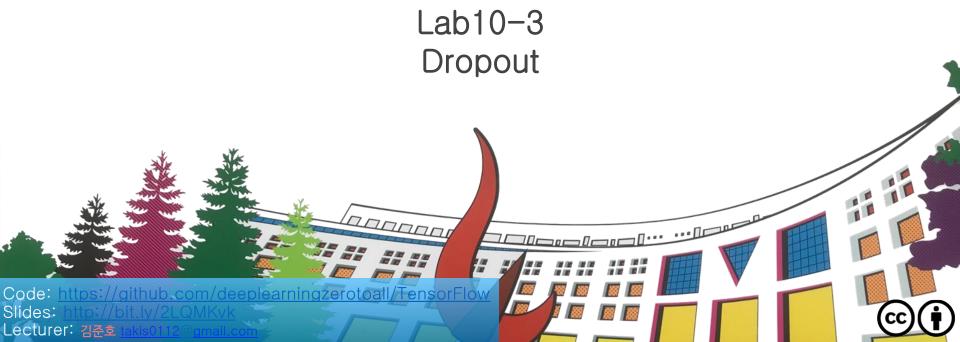
# ML/DL for Everyone Season2





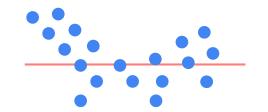
### Lab10-3: Dropout

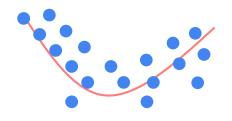
- Dropout
- Code
  - O load dataset
  - create network
  - O define loss function
  - O experiments
    - parameters
    - model
    - eager mode
- What's Next

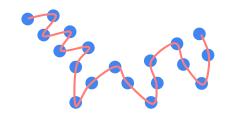


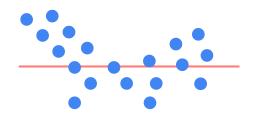




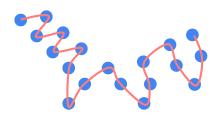








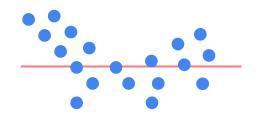


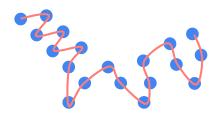


**Under-fitting** 

Good

Over-fitting





**Under-fitting** 

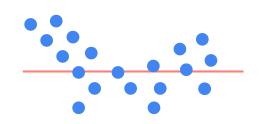
Train ↓

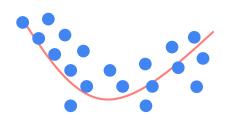
Good

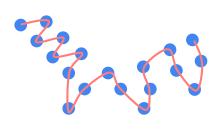
Train ≈

Over-fitting

Train 1







**Under-fitting** 

Good

Over-fitting

Train ↓

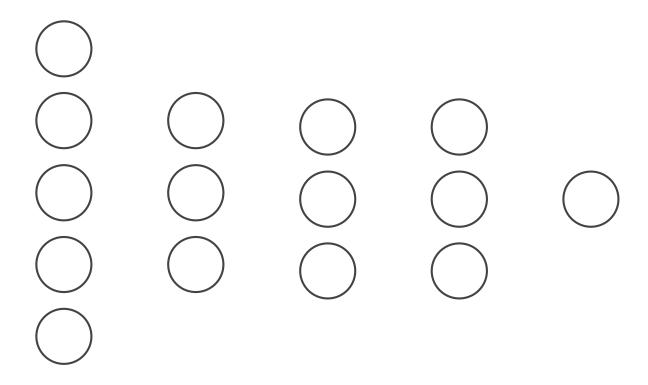
Train ≈

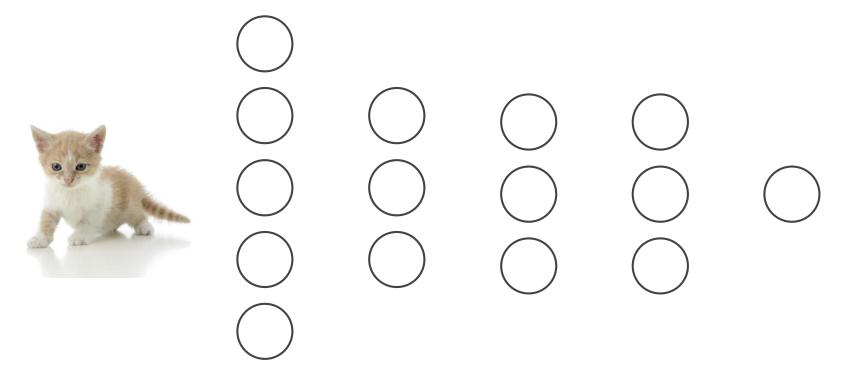
Train ↑

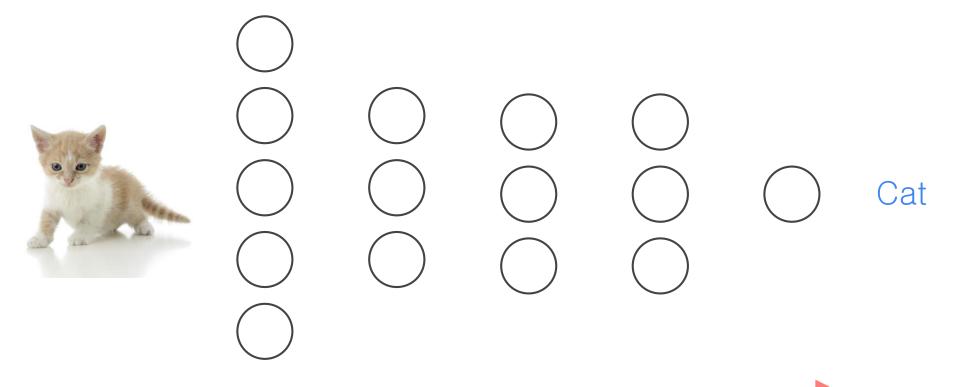
Test ↓

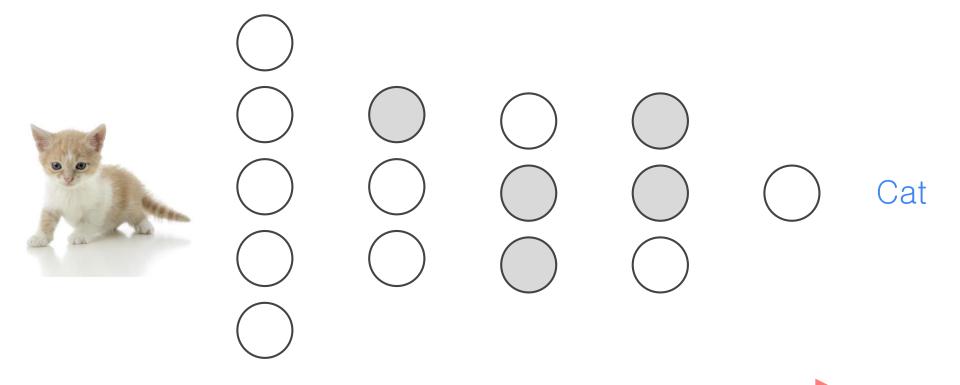
Test 1

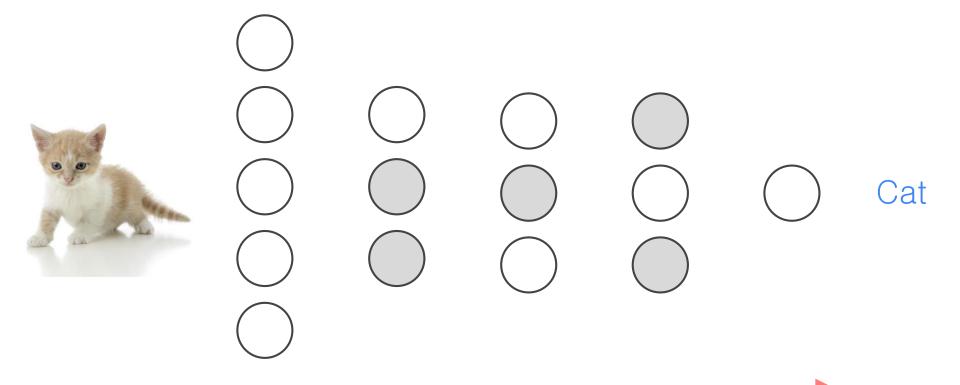
Test ↓

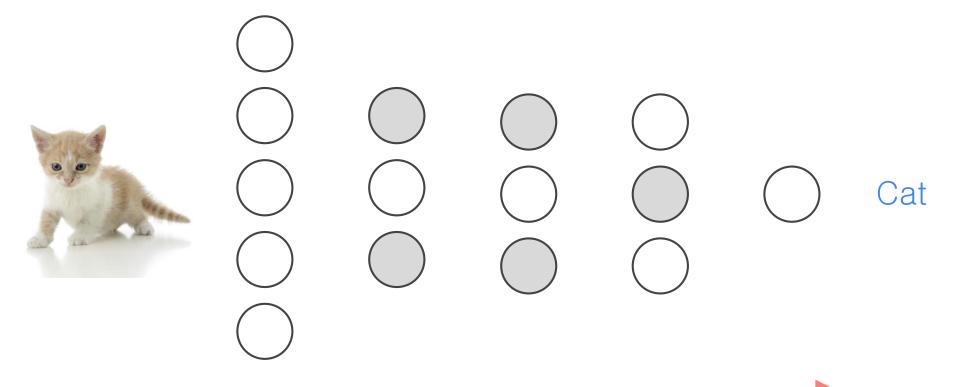


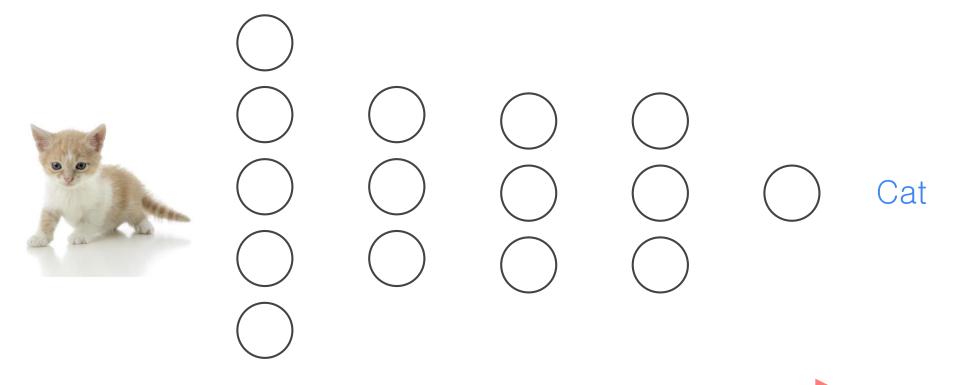


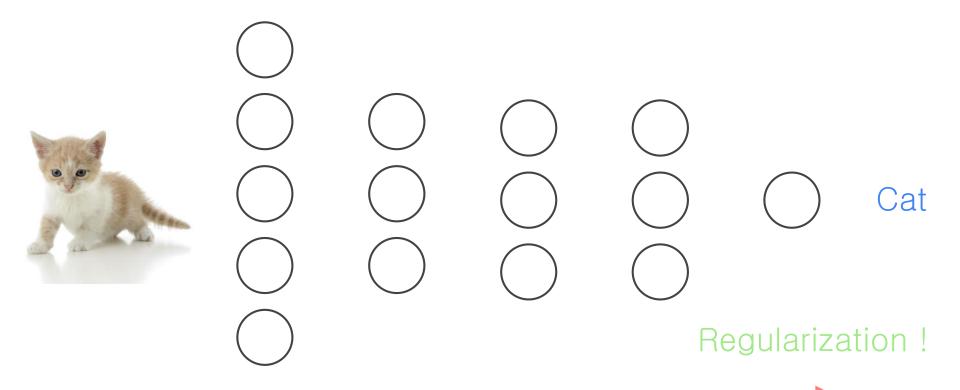












## Code

### Load mnist

#### Load mnist

```
import tensorflow as tf
import numpy as np
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.datasets import mnist # fasion mnist, cifar10, cifar100
tf.enable eager execution()
def load mnist() :
    (train data, train labels), (test data, test labels) = mnist.load data()
    train_data = np.expand_dims(train data, axis=-1) # \lceil N, 28, 28\rceil -> \lceil N, 28, 28, 1\rceil
    test data = np.expand dims(test data, axis=-1) # [N, 28, 28] \rightarrow [N, 28, 28, 1]
    train_data, test_data = normalize(train data, test data) # [0 ~ 255] -> [0 ~ 1]
    train labels = to categorical(train labels, 10) # [N,] -> [N, 10]
    test labels = to categorical(test labels, 10) # [N,] -> [N, 10]
    return train data, train labels, test data, test labels
  def normalize(train data, test data):
      train data = train data.astype(np.float32) / 255.0
      test data = test data.astype(np.float32) / 255.0
      return train data, test data
```

```
def flatten() :
    return tf.keras.layers.Flatten()

def dense(channel, weight_init) :
    return tf.keras.layers.Dense(units=channel, use_bias=True, kernel_initializer=weight_init)

def relu() :
    return tf.keras.layers.Activation(tf.keras.activations.relu)
```

```
def flatten() :
    return tf.keras.layers.Flatten()

def dense(channel, weight_init) :
    return tf.keras.layers.Dense(units=channel, use_bias=True, kernel_initializer=weight_init)

def relu() :
    return tf.keras.layers.Activation(tf.keras.activations.relu)

def dropout(rate) :
    return tf.keras.layers.Dropout(rate);
```

```
class create_model(tf.keras.Model):
    def init (self, label dim):
        super(create model, self). init ()
       weight init = tf.keras.initializers.glorot uniform()
        self.model = tf.keras.Sequential()
        self.model.add(flatten()) # [N, 28, 28, 1] -> [N, 784]
       for i in range(2):
           # [N, 784] -> [N, 256] -> [N, 256]
           self.model.add(dense(256, weight init))
           self.model.add(relu())
        self.model.add(dense(label dim, weight init)) # [N, 256] -> [N, 10]
    def call(self, x, training=None, mask=None):
       x = self.model(x)
        return x
```

```
class create model(tf.keras.Model):
    def init (self, label dim):
        super(create model, self). init ()
       weight init = tf.keras.initializers.glorot uniform()
        self.model = tf.keras.Sequential()
        self.model.add(flatten()) # [N, 28, 28, 1] -> [N, 784]
       for i in range(2):
           # [N, 784] -> [N, 256] -> [N, 256]
           self.model.add(dense(256, weight init))
           self.model.add(relu())
           self.model.add(dropout(rate=0.5))
        self.model.add(dense(label dim, weight init)) # [N, 256] -> [N, 10]
    def call(self, x, training=None, mask=None):
        x = self.model(x)
        return x
```

```
def create model(label dim) :
    weight init = tf.keras.initializers.glorot uniform()
    model = tf.keras.Sequential()
    model.add(flatten())
    for i in range(2) :
        model.add(dense(256, weight_init))
        model.add(relu())
        model.add(dropout(rate=0.5))
    model.add(dense(label dim, weight init))
    return model
```

### Define loss

#### Define loss

```
def loss fn(model, images, labels):
    logits = model(images, training=True)
    loss = tf.reduce mean(tf.nn.softmax_cross_entropy_with_logits_v2(logits=logits, labels=labels))
    return loss
def accuracy fn(model, images, labels):
    logits = model(images, training=False)
    prediction = tf.equal(tf.argmax(logits, -1), tf.argmax(labels, -1))
    accuracy = tf.reduce mean(tf.cast(prediction, tf.float32))
    return accuracy
def grad(model, images, labels):
   with tf.GradientTape() as tape:
        loss = loss fn(model, images, labels)
    return tape.gradient(loss, model.variables)
```

## Experiments (parameters)

## Experiments (parameters)

```
""" parameters """
                                                       learning rate = 0.001
                                                       batch size = 128
""" dataset """
                                                       training epochs = 1
train x, train y, test x, test y = load mnist()
                                                       training iterations = len(train x) // batch size
                                                       label dim = 10
""" Graph Input using Dataset API """
train dataset = tf.data.Dataset.from_tensor_slices((train_x, train_y)).\
    shuffle(buffer size=100000).\
    prefetch(buffer size=batch size).\
    batch(batch size).\
    repeat()
test dataset = tf.data.Dataset.from tensor slices((test x, test y)).\
    shuffle(buffer size=100000).\
    prefetch(buffer size=len(test x)).\
    batch(len(test x)).\
    repeat()
```

# Experiments (model)

### Experiments (model)

```
Dataset Iterator """
train_iterator = train_dataset.make_one_shot iterator()
test iterator = test dataset.make one shot iterator()
""" Model """
network = create_model(label_dim)
    Training """
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate)
```

## Experiments (Eager mode)

### Experiments (Eager mode)

```
checkpoint = tf.train.Checkpoint(dnn=network)
global step = tf.train.create global step()
for epoch in range(start epoch, training epochs):
    for idx in range(start iteration, training iterations):
       train input, train label = train iterator.get next()
        grads = grad(network, train input, train label)
        optimizer.apply gradients(grads and vars=zip(grads, network.variables), global step=global step)
       train loss = loss fn(network, train input, train label)
       train accuracy = accuracy fn(network, train input, train label)
       test input, test label = test iterator.get next()
       test accuracy = accuracy fn(network, test input, test label)
        print("Epoch: [%2d] [%5d/%5d], train_loss: %.8f, train_accuracy: %.4f, test_Accuracy: %.4f" \
           % (epoch, idx, training iterations, train loss, train accuracy, test accuracy))
        counter += 1
checkpoint.save(file prefix=checkpoint prefix + '-{}'.format(counter))
```

### What's Next?

- Normalization
  - Batch Normalization