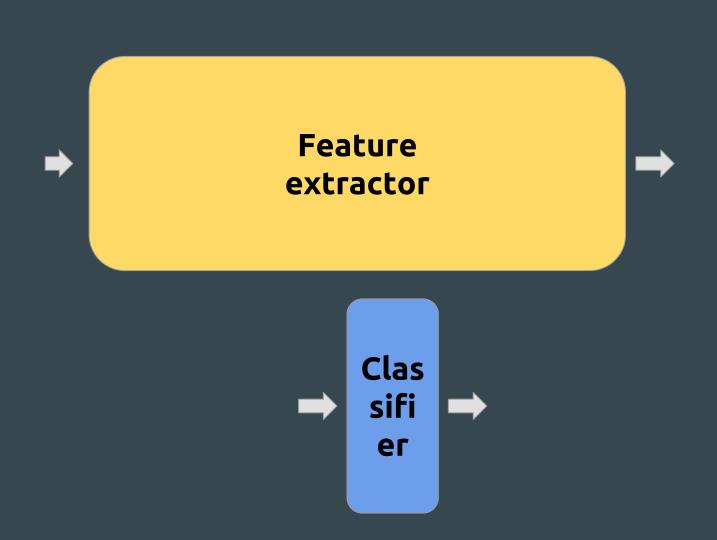


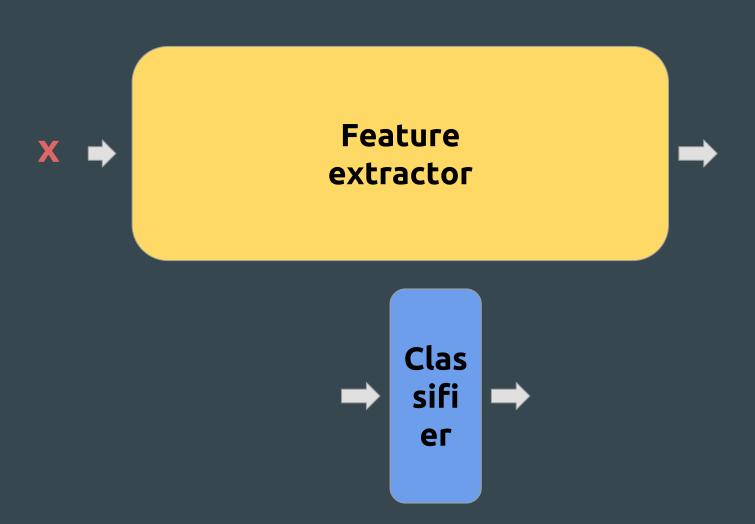


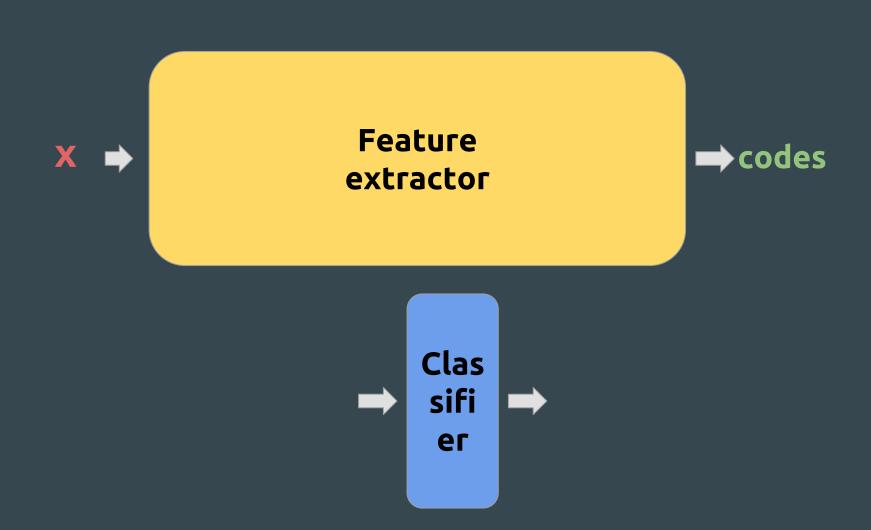


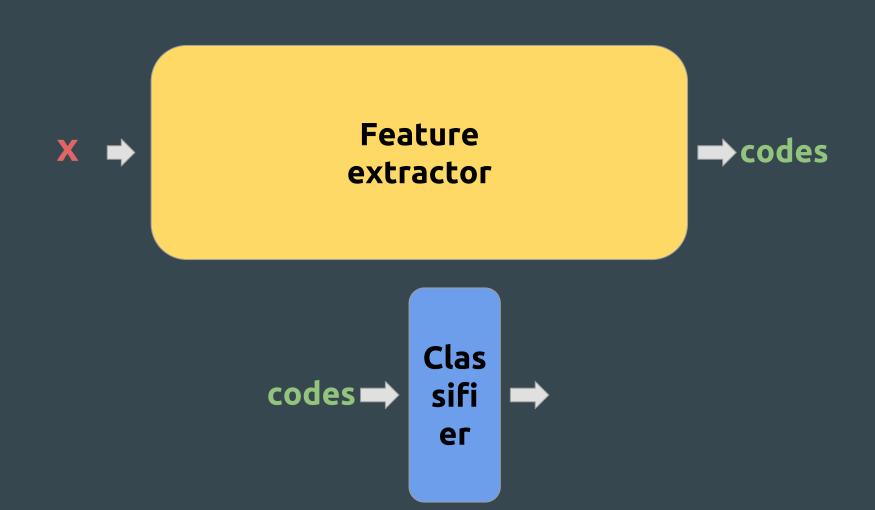


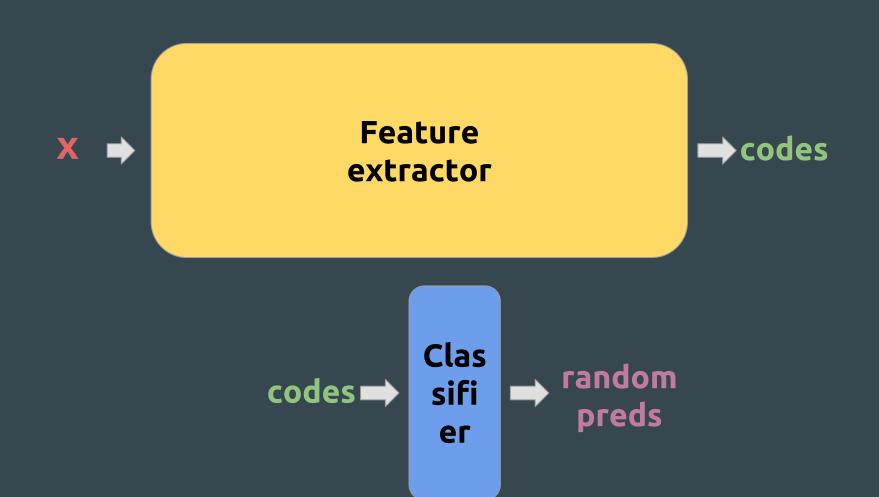


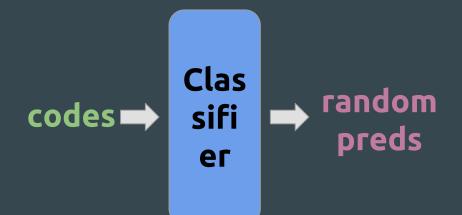


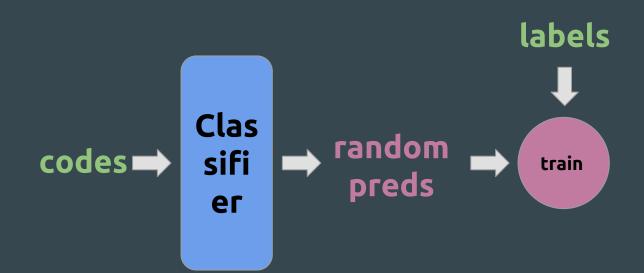


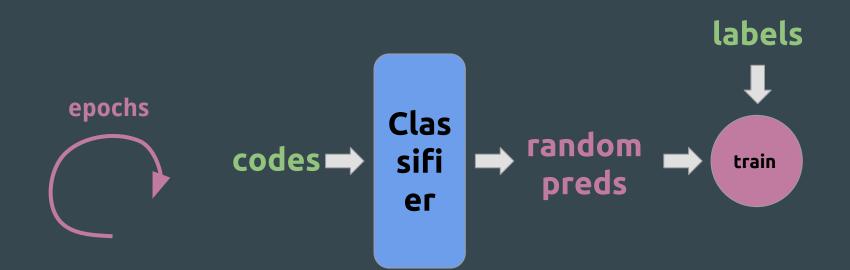












Extract features

```
with tf.Session() as sess:
    code = sess.run(bottleneck, feed_dict={
        input_placeholder: image
    })
```

Extract features

```
bottleneck =
tf.get_default_graph().get_tensor_by_name('name:0')
with tf.Session() as sess:
   code = sess.run(bottleneck, feed_dict={
        input_placeholder: image
     })
```

Extract features

```
bottleneck =
tf.get_default_graph().get_tensor_by_name('name:0'
with tf.Session() as sess:
   code = sess.run(bottleneck, feed_dict={
      input_placeholder: image
```

Retrain classifier

```
with tf.Session() as sess:
    for epoch in range(len(epochs)):
        sess.run(optimizer, feed_dict={
            inputs_: codes
            lables_: labels
       })
```

hopefully, demo





alesolano/imagenet_models_flask



alesolano/transfer_learning_webapp



alesolano/imagenet_models_flask



alesolano/transfer_learning_webapp

Thank you!