

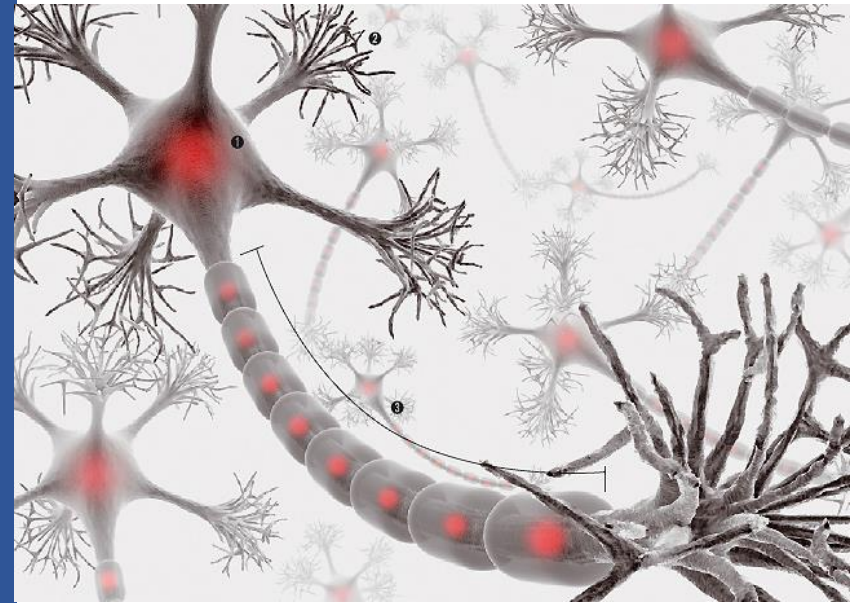
# (DNN) Spiral Data Classification

## 학습 목표

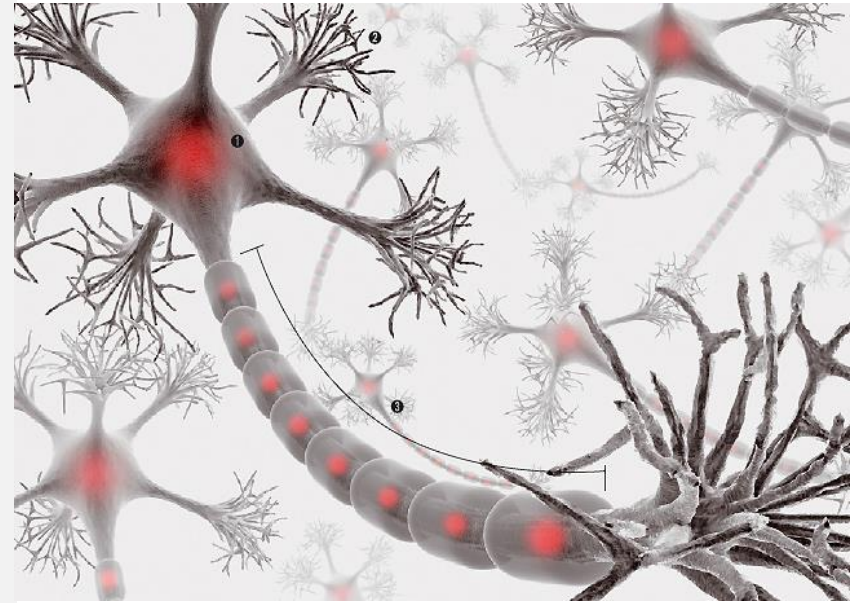
- Spiral 데이터를 분류하는 신경망 모델을 Keras로 만들어 본다.

## 주요 내용

- 1. 문제 정의
- 2. 데이터 준비
- 3. 모델 정의 및 훈련, 검증



# 1 문제 정의

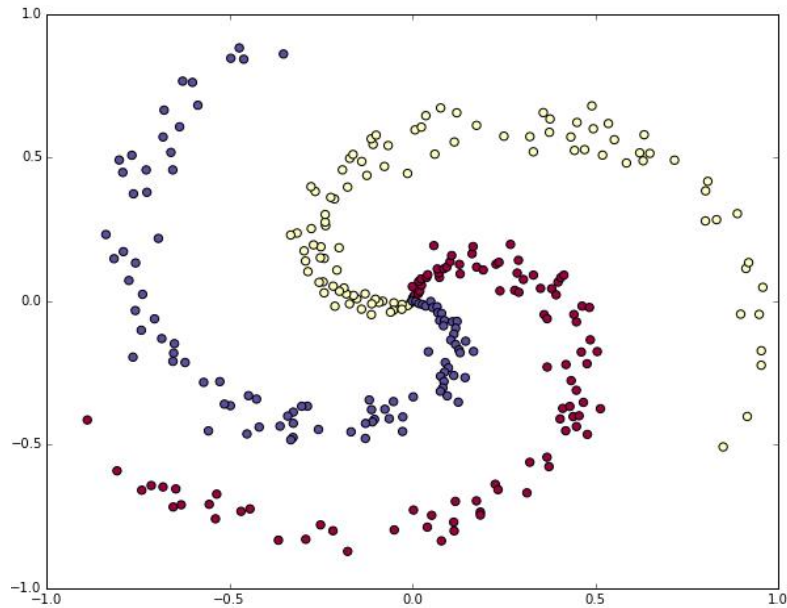


# Classification 문제

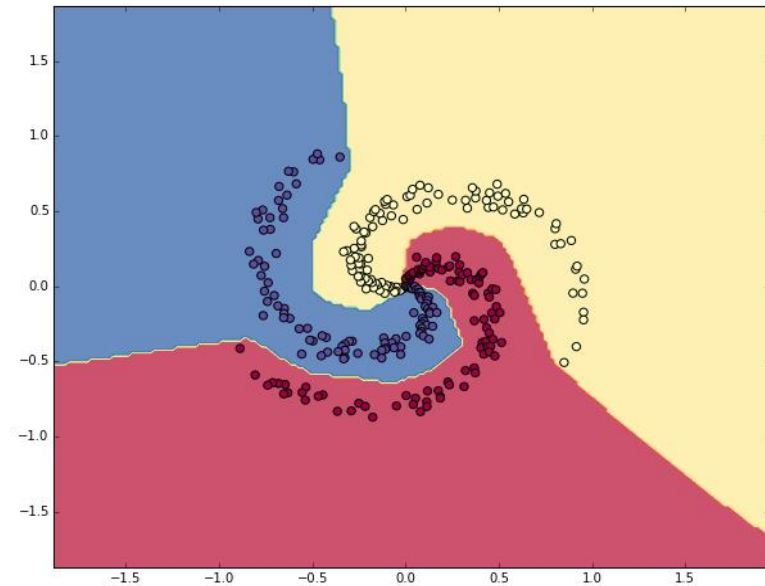


2차원 나선형 데이터를 분류하는 신경망을 개발해보자.

Data Generation



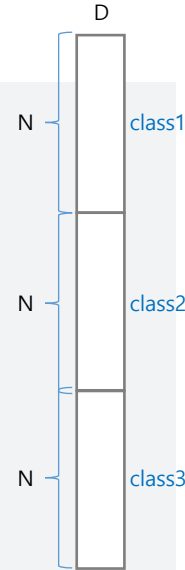
Classification





# Hint : Spiral Dataset Generation

행렬 X의 모양



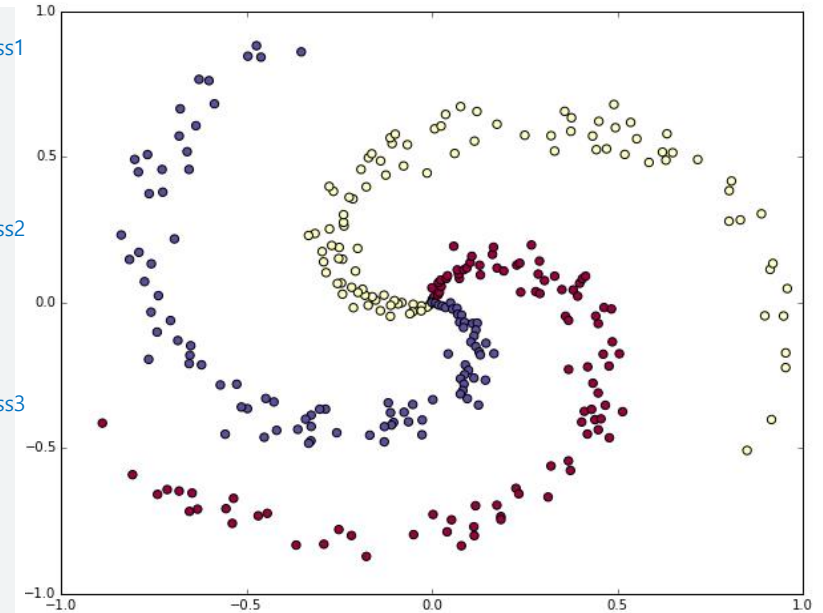
```
N = 100 # 클래스 별 포인트 개수
D = 2   # 차원
K = 3   # 클래스 개수
X = np.zeros((N*K,D)) # 데이터
y = np.zeros(N*K, dtype='uint8') # 레이블 (클래스)

for j in xrange(K):
    ix = range(N*j,N*(j+1)) # j번째 클래스
    r = np.linspace(0.0,1,N) # 반지름 [0,1]
    t = np.linspace(j*4,(j+1)*4,N) + np.random.randn(N)*0.2 # 각도 [0, 4]
    X[ix] = np.c_[r*np.sin(t), r*np.cos(t)]
    y[ix] = j
```

# lets visualize the data:

```
plt.scatter(X[:, 0], X[:, 1], c=y, s=40, cmap=plt.cm.Spectral)
plt.show()
```

color   marker size   color map



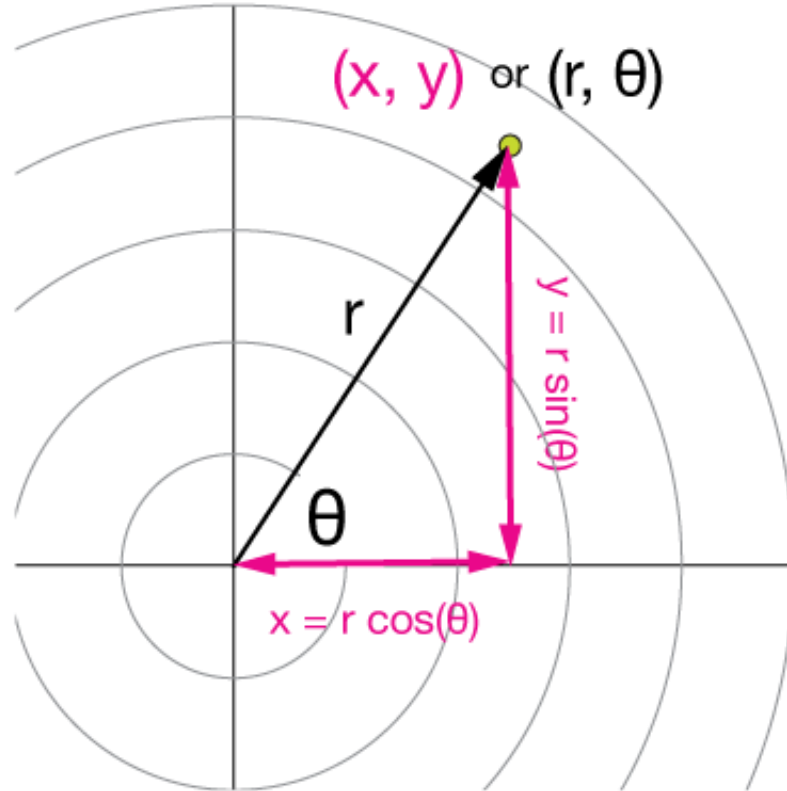
**X** (300,2) 입력 데이터   **Y** (300) 레이블

- 반지름  $r$ 은  $[0,1]$ 에서 생성
- 각도  $t$ 는  $[0, 4]$ 에서 생성 + 노이즈 추가
- 각도가 4이어야 등간격으로 예쁘게 나옴
- 좌표가  $r*\sin(t)$ ,  $r*\cos(t)$ 라서 시계 방향으로 회전하는 나선형이 됨
- `np.c_`는 두 배열을 column으로 방향으로 합침

<https://cs231n.github.io/neural-networks-case-study/>



# Hint : Spiral Dataset Generation

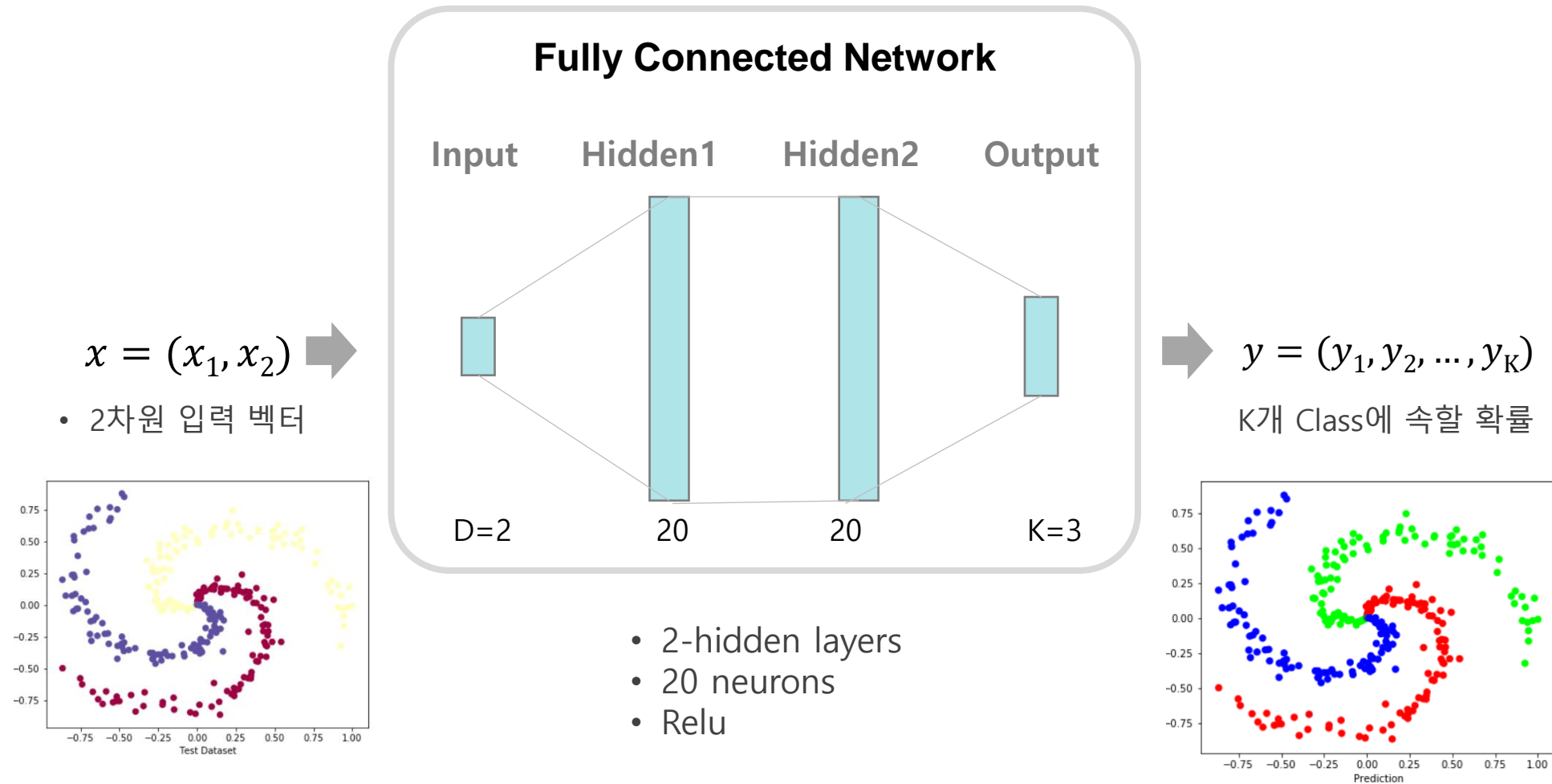


## 데이터 생성

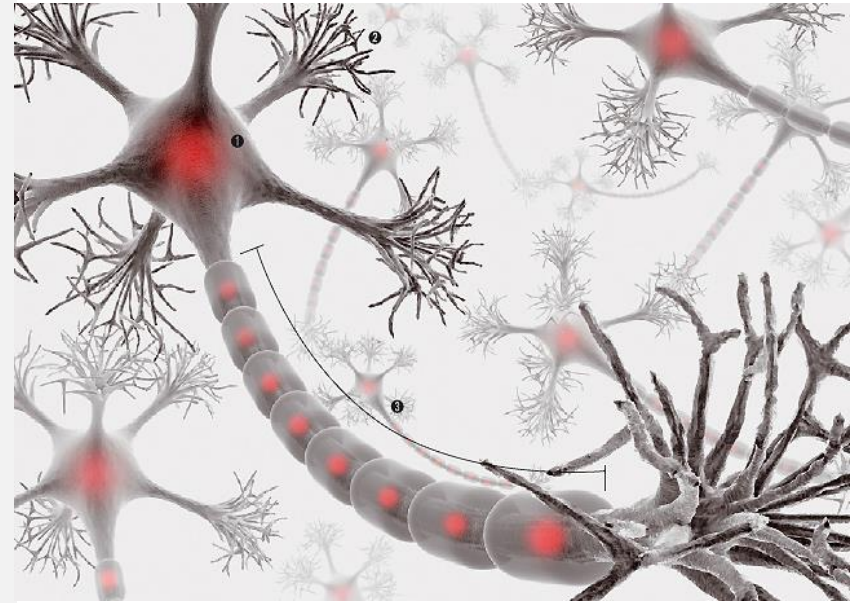
- 반지름  $r$ 은  $[0, 1]$ 에서 생성
- 각도  $t$ 는  $[0, 4]$ 에서 생성 + 노이즈 추가
- 각도가 4이어야 등간격으로 예쁘게 나옴



# Hint : Network 구성



## 2 데이터 준비



# 패키지 импорт

```
# tensorflow를 임포트합니다
import tensorflow as tf

# 헬퍼(helper) 라이브러리를 임포트합니다
import numpy as np
import matplotlib.pyplot as plt

print(tf.__version__)
```

2.0.0-dev20190524



# 데이터셋 생성

```
# generate the data
theta = 4 # class 별 시작 각도 : theta 배수

def generate_spiral_dataset(num_data, num_class, num_dim=2):
    input_data = np.zeros((num_data*num_class,num_dim)) # data matrix (each row = single example)
    output_data = np.zeros(num_data*num_class, dtype='uint8') # class labels

    for j in range(num_class):
        ix = range(num_data*j,num_data*(j+1)) # jth class data index

        r = np.linspace(0.0,1,num_data) # radius [0,1]
        t = np.linspace(j*theta,(j+1)*theta,num_data) + np.random.randn(num_data)*0.2 # theta [0, 4]

        input_data[ix] = np.c_[r*np.sin(t), r*np.cos(t)] # inputs (num_data*num_class, num_dim)
        output_data[ix] = j # output (num_data*num_class)

    return input_data, output_data
```

# 데이터셋 생성

```
N = 200 # number of points per class
D = 2   # dimensionality
K = 3   # number of classes
```

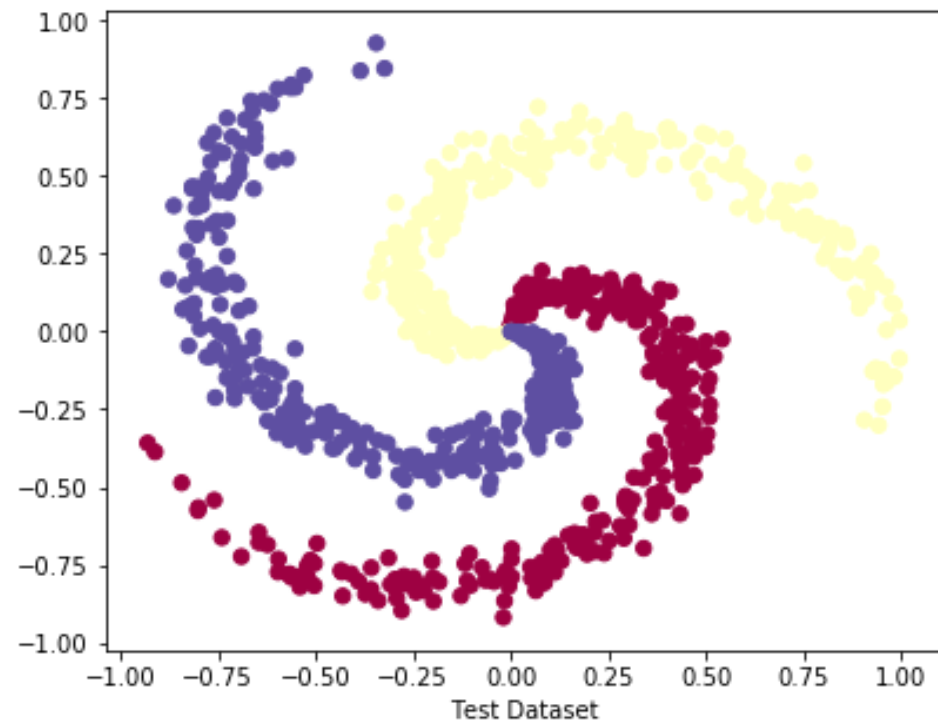
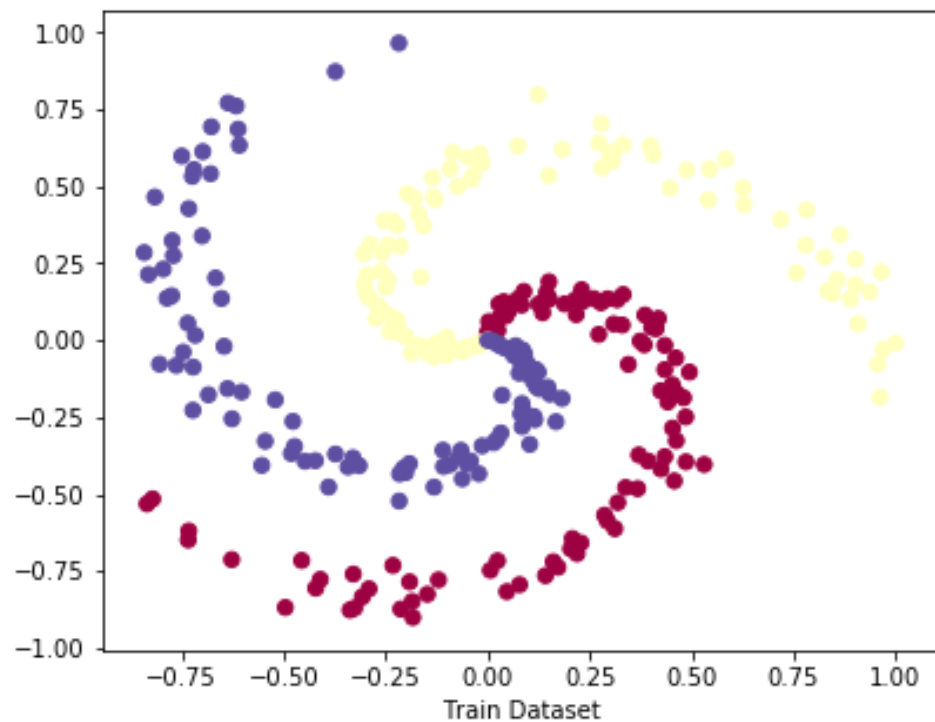
```
train_x, train_y = generate_spiral_dataset(N, K, D)
test_x, test_y = generate_spiral_dataset(N, K, D)
```

```
# lets visualize the data:
```

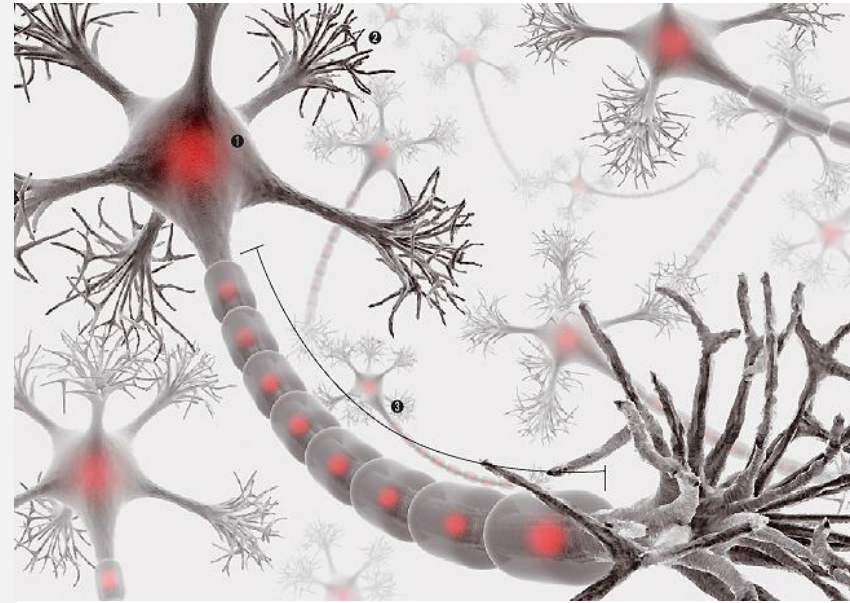
```
plt.figure(figsize=(14,5))
plt.subplot(1,2,1)
plt.scatter(train_x[:, 0], train_x[:, 1], c=train_y, s=40, cmap=plt.cm.Spectral)
plt.xlabel("Train Dataset")
```

```
plt.subplot(1,2,2)
plt.scatter(test_x[:, 0], test_x[:, 1], c=test_y, s=40, cmap=plt.cm.Spectral)
plt.xlabel("Test Dataset")
plt.show()
```

# 데이터셋 생성



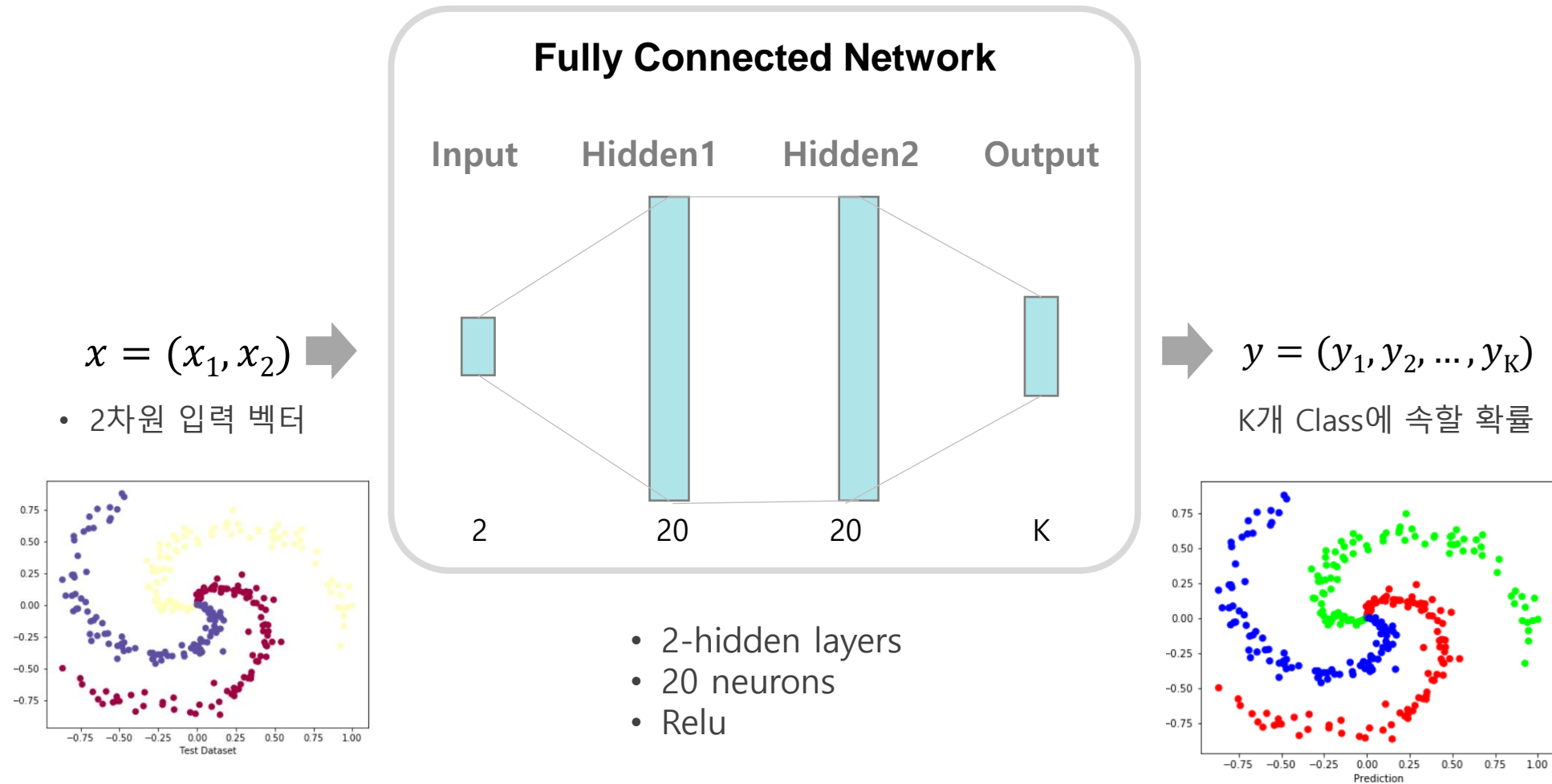
### 3 모델 정의 및 훈련, 검증





# 문제

Spiral Classification 코드를 Keras로 변경해보자!



# 모델 정의 (문제)



```
# Dense를 이용해서 3계층 신경망을 구축하시오.  
# hidden 20, hidden 20, output K  
# 첫번째 Dense 계층에 input_shape을 지정하시오  
model = keras.Sequential(#your code)
```

## 참고 tf.keras.layers.Dense

```
tf.keras.layers.Dense(  
    units, activation=None, use_bias=True, kernel_initializer='glorot_uniform',  
    bias_initializer='zeros', kernel_regularizer=None, bias_regularizer=None,  
    activity_regularizer=None, kernel_constraint=None, bias_constraint=None,  
    **kwargs  
)
```

- **units**: 뉴런 개수, Positive integer, dimensionality of the output space.
- **activation**: Activation function to use. If you don't specify anything, no activation is applied (ie. "linear" activation:  $a(x) = x$ ).
- **use\_bias**: Boolean, whether the layer uses a bias vector.
- **kernel\_initializer**: Initializer for the kernel weights matrix.
- **bias\_initializer**: Initializer for the bias vector.
- **kernel\_regularizer**: Regularizer function applied to the kernel weights matrix.
- **bias\_regularizer**: Regularizer function applied to the bias vector.
- **activity\_regularizer**: Regularizer function applied to the output of the layer (its "activation")..
- **kernel\_constraint**: Constraint function applied to the kernel weights matrix.
- **bias\_constraint**: Constraint function applied to the bias vector.

[https://www.tensorflow.org/api\\_docs/python/tf/keras/layers/Dense](https://www.tensorflow.org/api_docs/python/tf/keras/layers/Dense)

# 모델 컴파일 (문제)



훈련에 필요한 Optimizer, Loss, Metrics 설정

(단, Optimizer는 Adam, Loss는 Label을 One-Hot Encoding 해주는 Cross Entropy로 Metric은 accuracy로 설정)

```
model.compile(optimizer= #your code,  
              loss= #your code,  
              metrics= #your code)
```



## 참고 tf.keras.losses.SparseCategoricalCrossentropy

```
tf.keras.losses.SparseCategoricalCrossentropy(  
    from_logits=False, reduction=losses_utils.ReductionV2.AUTO,  
    name='sparse_categorical_crossentropy'  
)
```

**from\_logits:** Model의 출력 값이 Softmax를 거치기 전이면 True, Softmax를 실행했다면 False



### SparseCategoricalCrossentropy vs. CategoricalCrossentropy

- SparseCategoricalCrossentropy는 label을 One-Hot vector로 자동 변환해 줌
- CategoricalCrossentropy는 label이 이미 One-Hot vector로 준비된 경우에 사용

[https://www.tensorflow.org/api\\_docs/python/tf/keras/losses/SparseCategoricalCrossentropy](https://www.tensorflow.org/api_docs/python/tf/keras/losses/SparseCategoricalCrossentropy)



# 모델 훈련 (문제)

모델을 훈련하기 위해 training set, label, epoch 등을 지정

```
# epoch는 100 이상  
model.fit(#your code)
```

```
Epoch 1/100  
19/19 [=====] - 0s 1ms/step - loss: 1.0984 - accuracy: 0.3017  
Epoch 2/100  
19/19 [=====] - 0s 1ms/step - loss: 1.0649 - accuracy: 0.4033  
Epoch 3/100  
19/19 [=====] - 0s 1ms/step - loss: 1.0334 - accuracy: 0.5117  
...  
Epoch 98/100  
19/19 [=====] - 0s 1ms/step - loss: 0.1024 - accuracy: 0.9867  
Epoch 99/100  
19/19 [=====] - 0s 1ms/step - loss: 0.1005 - accuracy: 0.9917  
Epoch 100/100  
19/19 [=====] - 0s 2ms/step - loss: 0.0991 - accuracy: 0.9883
```

[https://www.tensorflow.org/api\\_docs/python/tf/keras/Model](https://www.tensorflow.org/api_docs/python/tf/keras/Model)

# 모델 정의 (테스트)

```
test_output_pred = model.predict(test_x)
```

```
plt.figure(figsize=(14,5))
```

```
plt.subplot(1,2,1)
```

```
plt.scatter(test_x[:, 0], test_x[:, 1], c=test_y, s=40, cmap=plt.cm.Spectral)
```

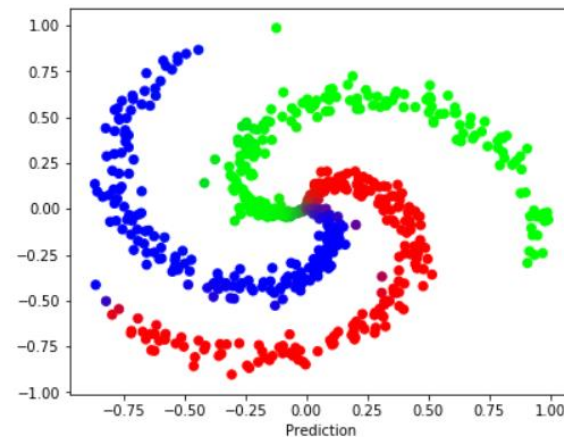
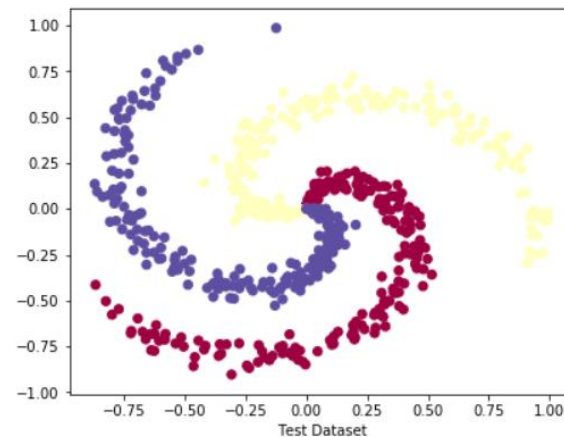
```
plt.xlabel("Test Dataset")
```

```
plt.subplot(1,2,2)
```

```
plt.scatter(test_x[:, 0], test_x[:, 1], c=test_output_pred, s=40, cmap=plt.cm.Spectral)
```

```
plt.xlabel("Prediction")
```

```
plt.show()
```



**Thank you!**

