

```
In [17]: ▶ ##matplotlib notebook
import os
import sys
sys.path.append("numpy_path")
import numpy as np
import struct
from matplotlib import pyplot as plt
import keras
from keras.models import Sequential, load_model
from keras.layers import Dense, Dropout, Activation
from keras.optimizers import RMSprop
import keras.callbacks as cb
from keras.callbacks import EarlyStopping, ModelCheckpoint
from math import cos, sin, pi
from statistics import mean
import os.path
import math
shape_size = 48
```

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In [18]: # define loss history
class LossHistory(cb.Callback):
    def on_train_begin(self, logs={}):
        self.losses = []
    def on_batch_end(self, batch, logs={}):
        batch_loss = logs.get('loss')
        self.losses.append(batch_loss)

#plot losses
def plot_losses(losses):
    plt.plot(losses)
    plt.title('Loss per batch')
    plt.show()

def feature_scaling(X):
    X = X.T
    for i in range(7):
        mean = X[i].mean()
        std = X[i].std()
        X[i] = [(x - mean)/std for x in X[i]]
    return X.T

# input dimension
in_dim = 6
out_dim = 200
def init_model():
    model = Sequential()
    model.add(Dense(20, input_dim=in_dim))
    model.add(Dropout(0.2))
    model.add(Activation('relu'))
    model.add(Dense(500))
    model.add(Dropout(0.5))
    model.add(Activation('relu'))
    model.add(Dense(500))
    model.add(Dropout(0.5))
    model.add(Activation('relu'))
    model.add(Dense(200))
    model.add(Dropout(0.5))
    model.add(Activation('relu'))
    model.add(Dense(200))
    model.add(Dropout(0.2))
    model.add(Activation('relu'))
    #     model.add(Dense(70))
    #     model.add(Dropout(0.2))
    #     model.add(Activation('relu'))
    #     model.add(Dense(100))
    #     model.add(Activation('relu'))
    model.add(Dense(out_dim))
    model.add(Activation('sigmoid'))
    # use mean squared error to measure the losses
    model.compile(loss=keras.losses.mean_squared_error,
                  optimizer=keras.optimizers.Adam(lr = 0.001),
                  metrics=['accuracy'])
    return model

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In [19]: data_size = 0
dummy1 = [0]*200
dummy2 = [0]*6
SP = np.array(np.reshape(dummy1, (1, 200)))
SH = np.array(np.reshape(dummy2, (1, 6)))
for i in range(2, 65):
    path = 'meep_code/data/DATA'+str(i)
    if not os.path.exists(path):
        #miss.append(i)
        print('Missing batch:' + str(i))
        continue

    files = next(os.walk(path))[2] #dir is your directory path as string]
    num_data = len(files)
    data_size += num_data
    skip = []

    coordinates = np.genfromtxt('meep_code/data/DATA'+str(i)+'_sh.txt')
    xc, yc = coordinates[:, 0], coordinates[:, 1]
    xc = np.reshape(xc, (num_data, shape_size))
    yc = np.reshape(yc, (num_data, shape_size))

    for j in range(num_data):
        tmp = np.genfromtxt(path+'/'+'DATA'+str(i)+'_sp'+str(j)+'.txt')
        valid = True
        for q in range(200):
            if math.isnan(float(tmp[q])):
                print('Batch '+str(i)+'\tsample '+str(j)+' has NAN value')
                valid = False
                break
            if tmp[q] > 3:
                print('Batch '+str(i)+'\tsample '+str(j)+' has extreme value')
                valid = False
                break
        if not valid:
            #skip.append(j)
            continue
        SP = np.concatenate((SP, np.reshape(tmp, (1, 200))))
        tmp = []
        for q in range(6):
            tmp.append(math.sqrt(xc[j][q]**2 + yc[j][q]**2))
        SH = np.concatenate((SH, np.reshape(np.array(tmp), (1, 6))))
    print('Batch '+str(i)+' has \t'+str(num_data))
print('Total # of data: ' + str(len(SH)))

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Batch 45 has 35
Batch 46 has 100
Batch 47 has 100
Batch 48 has 287
Batch 49 has 13
Batch 50 has 37
Batch 51 has 37
Batch 52 has 106
Batch 53 has 35
Batch 54 has 100
Batch 55 has 100
Batch 56 has 287

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Batch 57 has 35
Batch 58 has 100
Batch 59 has 100
Batch 60 has 287
Batch 61 has 95
Batch 62 has 272
Batch 63 has 272
Batch 64 has 781

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In [27]: distribution = []
print('Total # of data: ' + str(len(SP)))
x = np.genfromtxt('meep_code/data/SP_xaxis.txt')
SP_F, SH_F = np.reshape(SP[1], (1, 200)), np.reshape(SH[1], (1, 6))
for i in range(2, len(SP)):
    peak = 0
    p_index = 0
    p_pos = [(0,0),(0,0)]
    for j in range(1, 200):
        if SP[i][j] < SP[i][p_index]:
            p_index = j
        if SP[i][j - 1] >= 0.6 >= SP[i][j]:
            peak += 1
            p_pos = [((j-1)/2+200, SP[i][j - 1]), (j/2+200, SP[i][j])]
    if peak == 1:
        distribution.append([1, p_index/2+200])
        SP_F = np.concatenate((SP_F, np.reshape(SP[i], (1, 200))))
        SH_F = np.concatenate((SH_F, np.reshape(SH[i], (1, 6))))

```

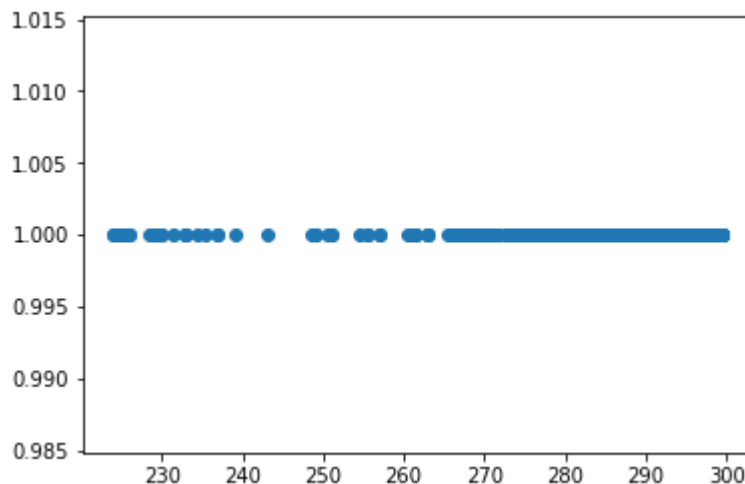
Total # of data: 4989

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In [28]: distribution = np.array(distribution)
plt.scatter(distribution[:,1], distribution[:,0])

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Out[28]: <matplotlib.collections.PathCollection at 0x17701ee0a90>



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In [29]: DATA = np.append(SP_F, SH_F, axis = 1)
          np.random.shuffle(DATA)

          Y = DATA[:, :200]
          X = DATA[:, 200:]

          train_size = int(len(DATA) * 0.8)

          train_X = X[0:train_size, :]
          train_Y = Y[0:train_size, :]
          test_X = X[train_size:, :]
          test_Y = Y[train_size:, :]

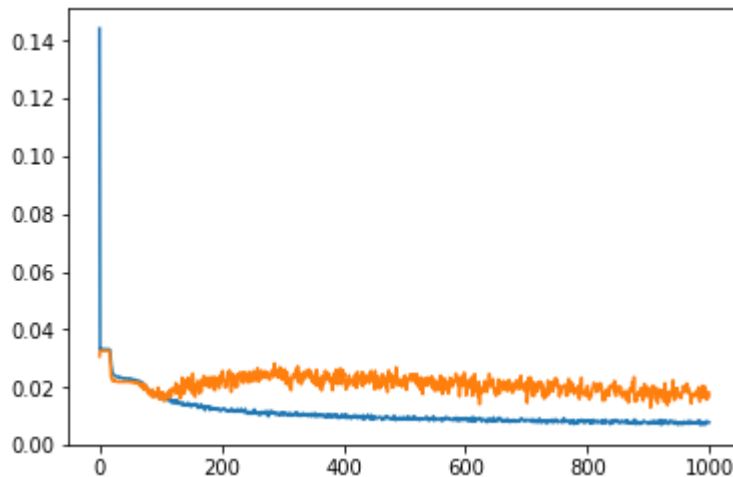
In [31]: model = init_model()
          # history = LossHistory()
          # when training, using minibatch seems to be pretty good
          history = model.fit(train_X, train_Y,
                              epochs=1000,
                              batch_size=20,
                              validation_data=(test_X, test_Y),
                              verbose=2)

          train_score = model.evaluate(train_X, train_Y, batch_size=100)
          test_score = model.evaluate(test_X, test_Y, batch_size= 50)
          print(train_score)
          print(test_score)
          plt.plot(history.history['loss'], label='train')
          plt.plot(history.history['val_loss'], label='test')

[0.015331682799163686, 0.0241935479264426]
[0.0180212862749574, 0.03225806399538953]

```

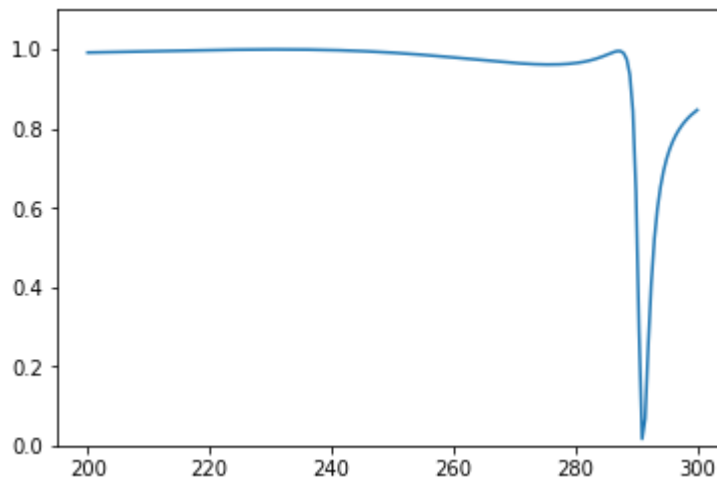
Out[31]: [<matplotlib.lines.Line2D at 0x177008ddc88>]



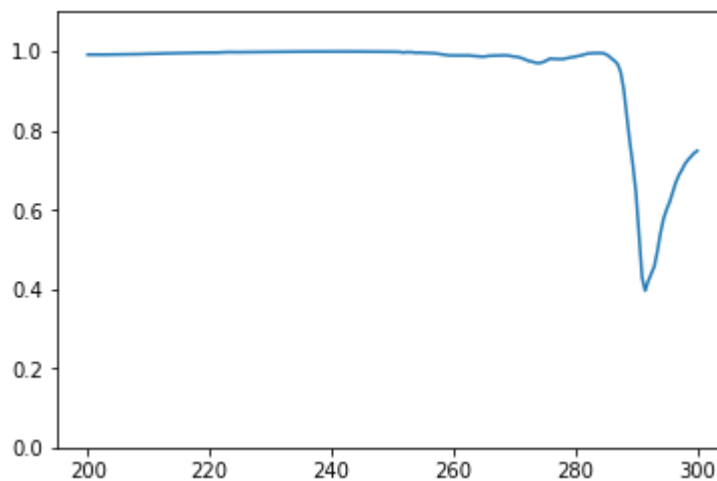
```
In [32]: x = np.genfromtxt('meep_code/data/SP_xaxis.txt')
for i in range(len(test_X)):
    print('Test '+str(i))
    print('True spectrum: ')
    plt.ylim(0, 1.1)
    plt.plot(x, test_Y[i])
    plt.show()
    print('Predicted spectrum: ')
    plt.ylim(0, 1.1)
    plt.plot(x, np.reshape(model.predict(np.reshape(test_X[i], (1, 6))), (200, 1)))
    plt.show()
```

Test 0

True spectrum:

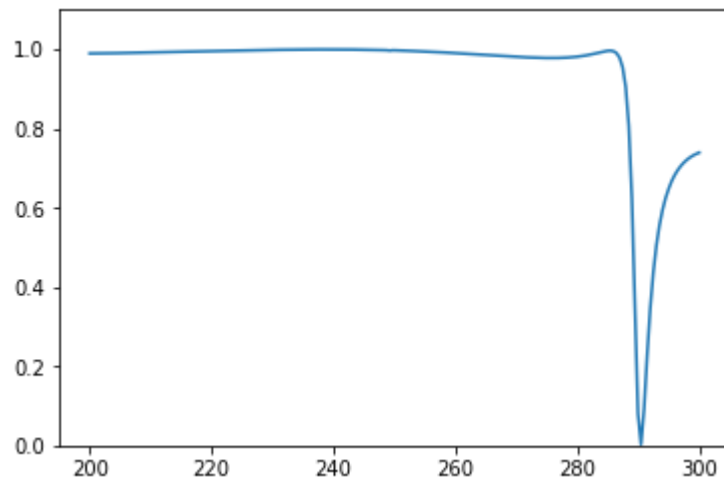


Predicted spectrum:

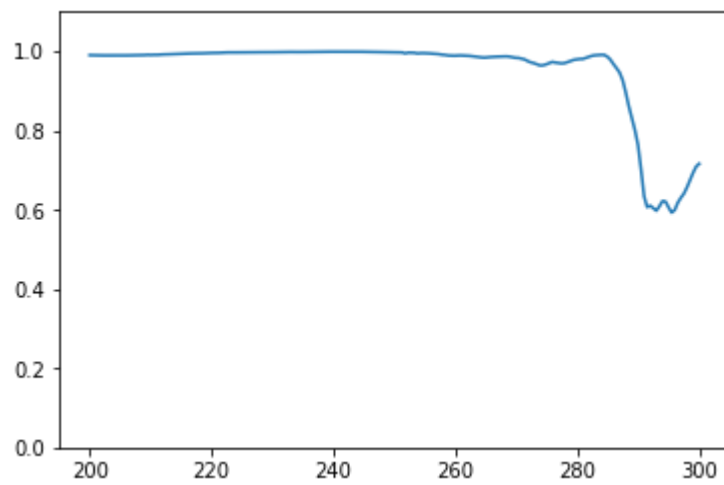


Test 1

True spectrum:

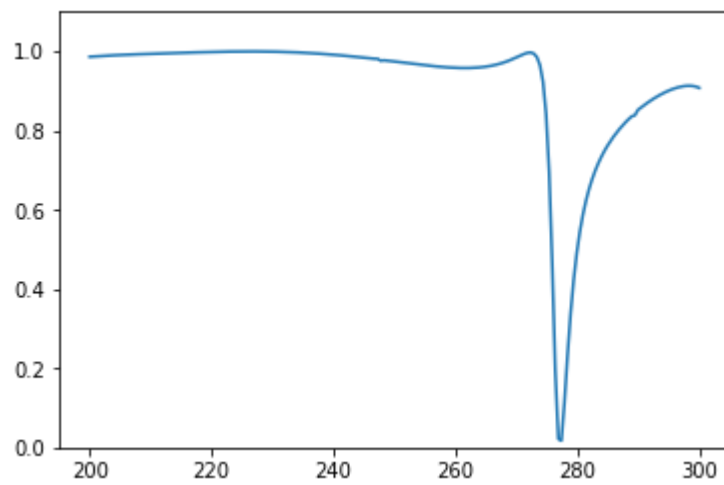


Predicted spectrum:

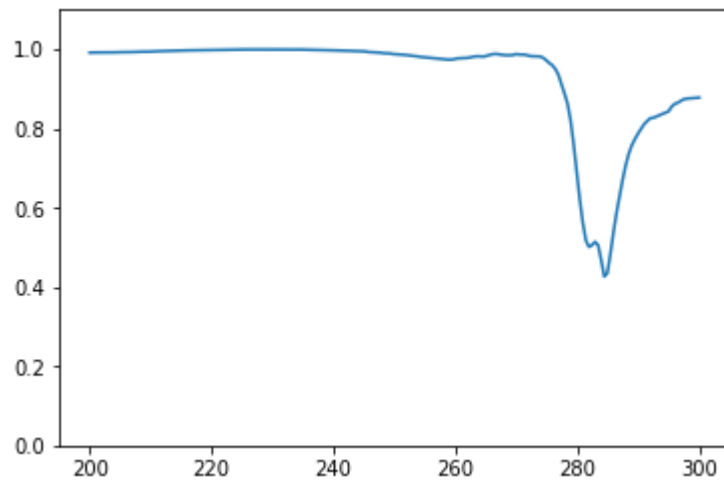


Test 2

True spectrum:

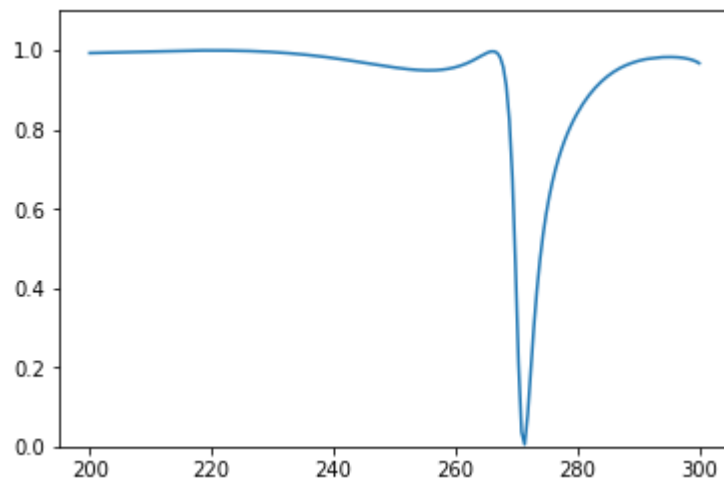


Predicted spectrum:

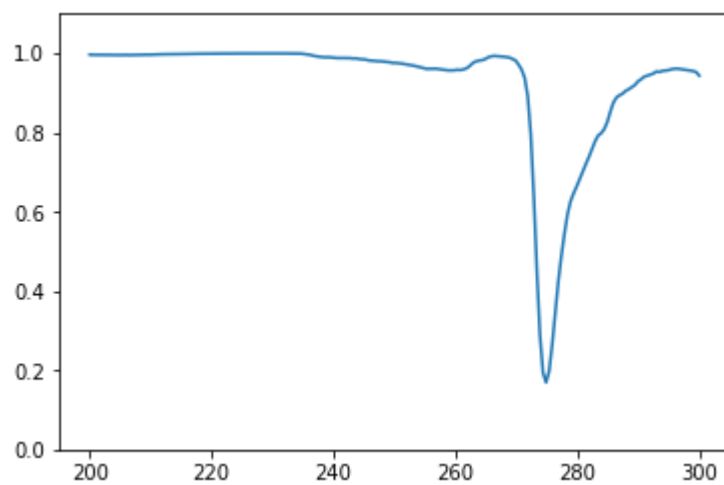


Test 3

True spectrum:

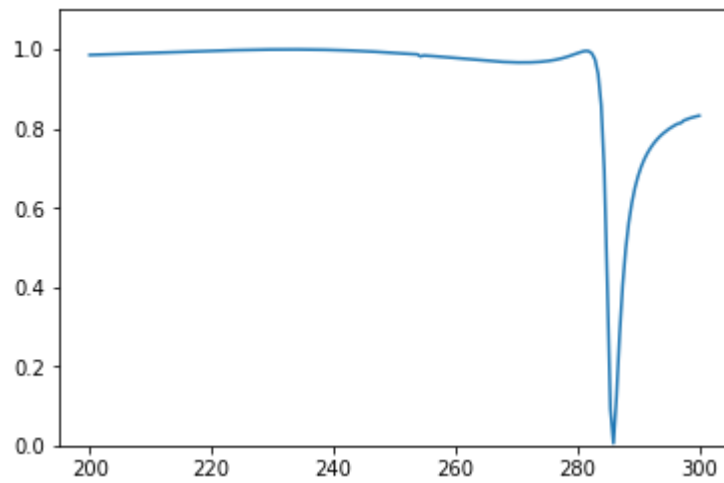


Predicted spectrum:

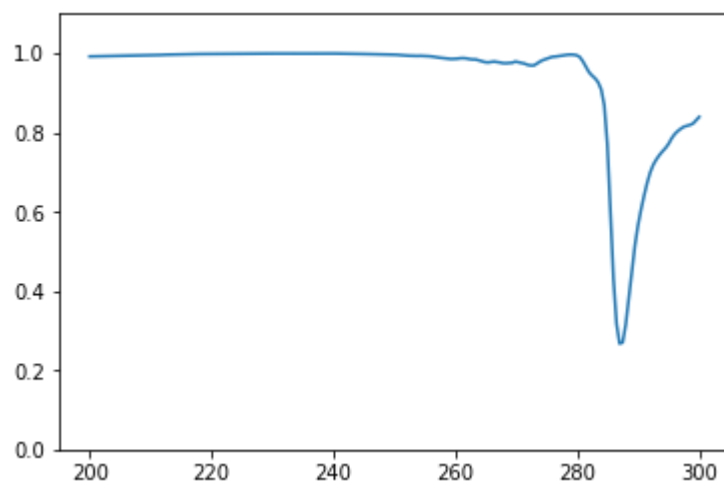


Test 4

True spectrum:

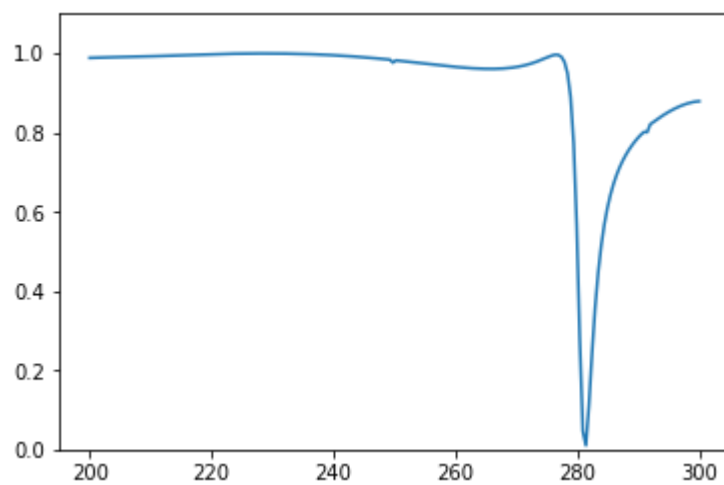


Predicted spectrum:

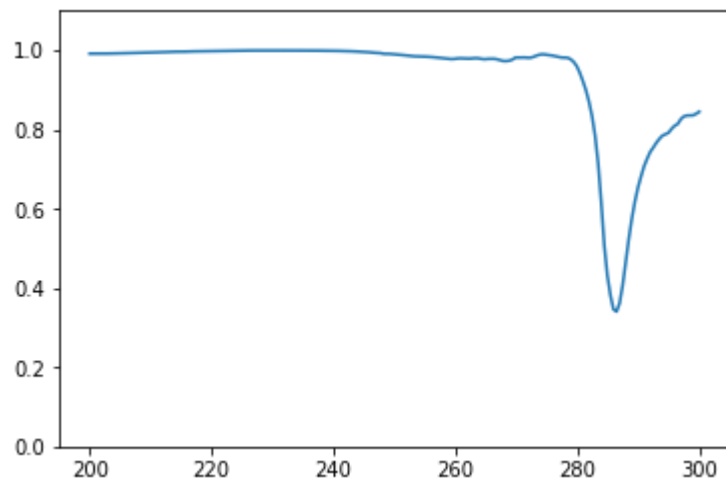


Test 5

True spectrum:

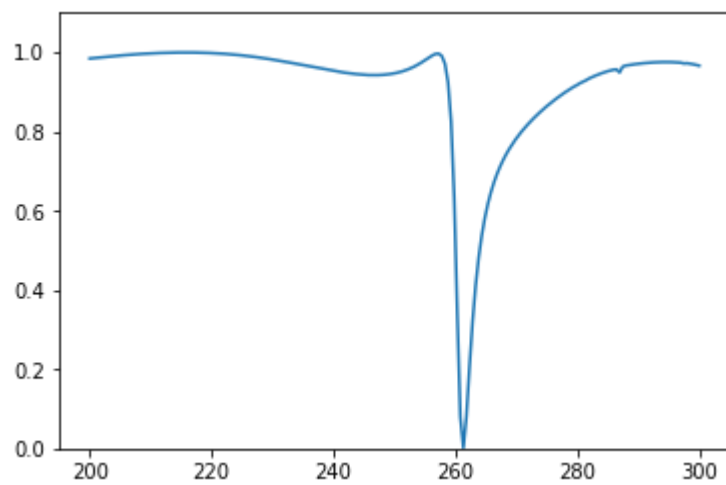


Predicted spectrum:

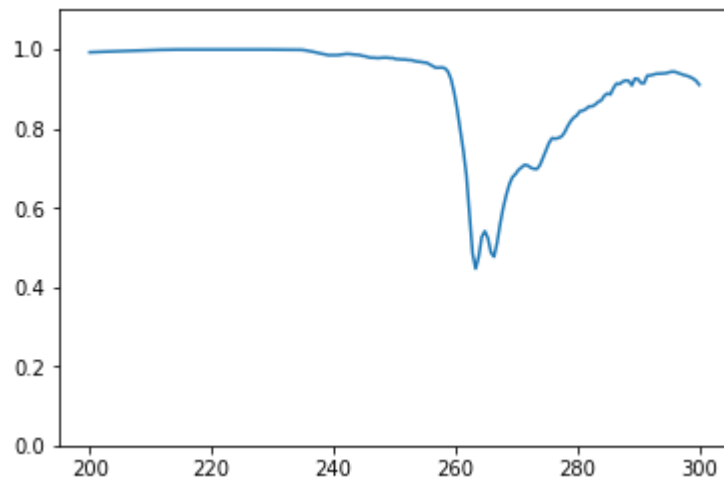


Test 6

True spectrum:

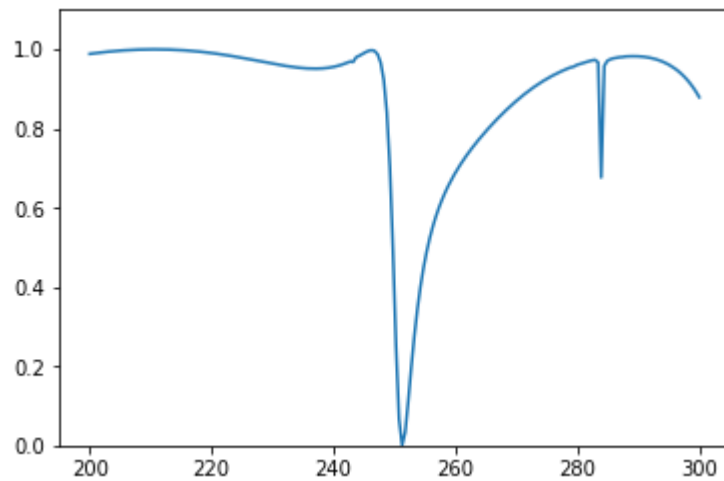


Predicted spectrum:

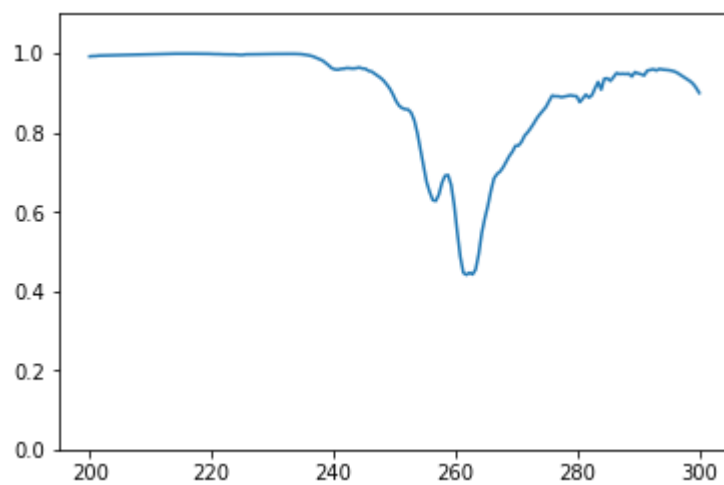


Test 7

True spectrum:

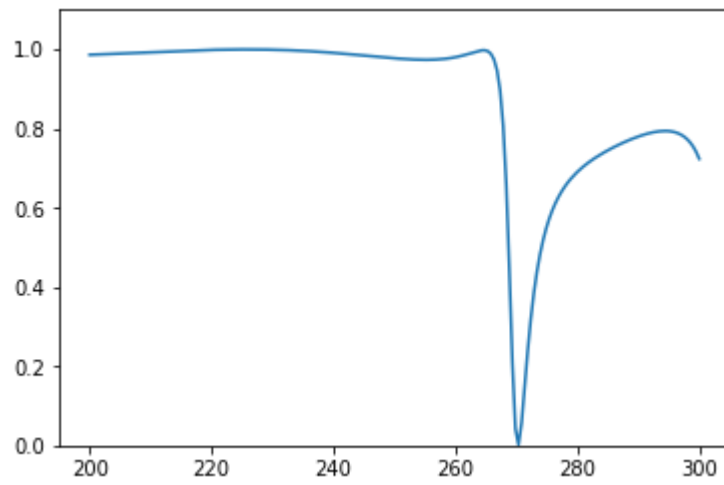


Predicted spectrum:

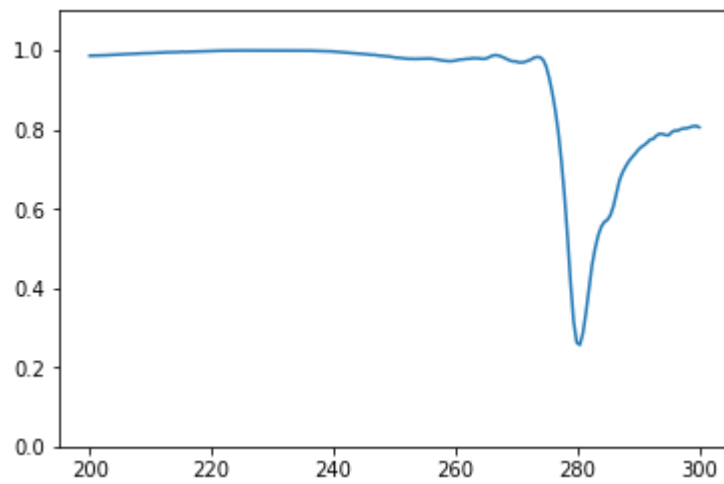


Test 8

True spectrum:

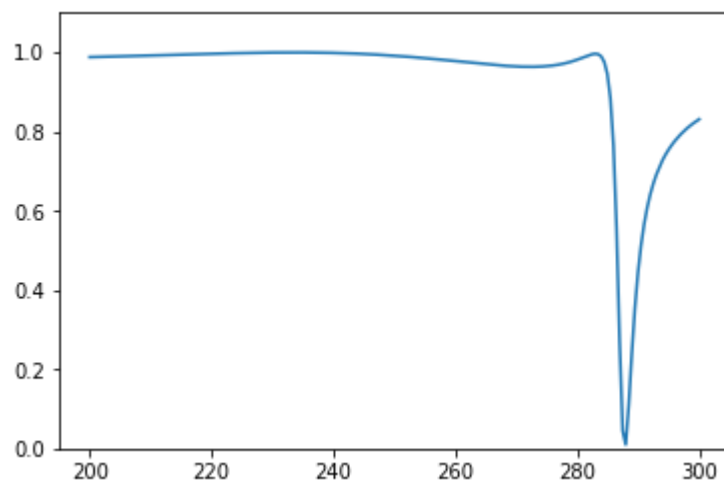


Predicted spectrum:

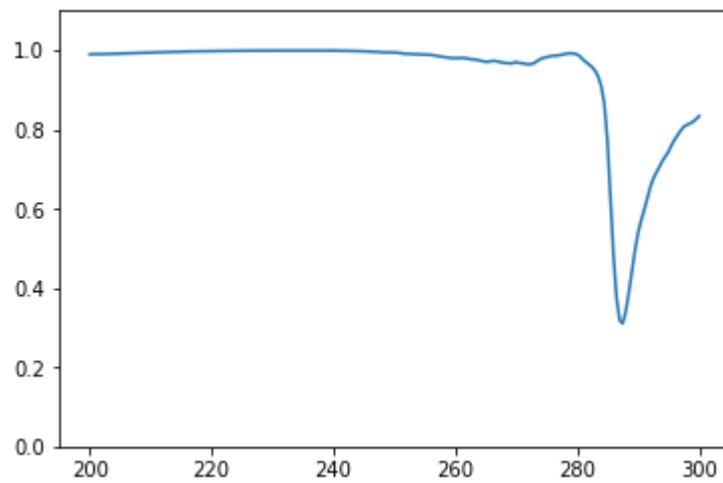


Test 9

True spectrum:

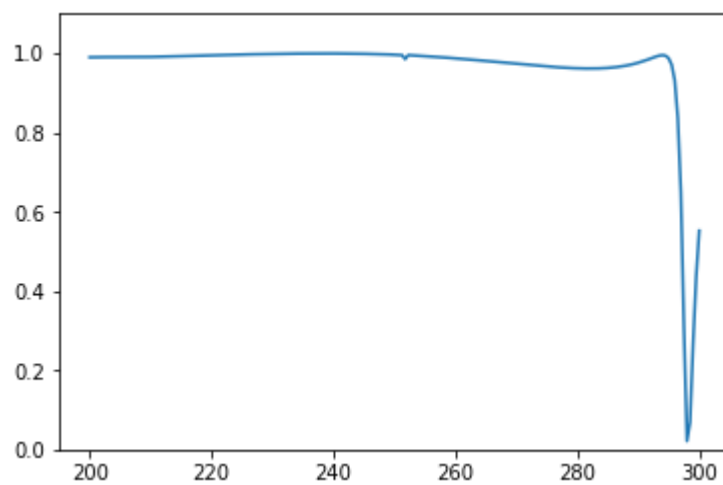


Predicted spectrum:

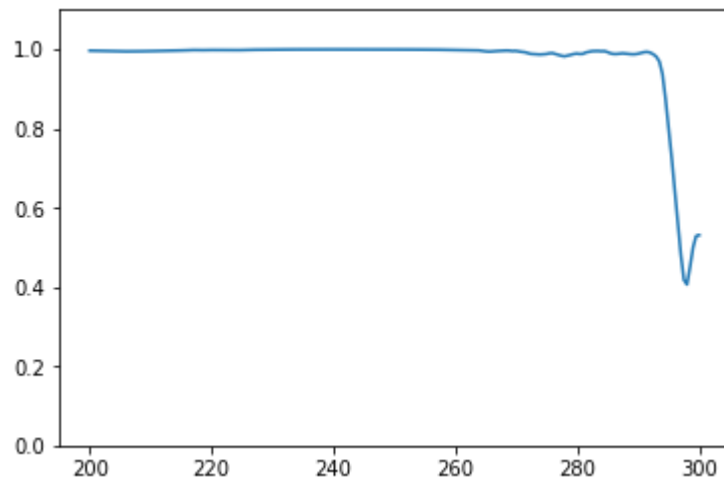


Test 10

True spectrum:

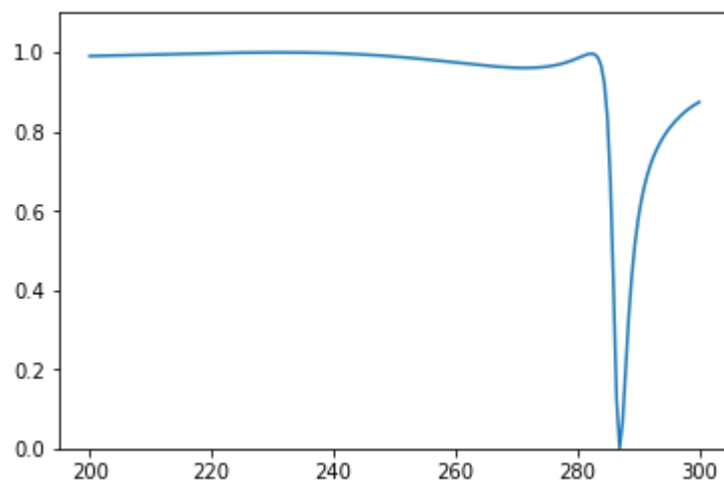


Predicted spectrum:

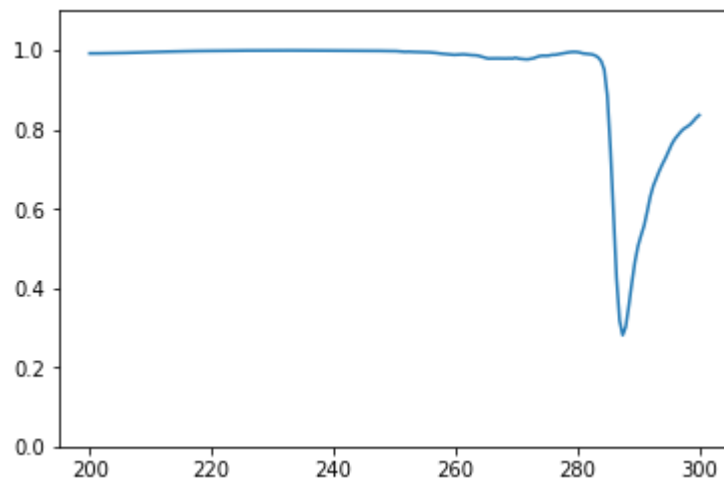


Test 11

True spectrum:

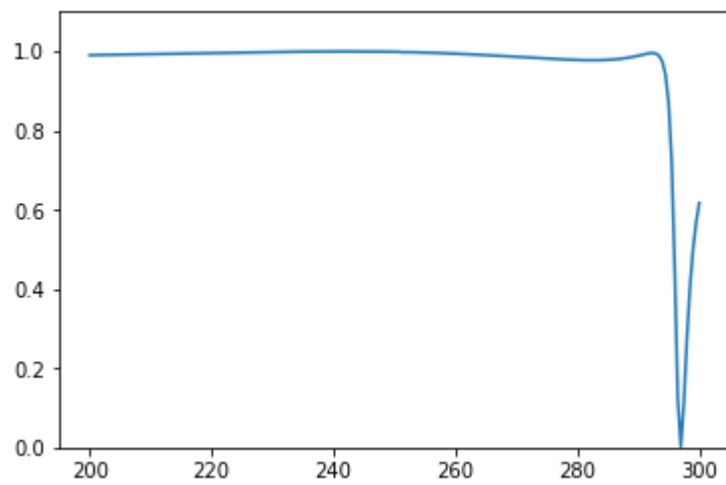


Predicted spectrum:

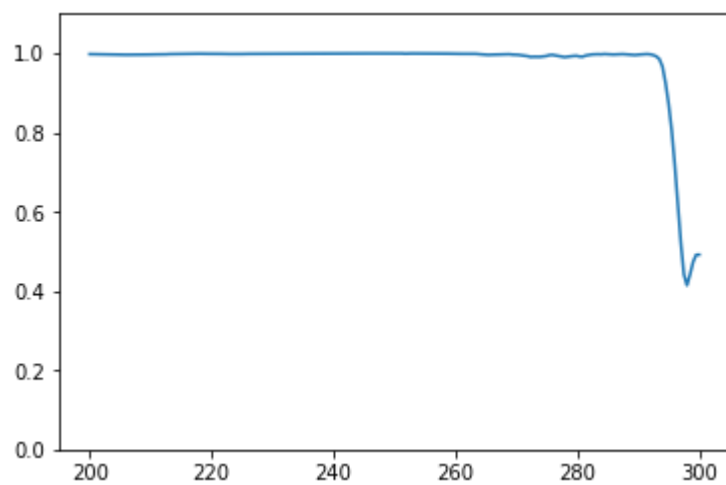


Test 12

True spectrum:



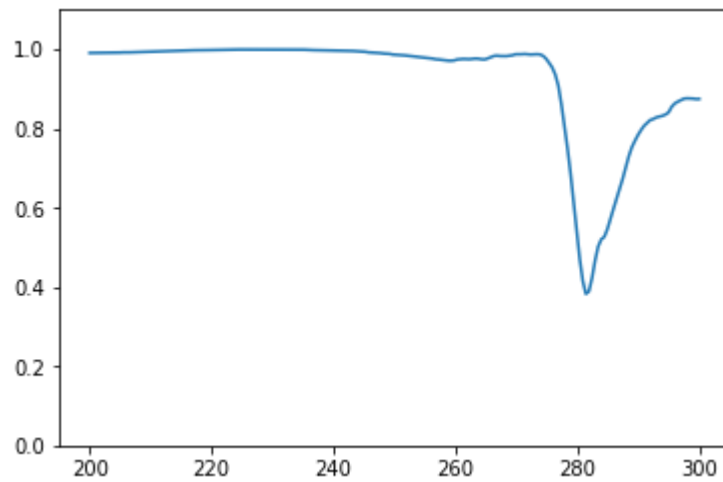
Predicted spectrum:



Test 13

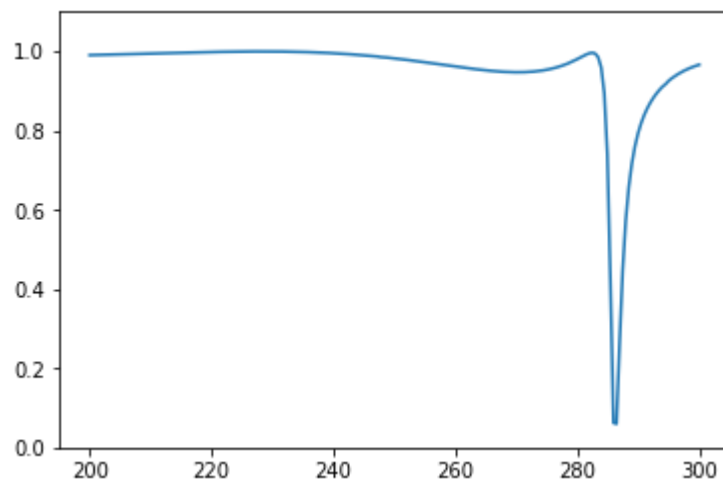
True spectrum:

Predicted spectrum:

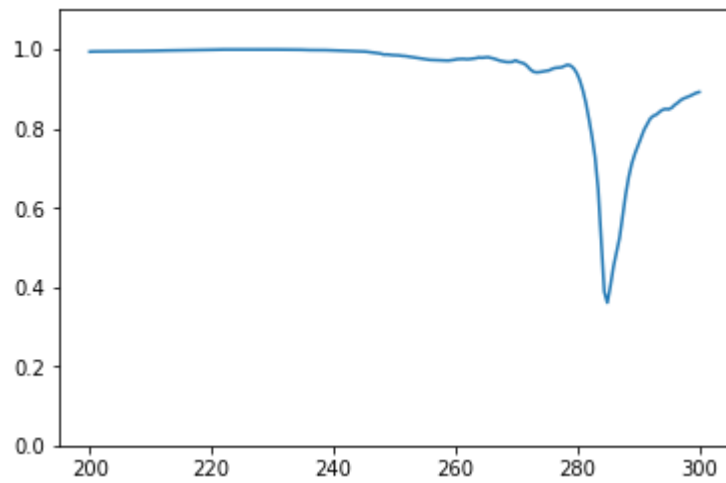


Test 14

True spectrum:

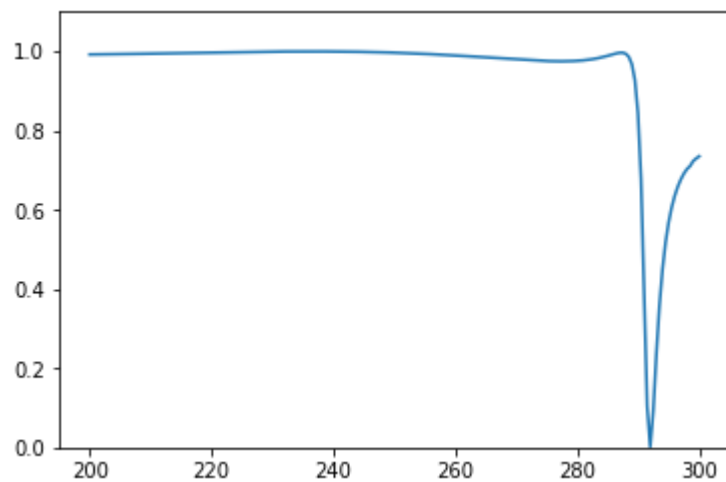


Predicted spectrum:

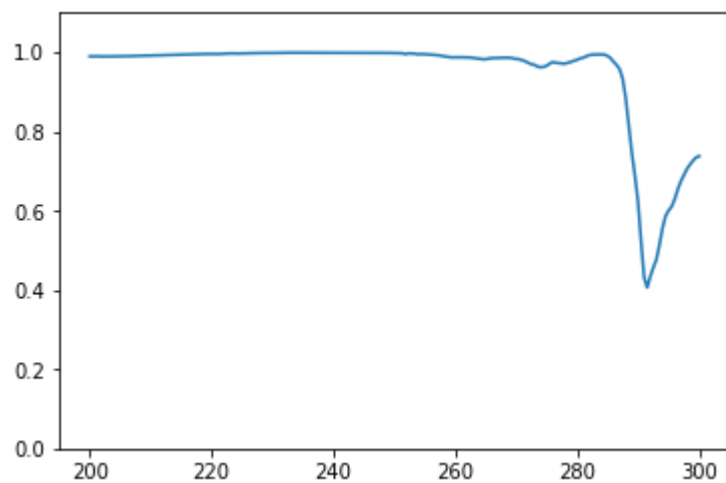


Test 15

True spectrum:

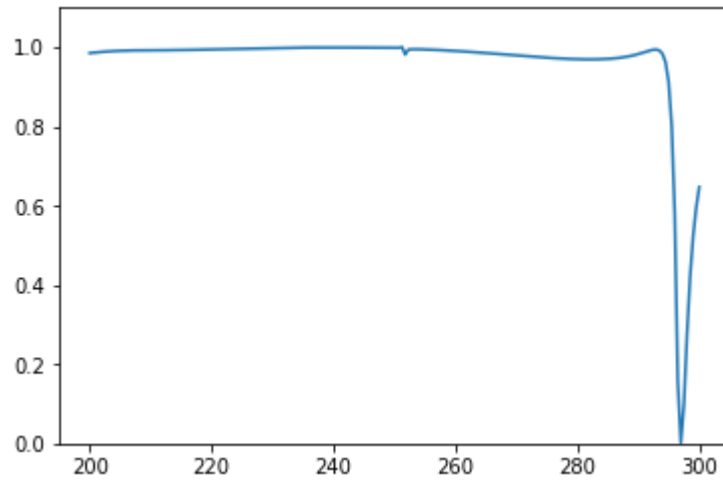


Predicted spectrum:

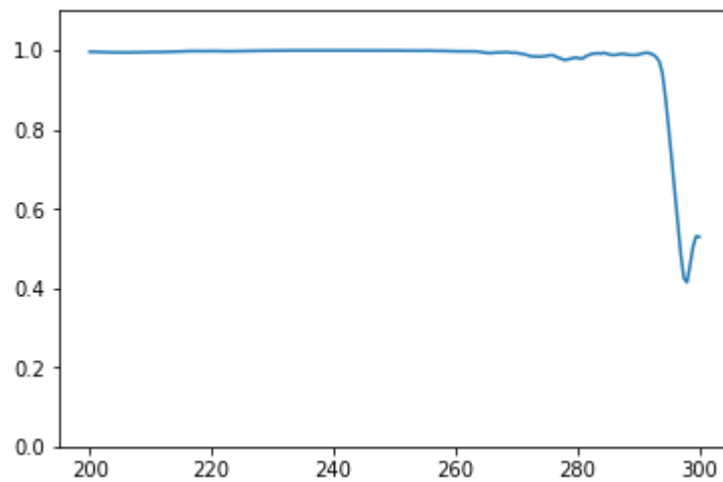


Test 16

True spectrum:

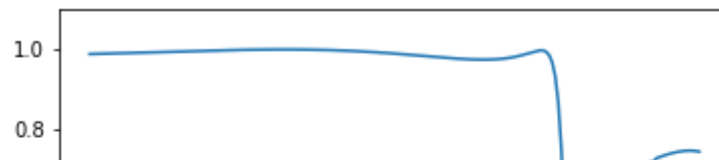


Predicted spectrum:

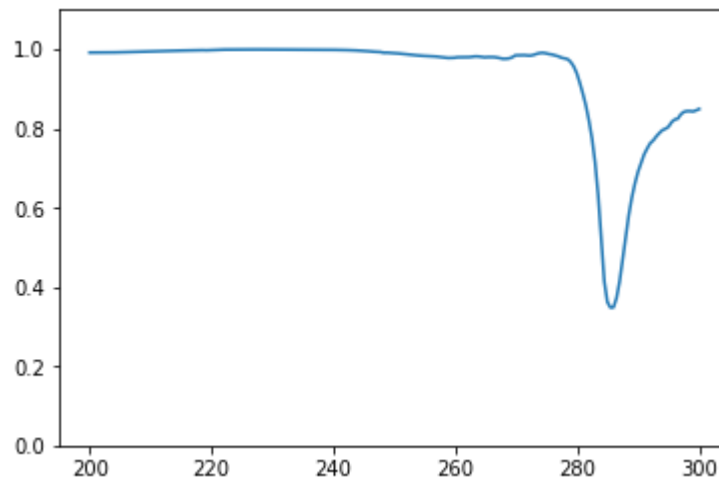


Test 17

True spectrum:

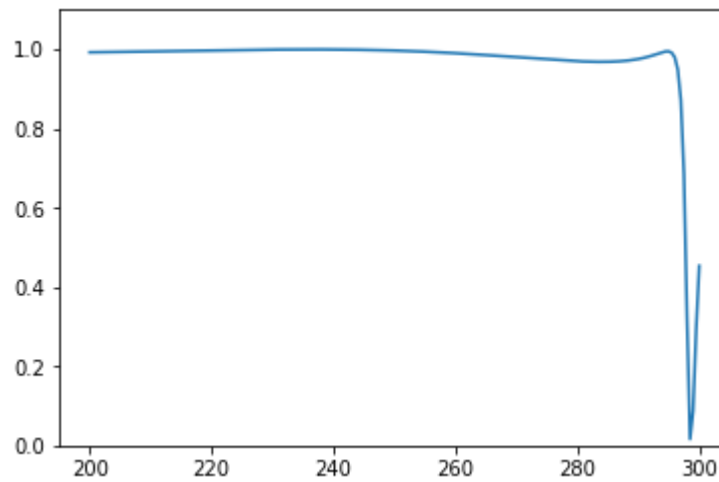


Predicted spectrum:

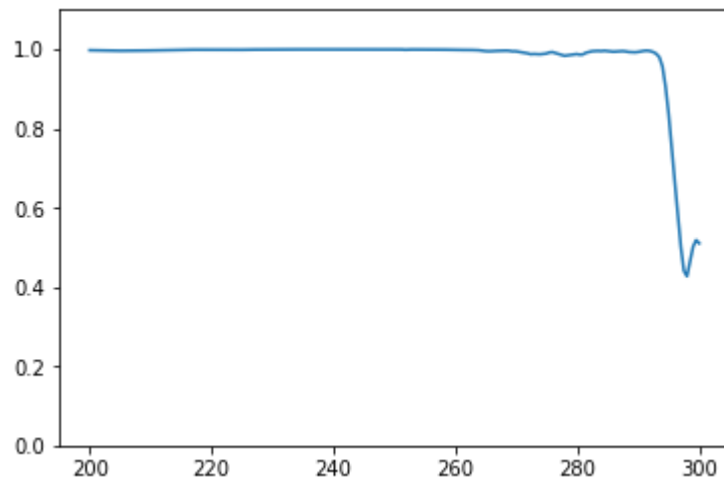


Test 18

True spectrum:

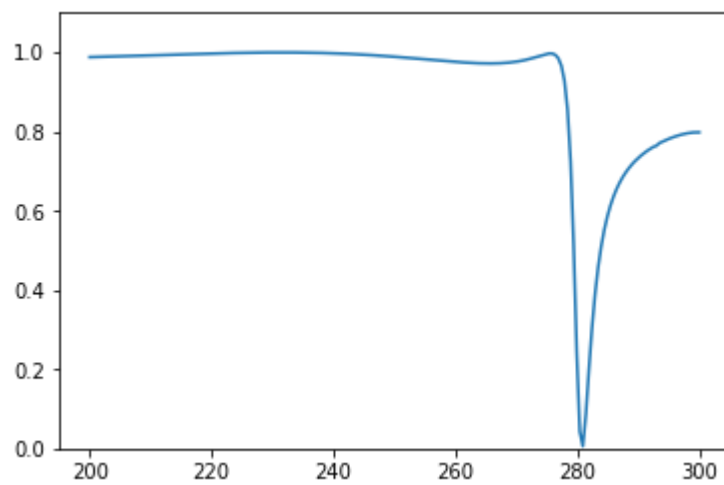


Predicted spectrum:

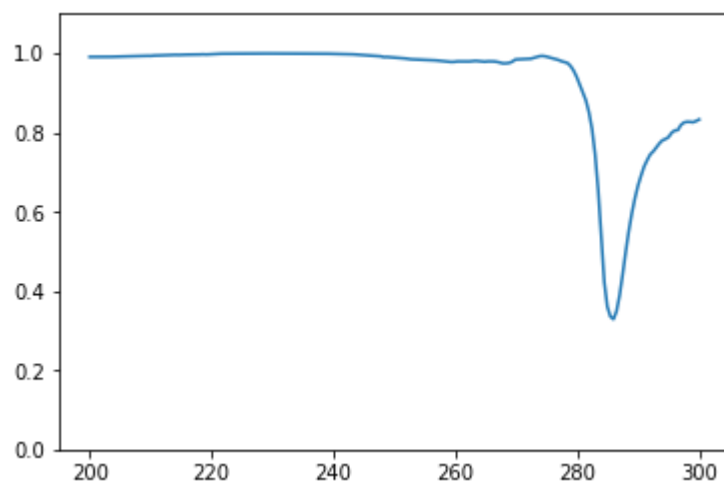


Test 19

True spectrum:

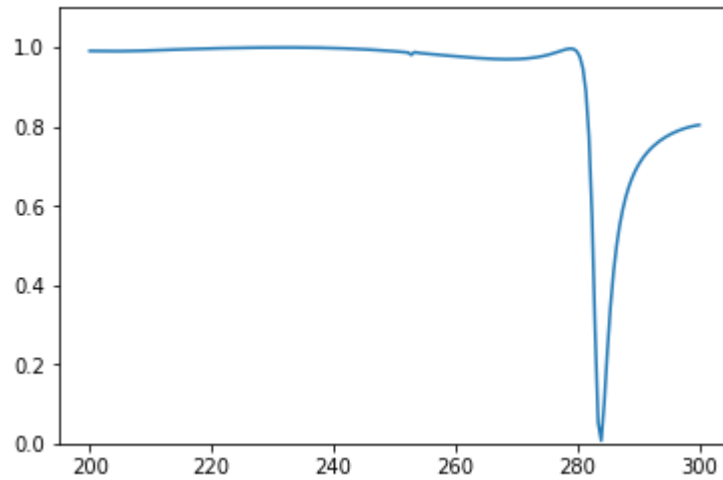


Predicted spectrum:

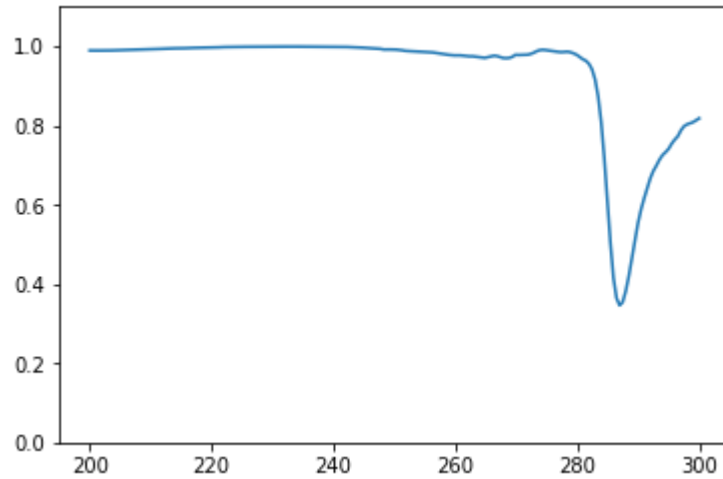


Test 20

True spectrum:



Predicted spectrum:

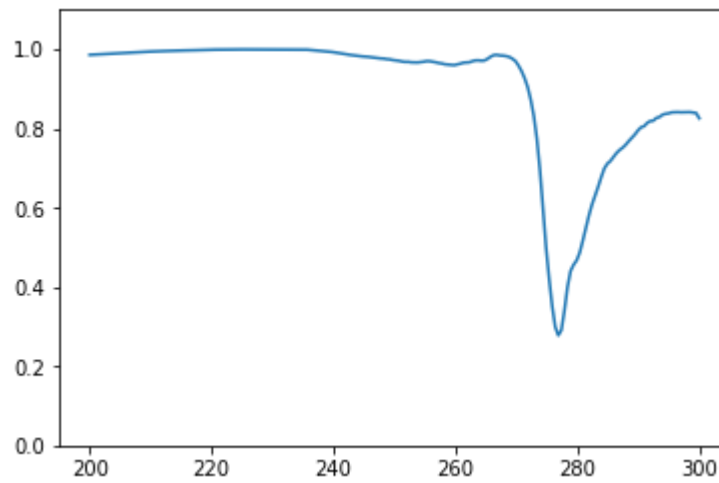


Test 21

True spectrum:

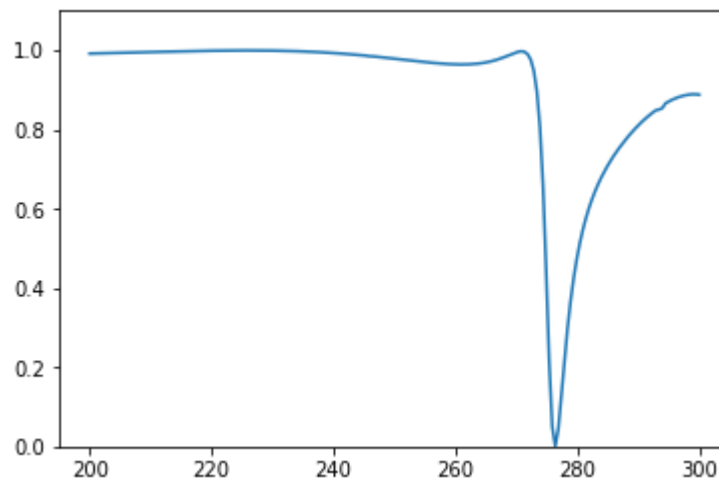


Predicted spectrum:

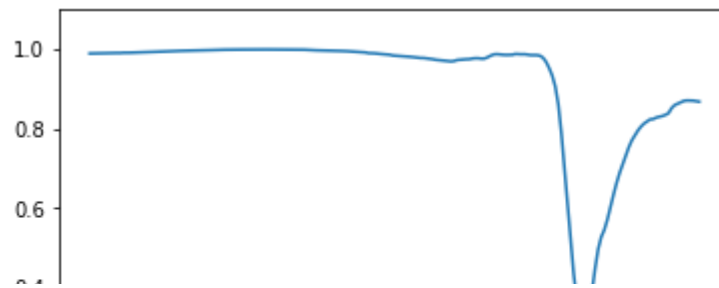


Test 22

True spectrum:

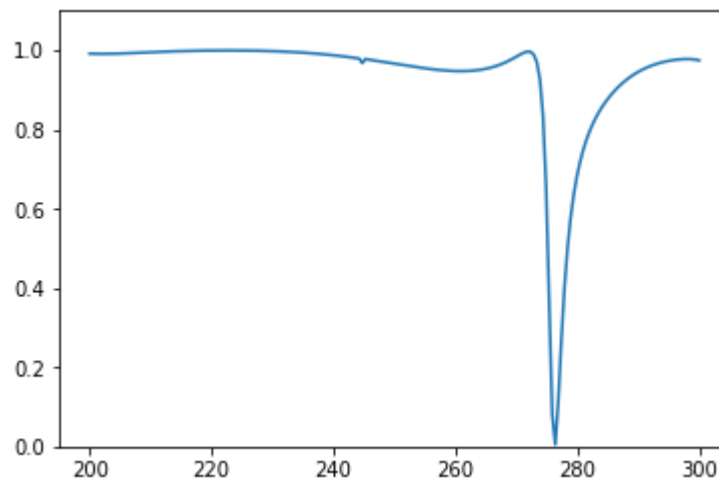


Predicted spectrum:

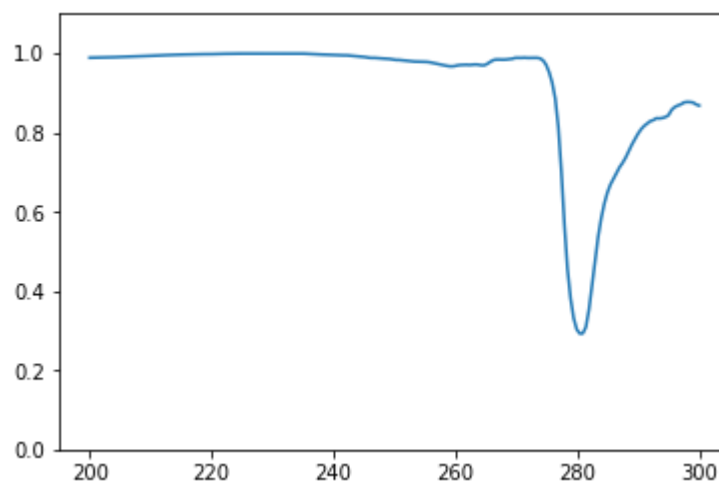


Test 23

True spectrum:

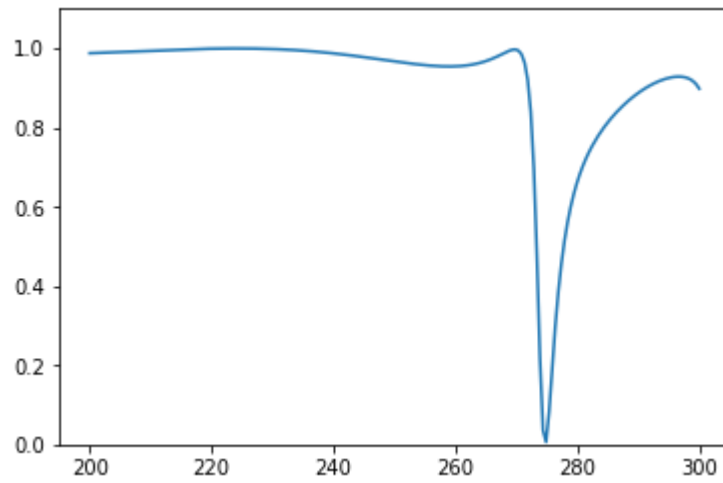


Predicted spectrum:

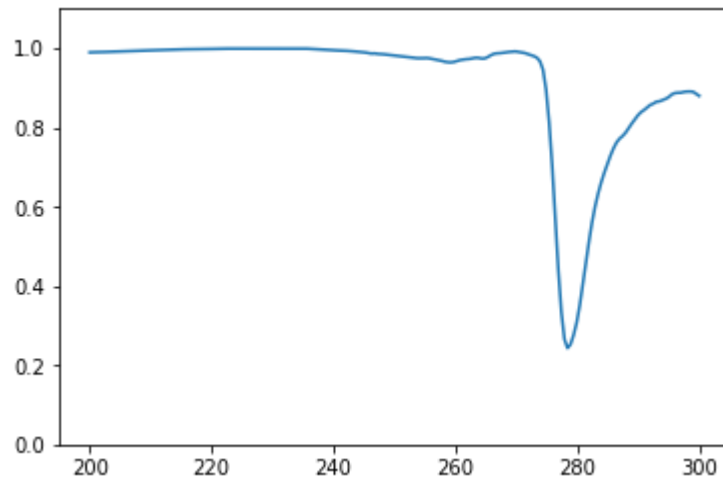


Test 24

True spectrum:

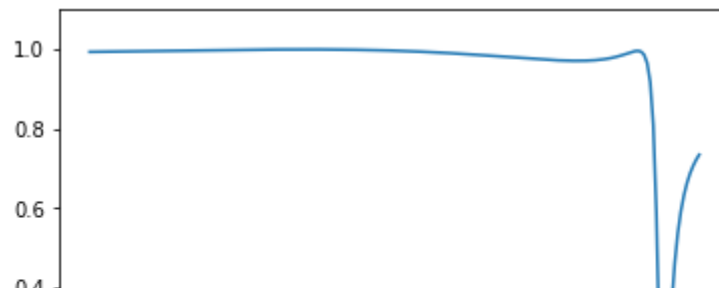


Predicted spectrum:

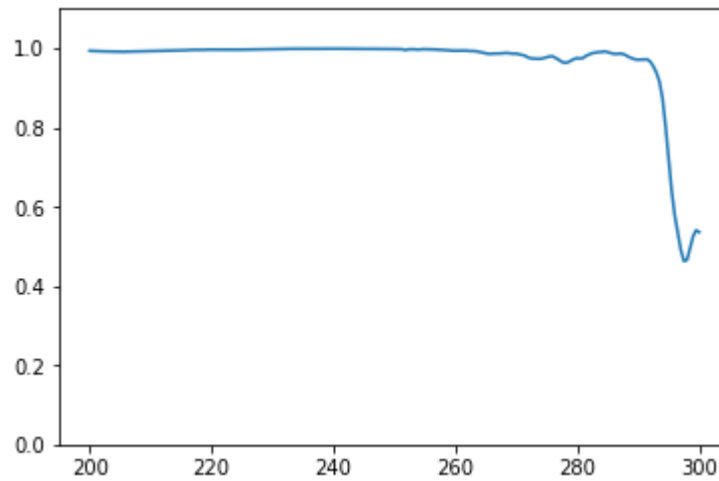


Test 25

True spectrum:

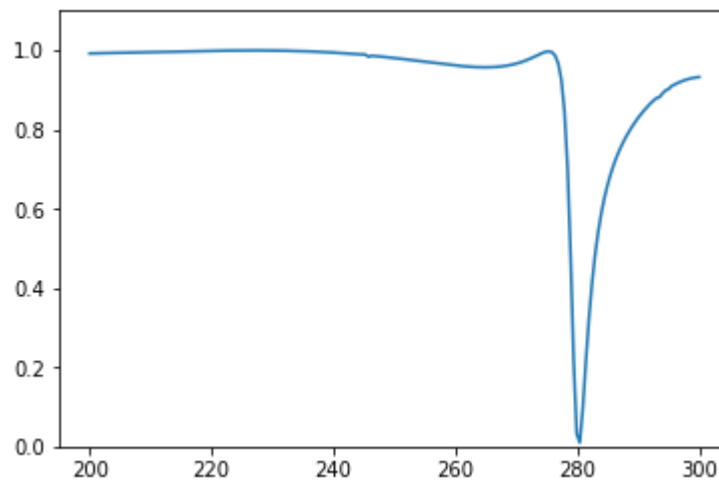


Predicted spectrum:

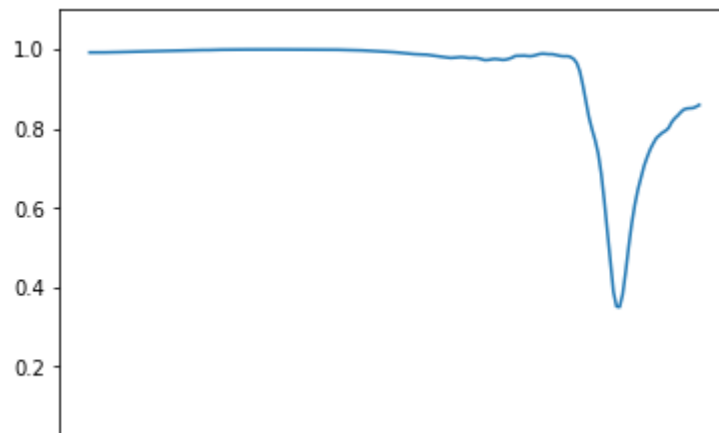


Test 26

True spectrum:

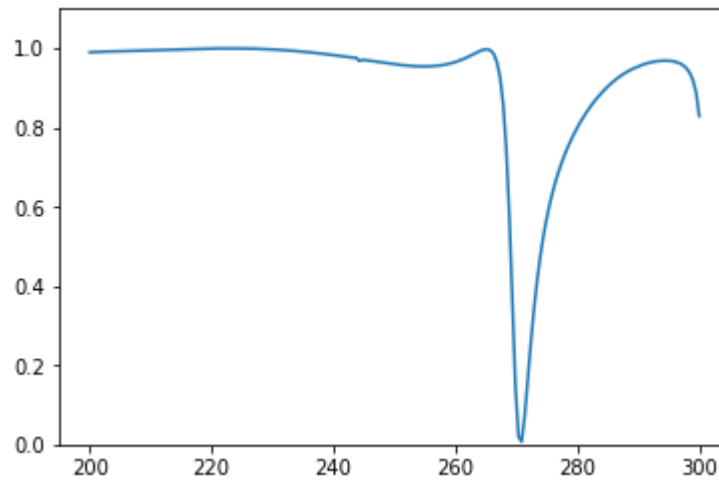


Predicted spectrum:

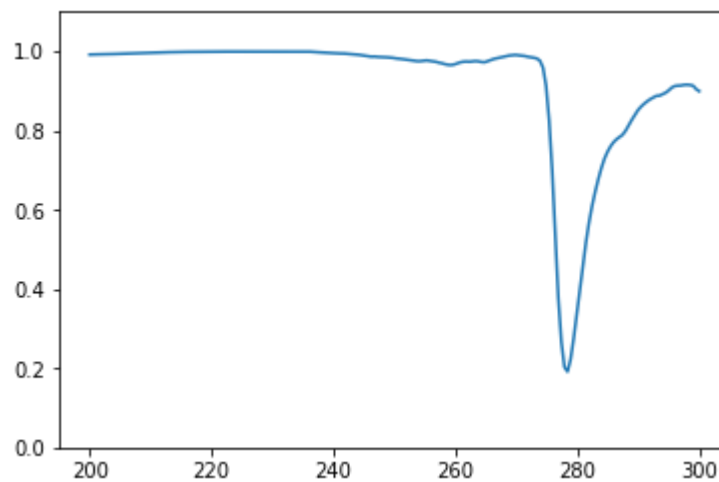


Test 27

True spectrum:

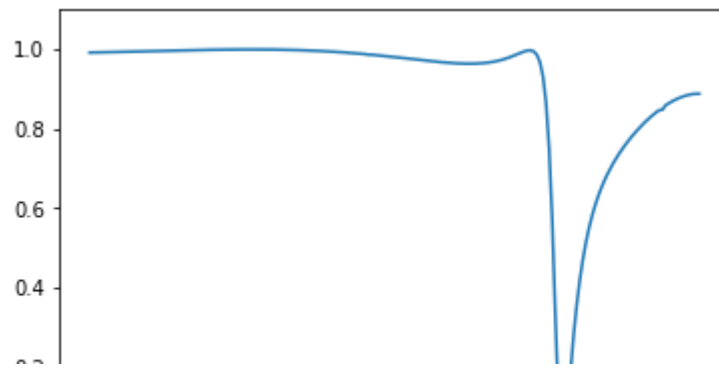


Predicted spectrum:

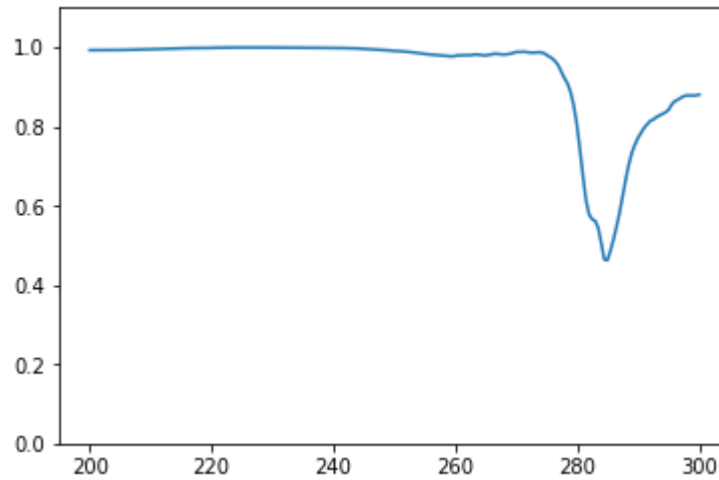


Test 28

True spectrum:

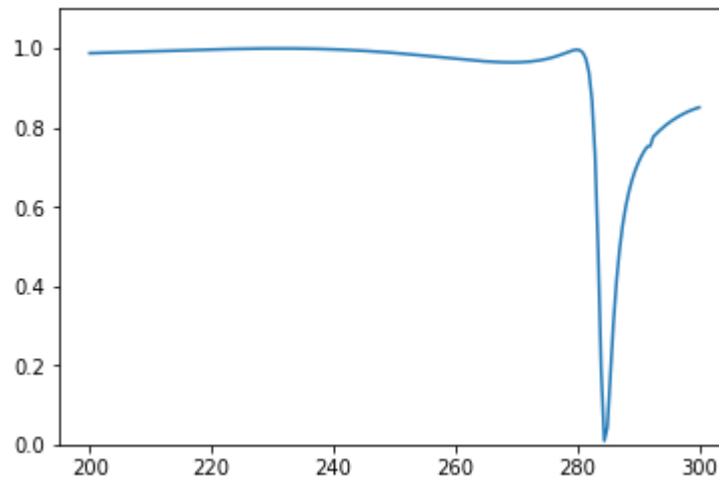


Predicted spectrum:

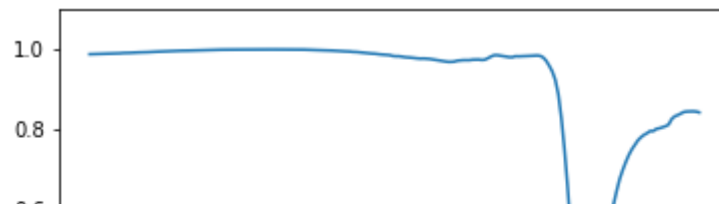


Test 29

True spectrum:

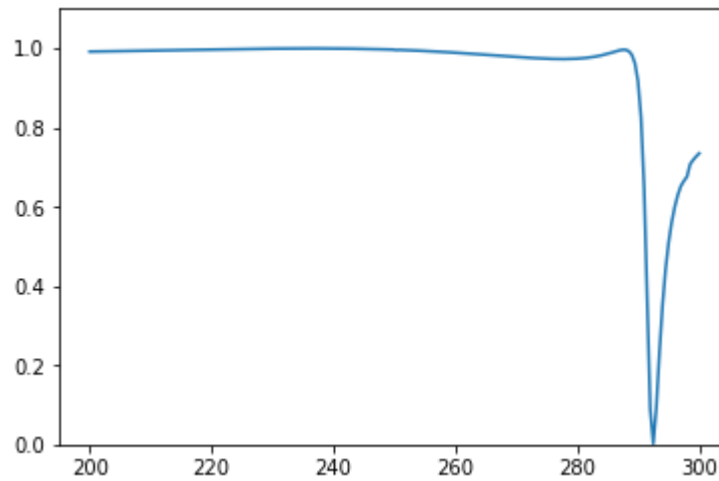


Predicted spectrum:

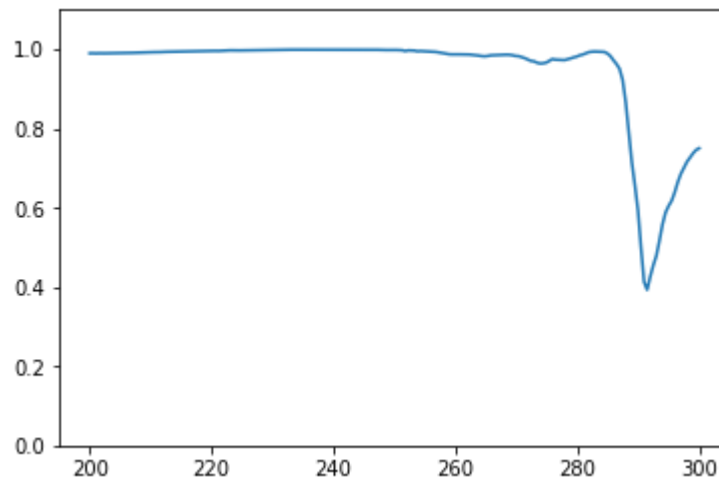


Test 30

True spectrum:

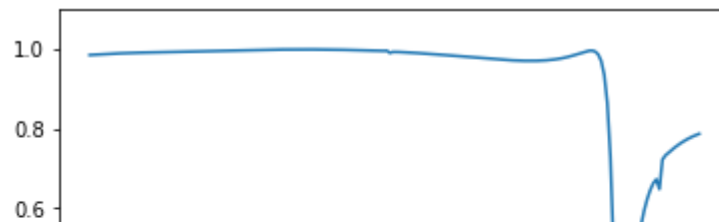


Predicted spectrum:

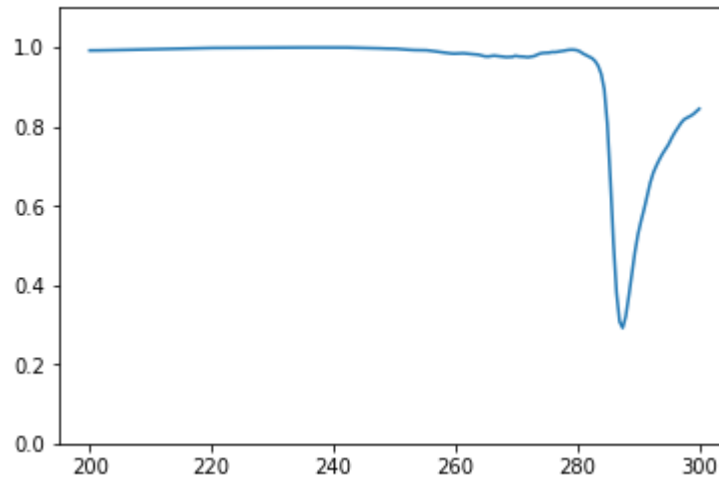


Test 31

True spectrum:

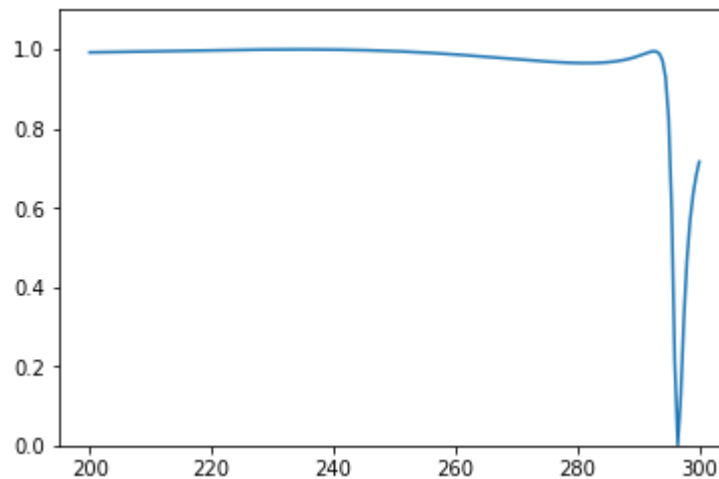


Predicted spectrum:

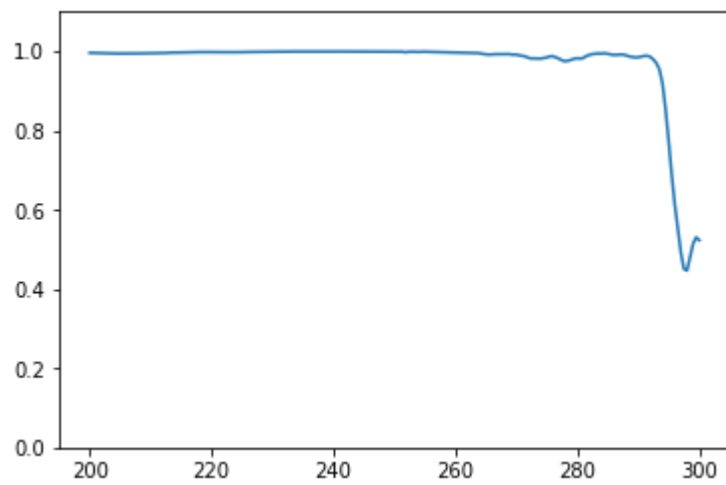


Test 32

True spectrum:

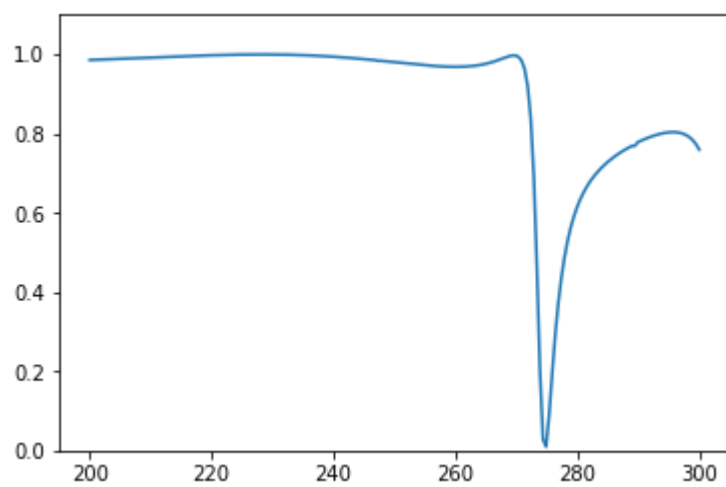


Predicted spectrum:

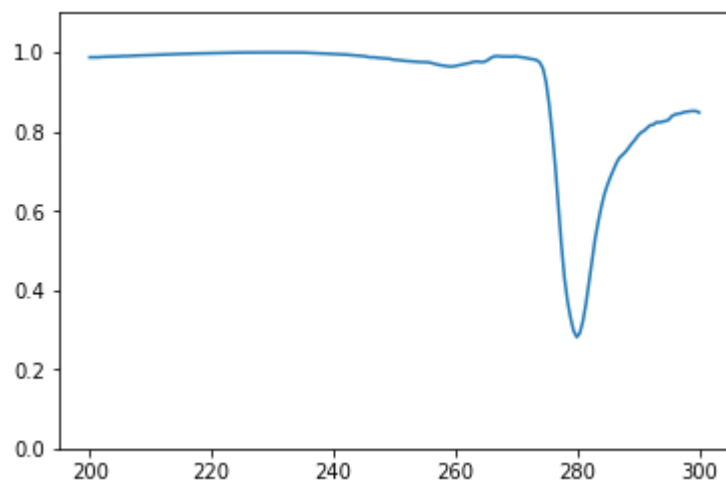


Test 33

True spectrum:

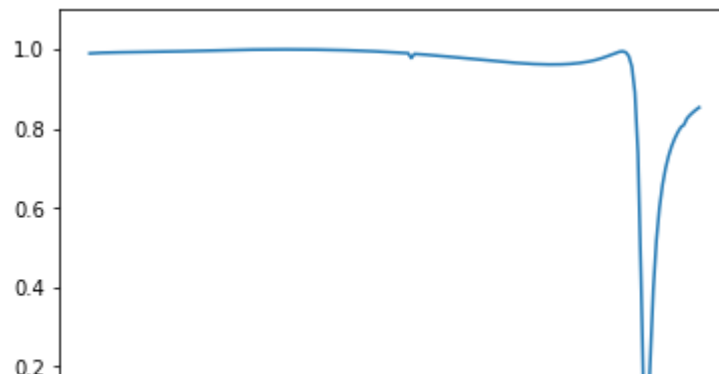


Predicted spectrum:

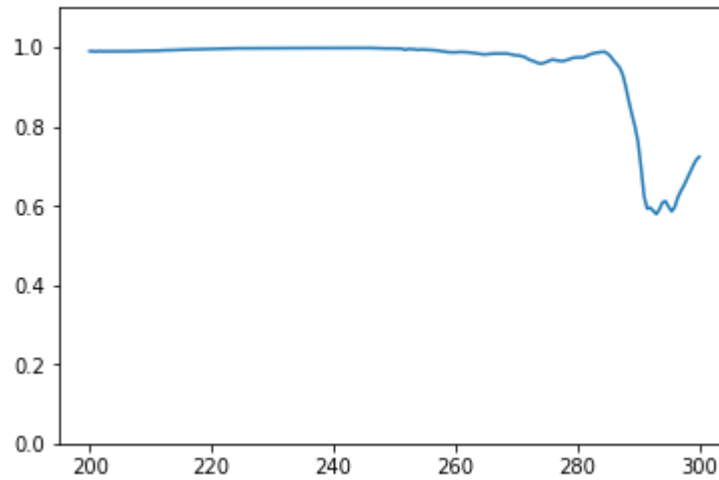


Test 34

True spectrum:

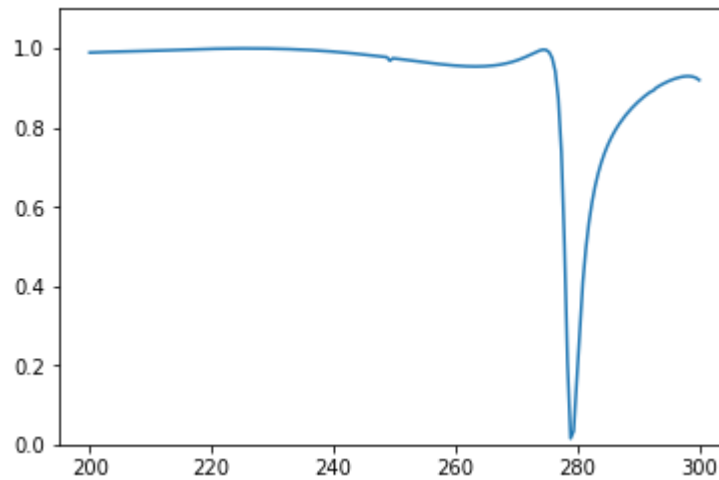


Predicted spectrum:

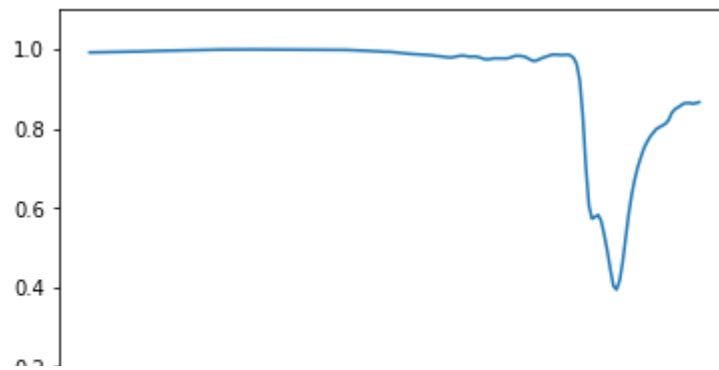


Test 35

True spectrum:

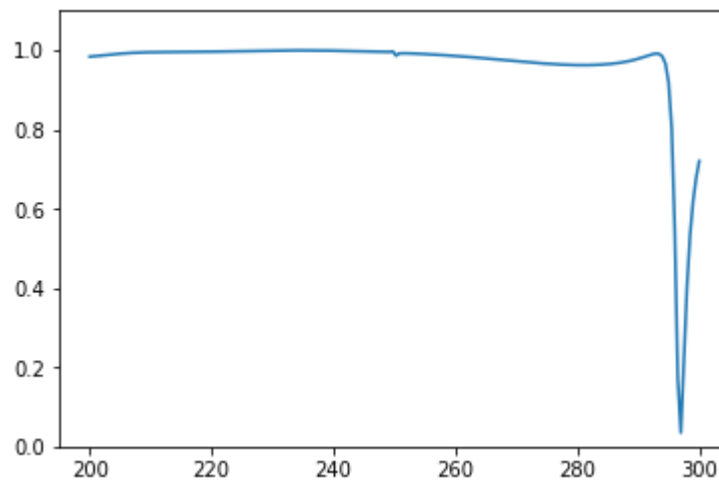


Predicted spectrum:

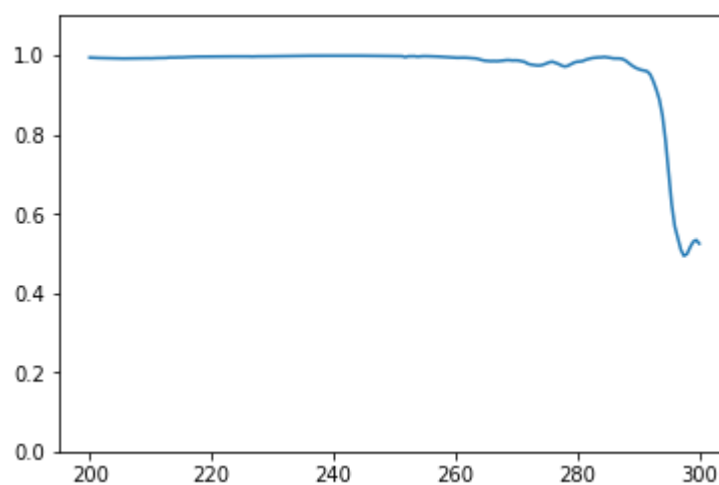


Test 36

True spectrum:

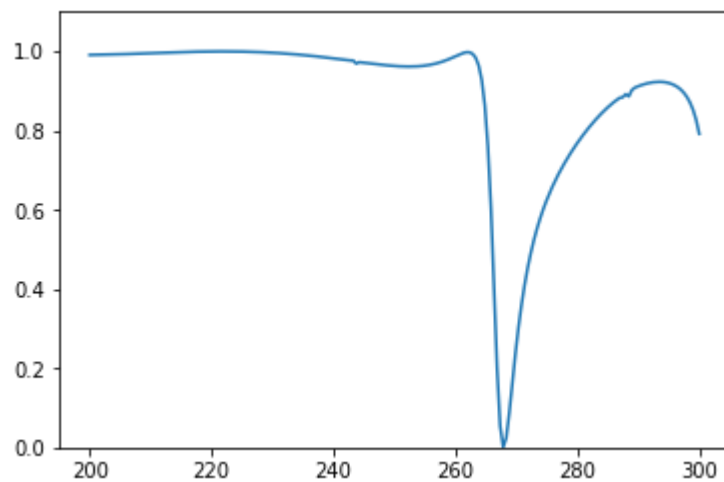


Predicted spectrum:

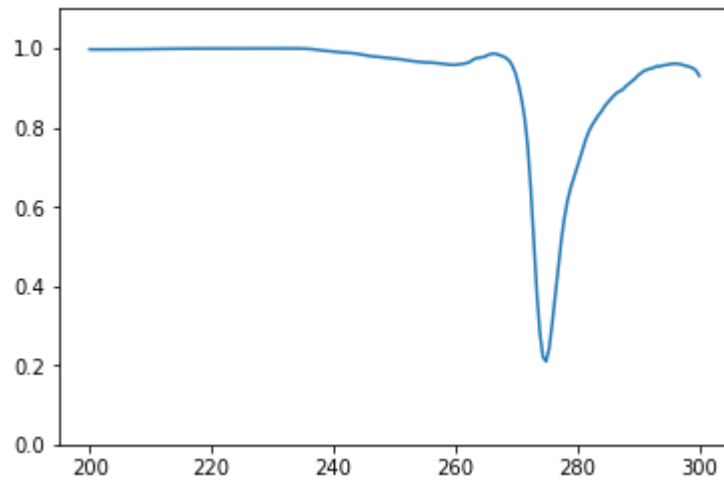


Test 37

True spectrum:

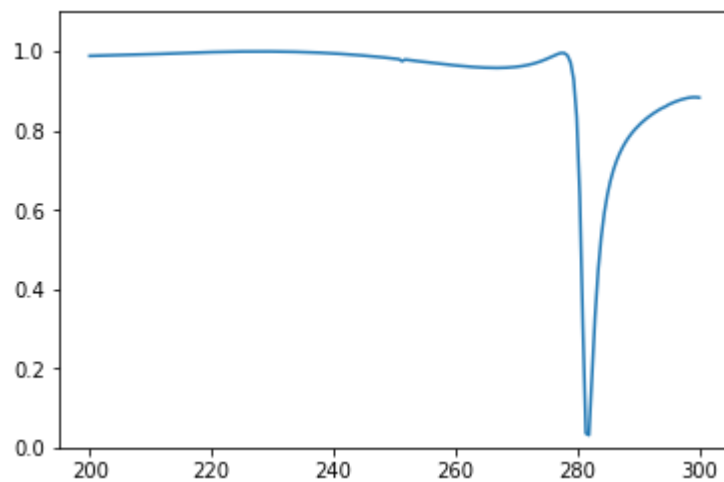


Predicted spectrum:

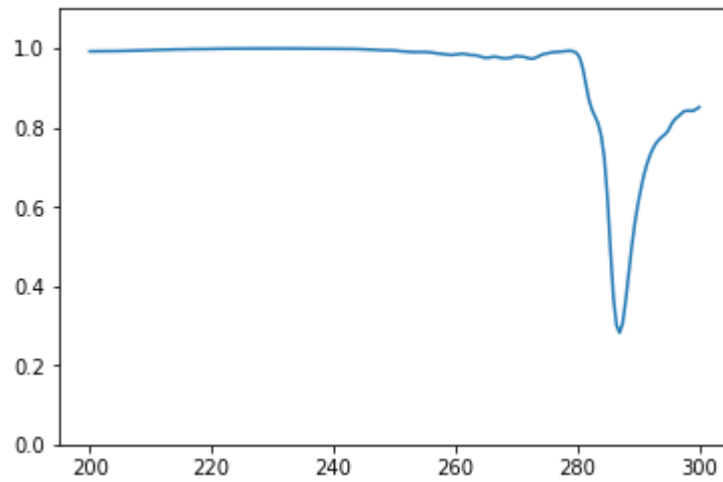


Test 38

True spectrum:

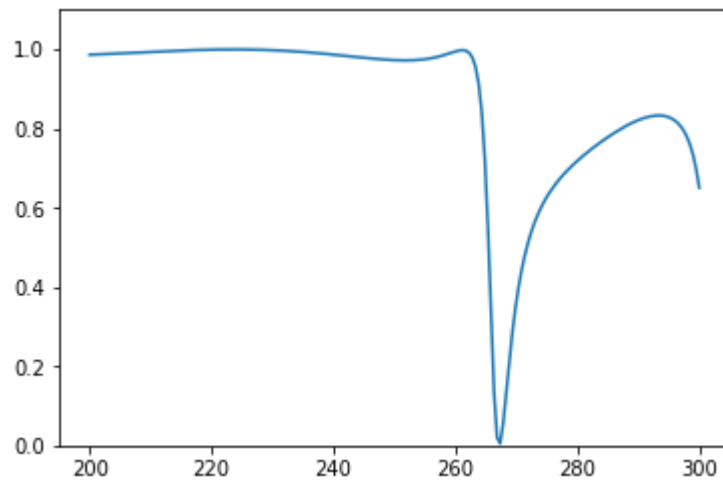


Predicted spectrum:

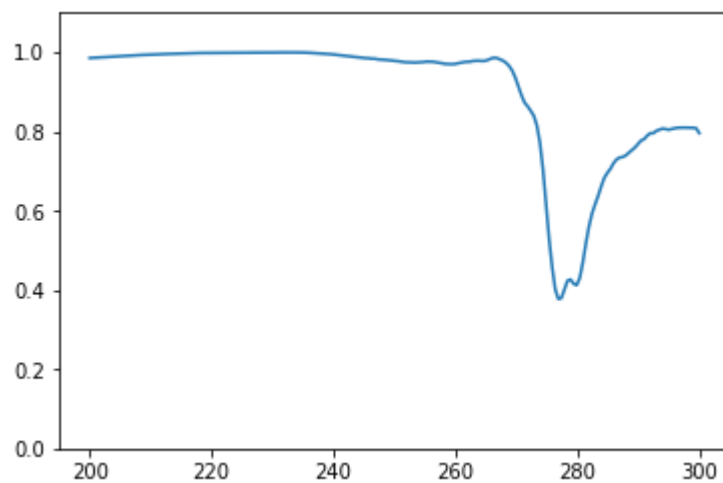


Test 39

True spectrum:

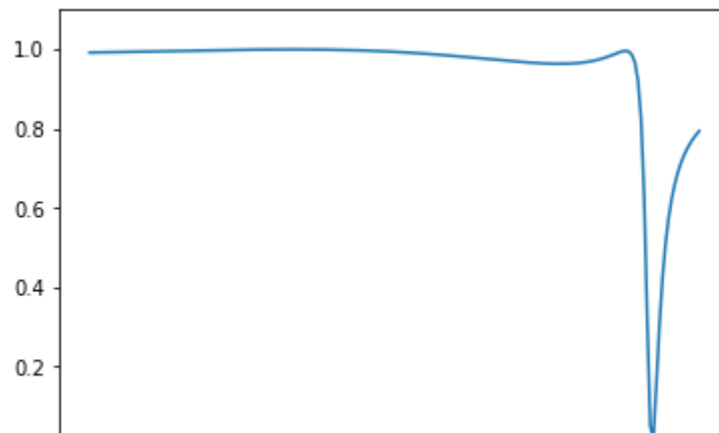


Predicted spectrum:

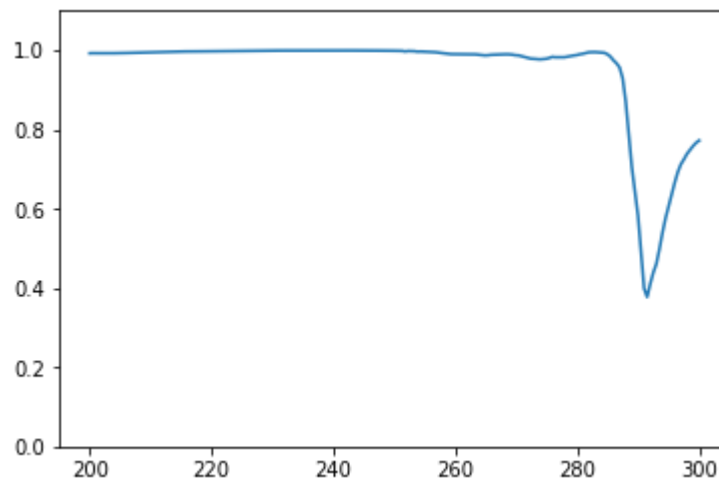


Test 40

True spectrum:

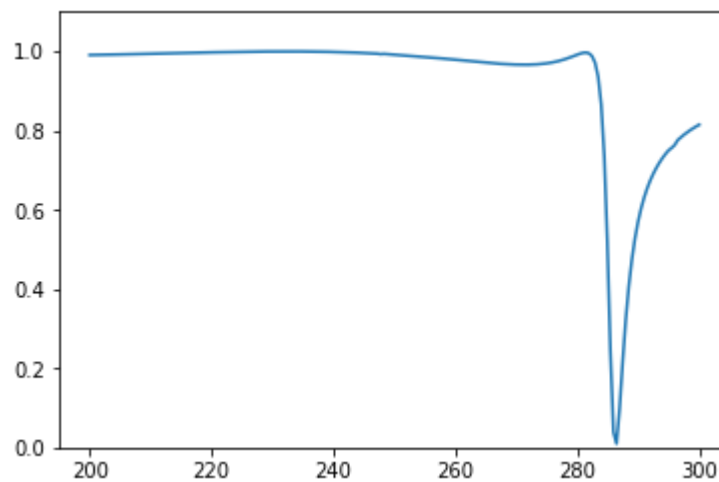


Predicted spectrum:

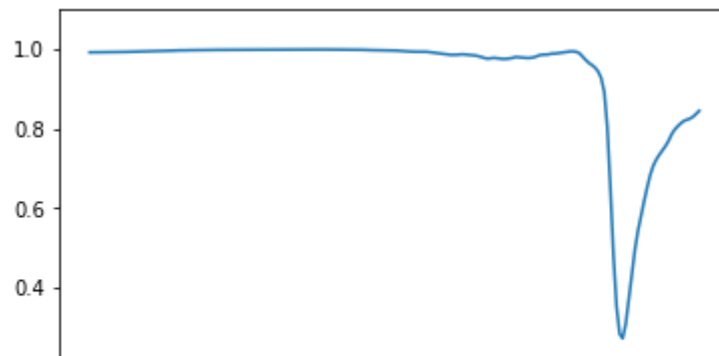


Test 41

True spectrum:

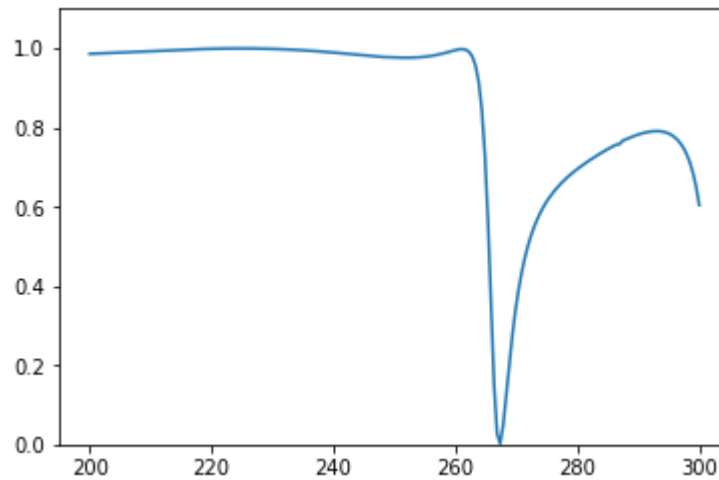


Predicted spectrum:

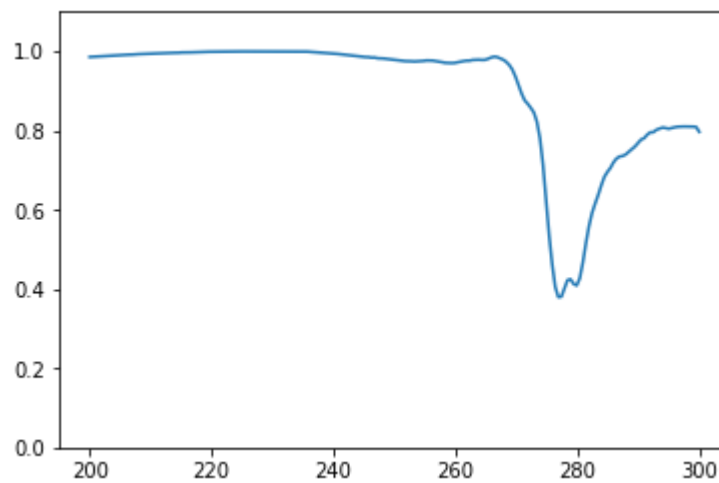


Test 42

True spectrum:

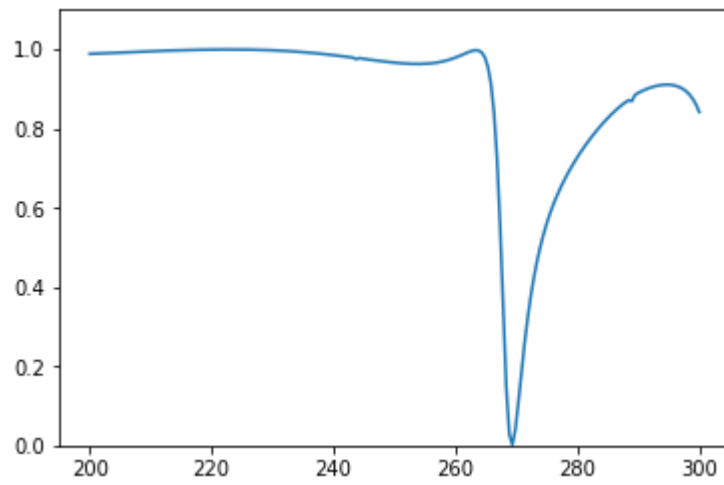


Predicted spectrum:

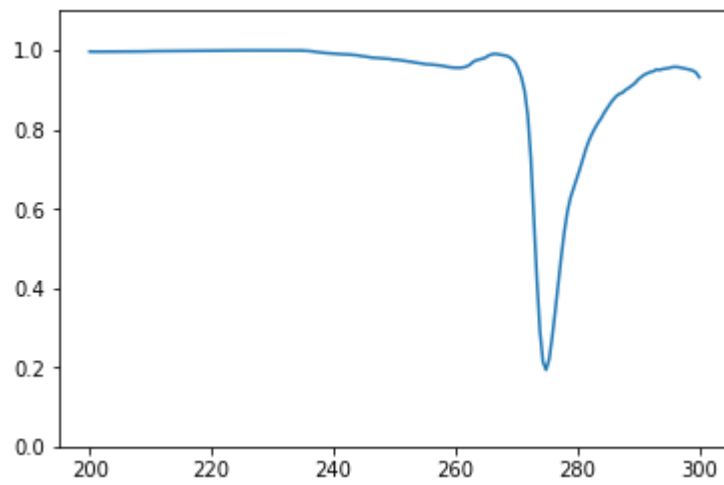


Test 43

True spectrum:

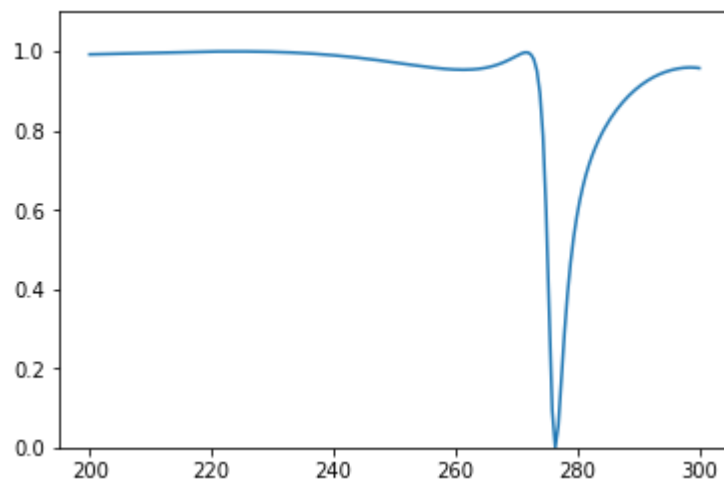


Predicted spectrum:

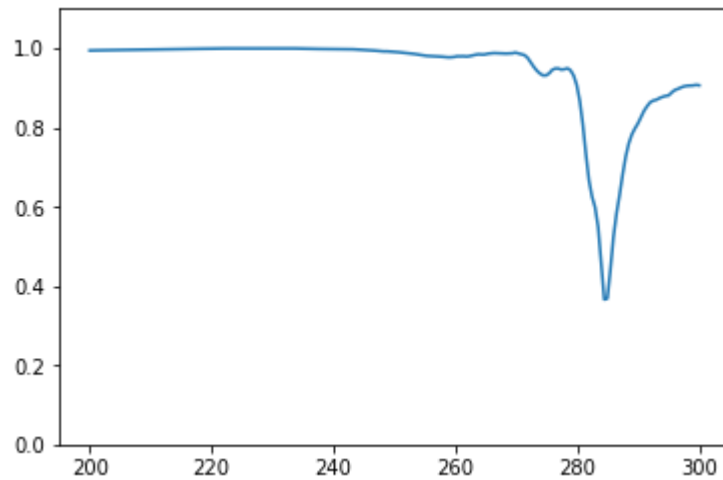


Test 44

True spectrum:

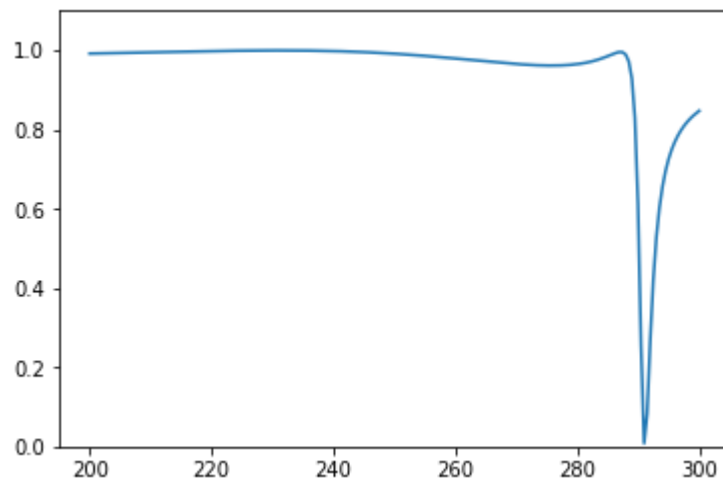


Predicted spectrum:

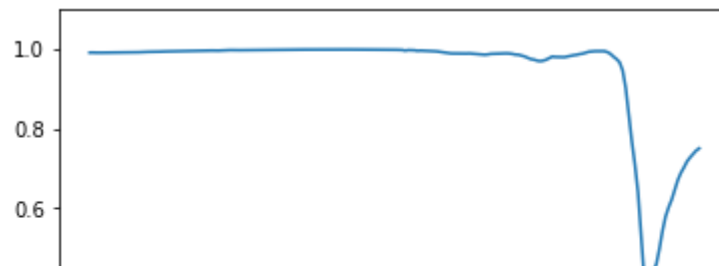


Test 45

True spectrum:

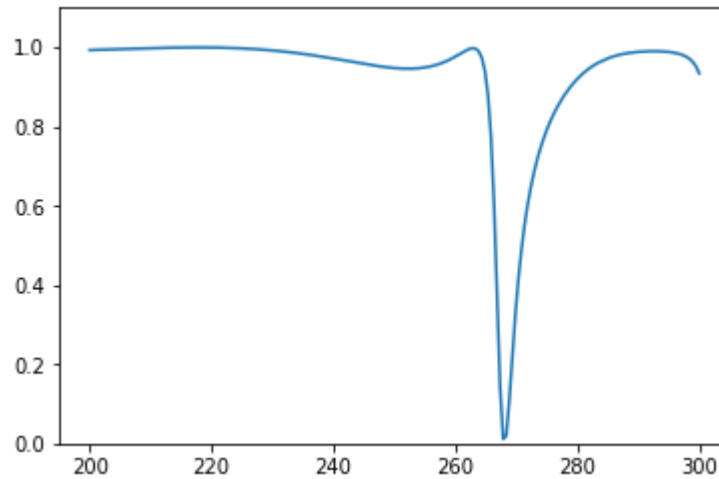


Predicted spectrum:

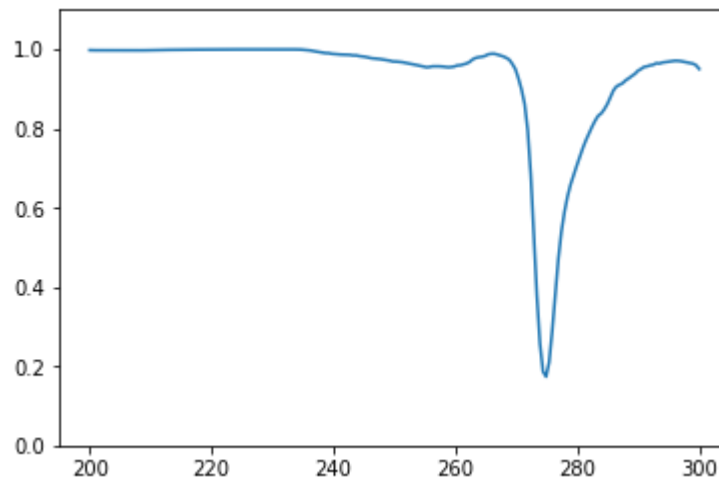


Test 46

True spectrum:

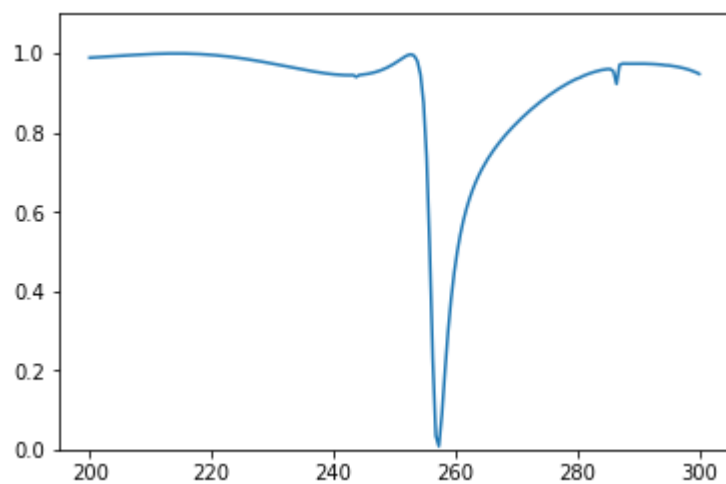


Predicted spectrum:

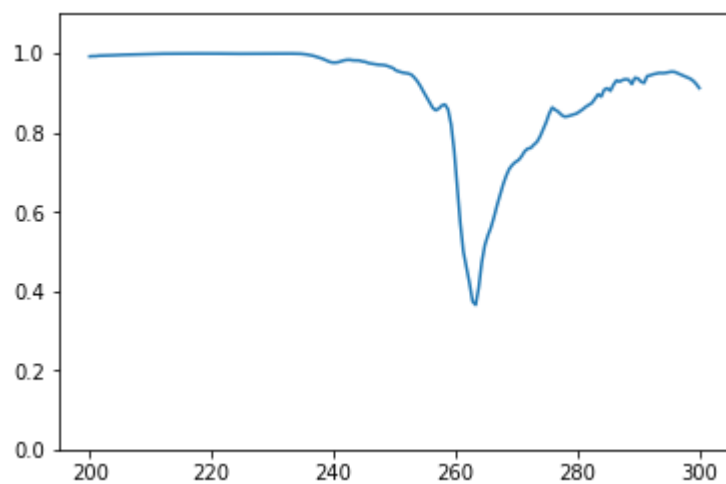


Test 47

True spectrum:

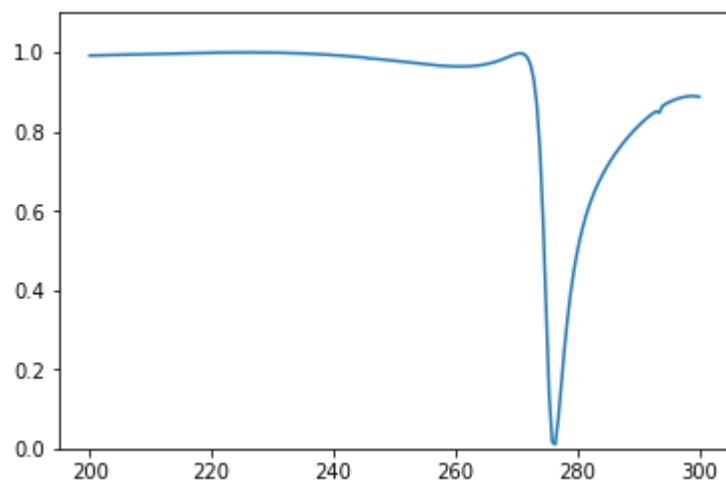


Predicted spectrum:

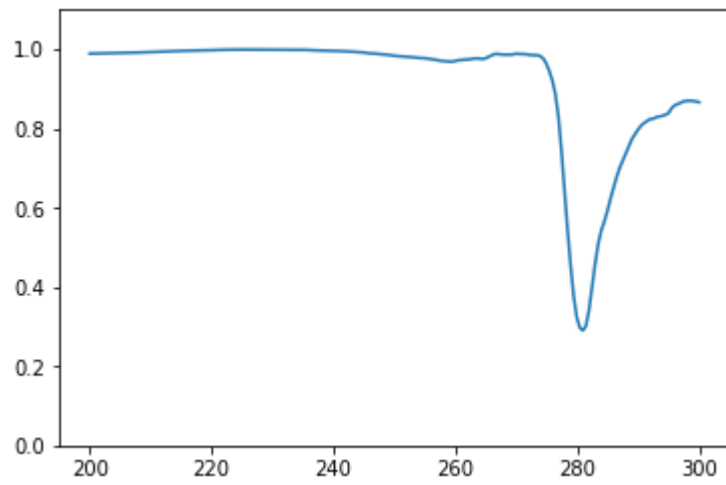


Test 48

True spectrum:

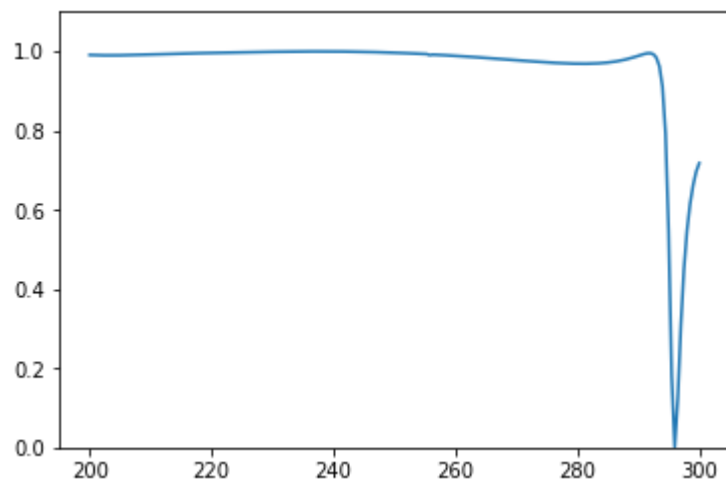


Predicted spectrum:

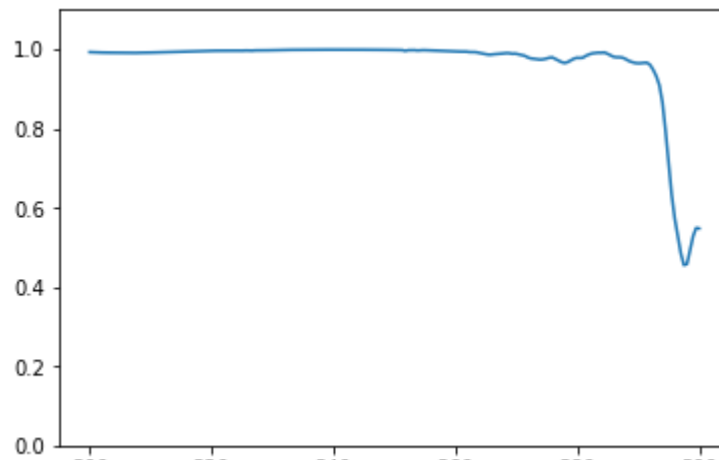


Test 49

True spectrum:

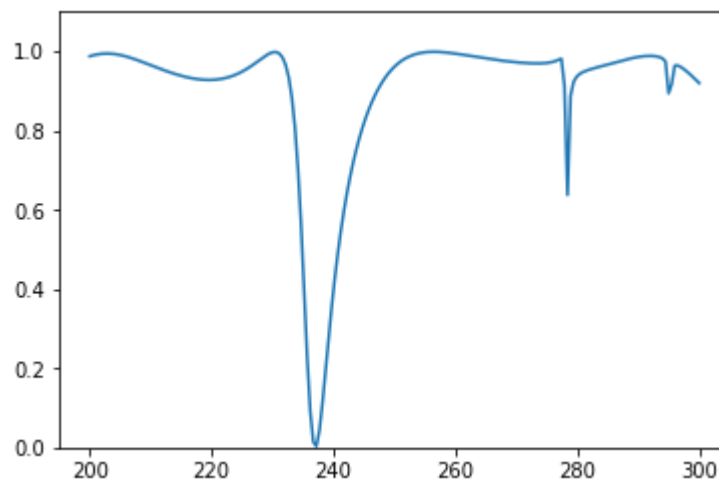


Predicted spectrum:

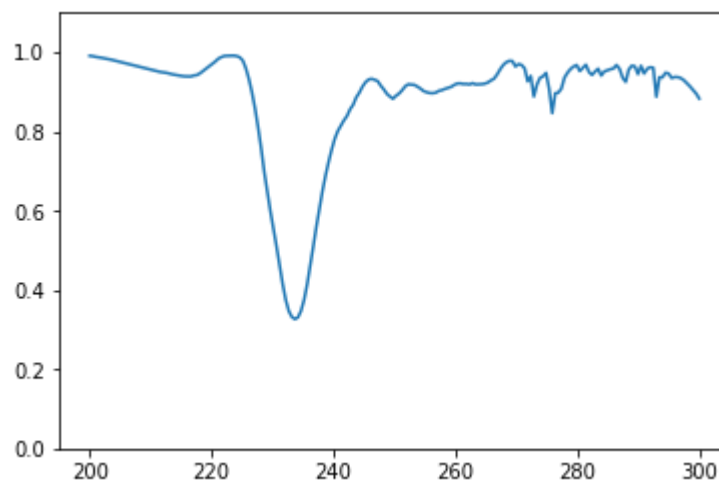


Test 50

True spectrum:

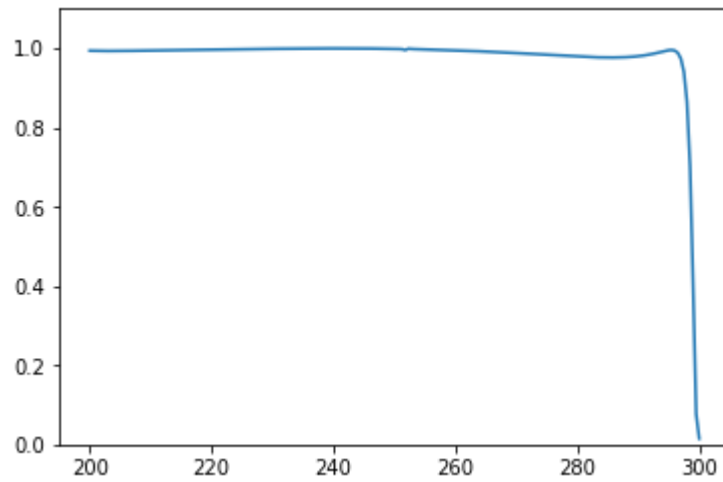


Predicted spectrum:

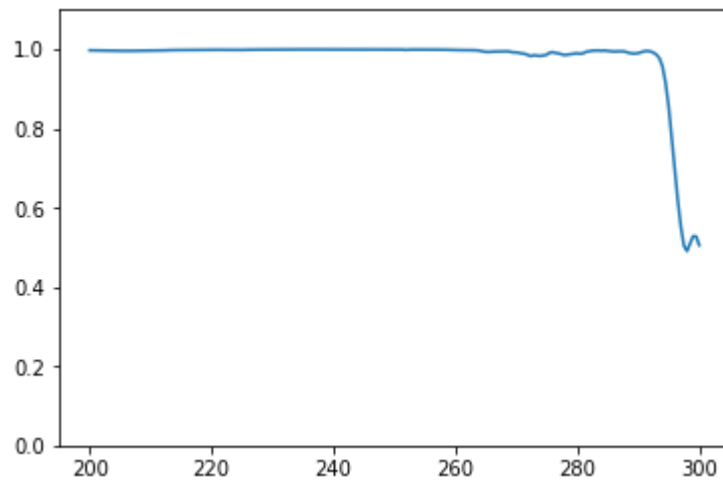


Test 51

True spectrum:



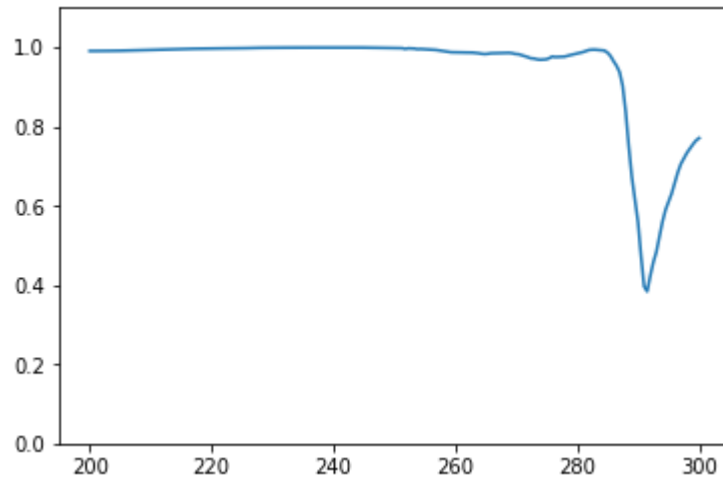
Predicted spectrum:



Test 52

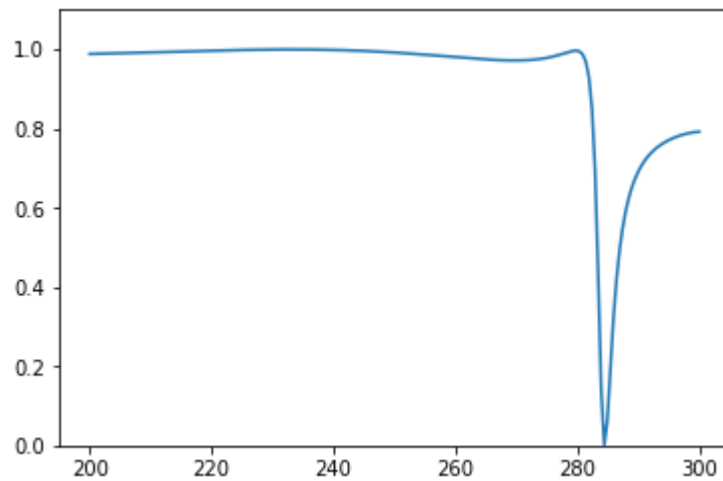
True spectrum:

Predicted spectrum:

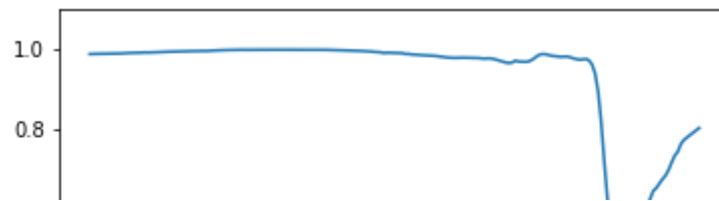


Test 53

True spectrum:

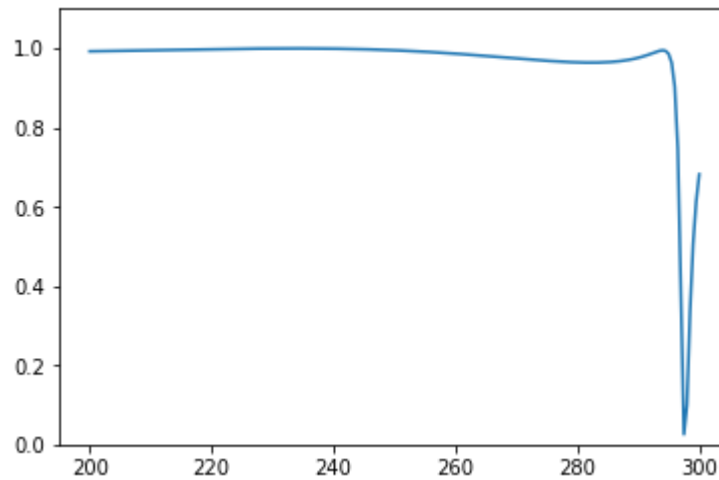


Predicted spectrum:

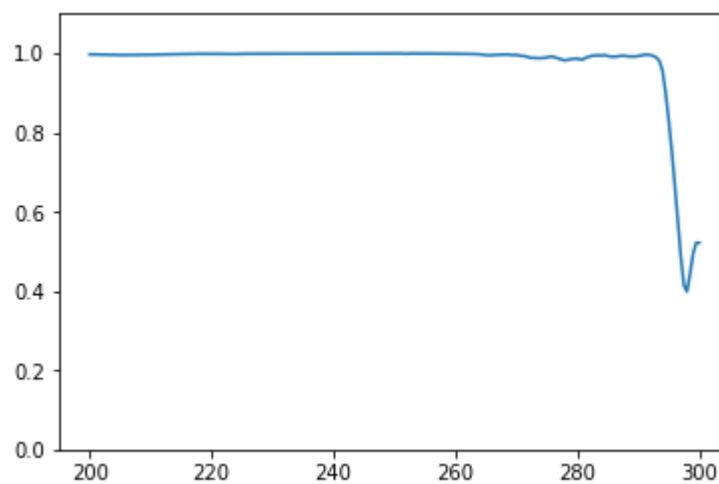


Test 54

True spectrum:

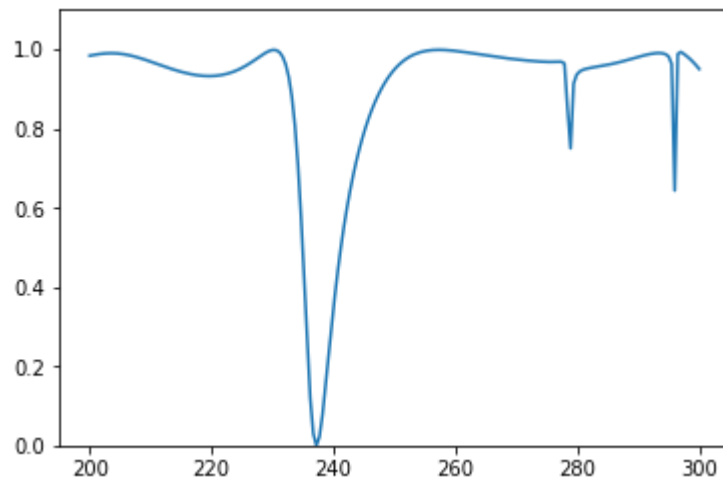


Predicted spectrum:

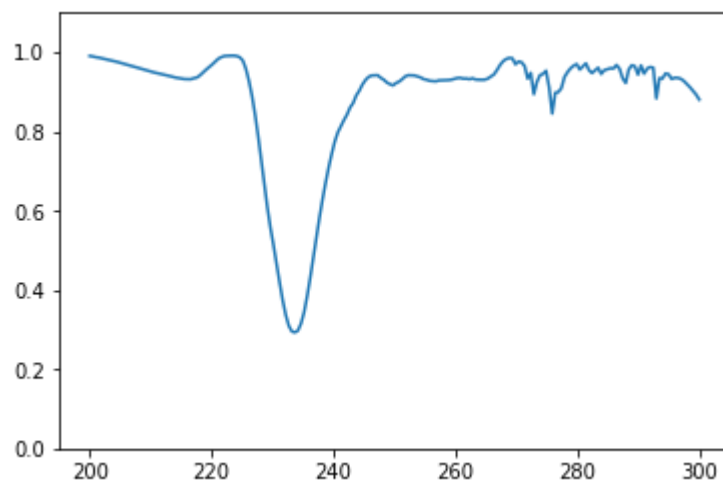


Test 55

True spectrum:

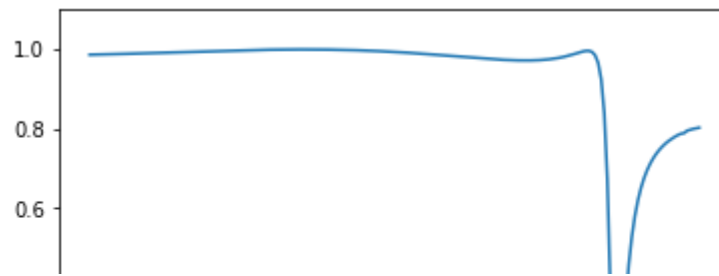


Predicted spectrum:

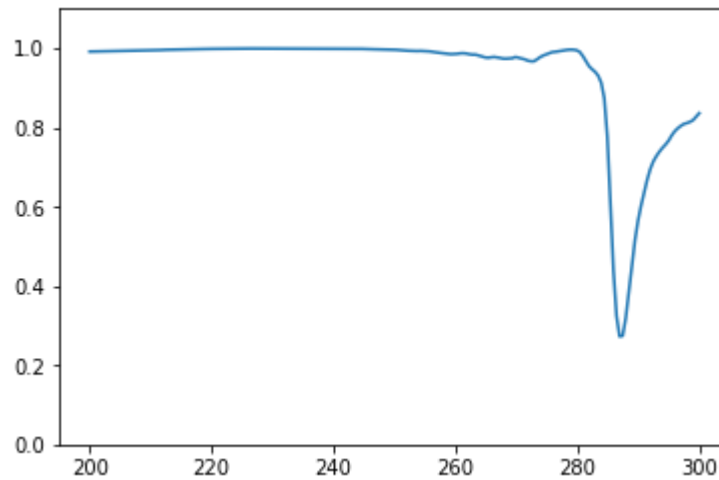


Test 56

True spectrum:

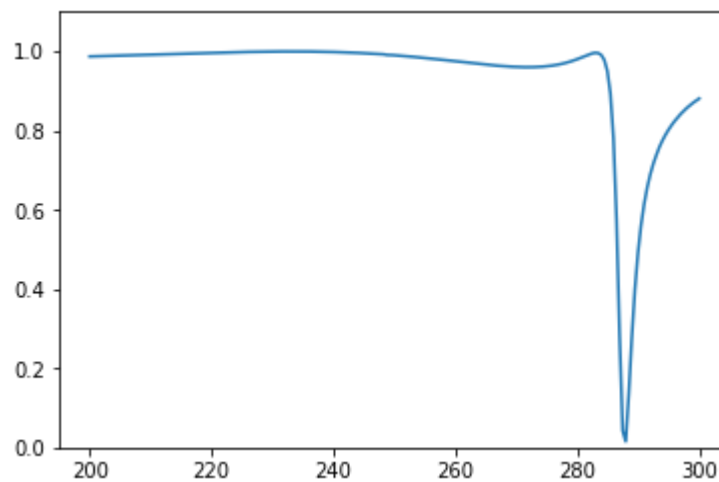


Predicted spectrum:

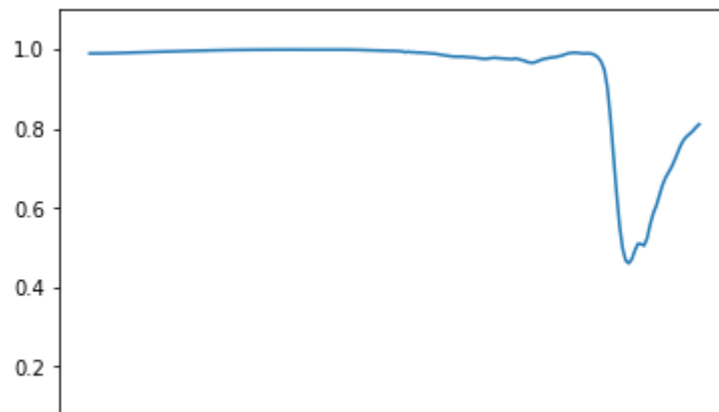


Test 57

True spectrum:

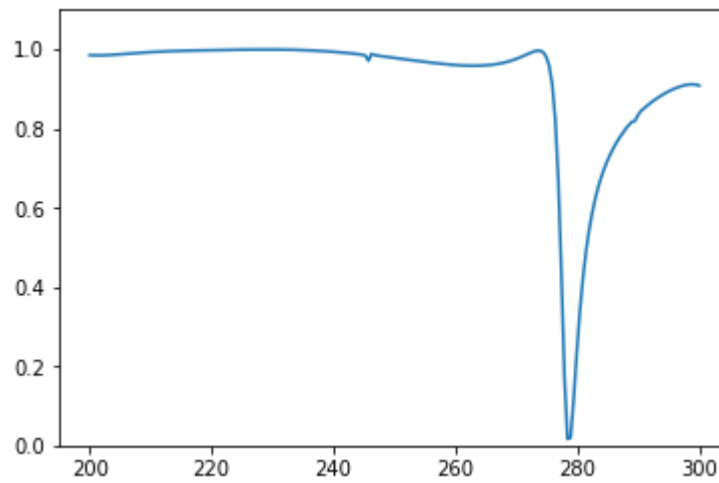


Predicted spectrum:

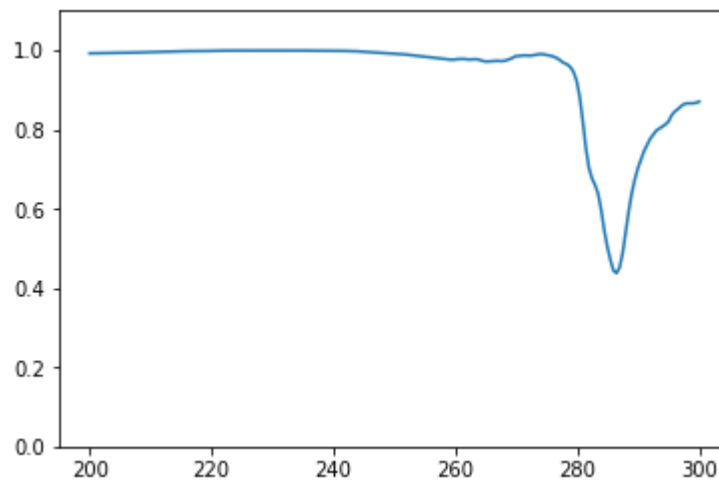


Test 58

True spectrum:

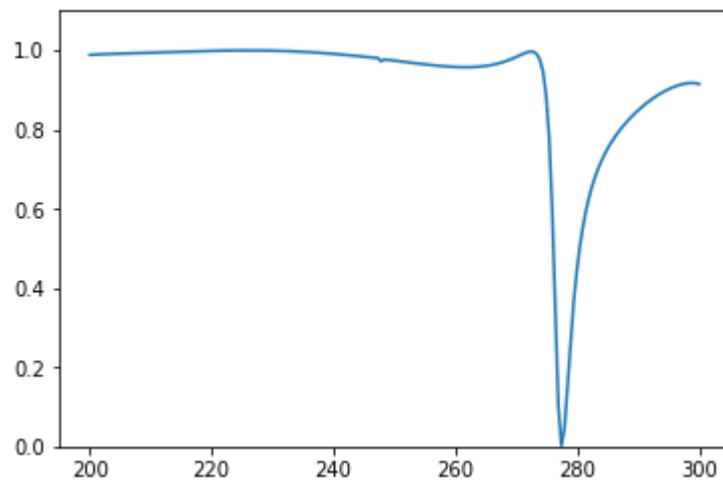


Predicted spectrum:

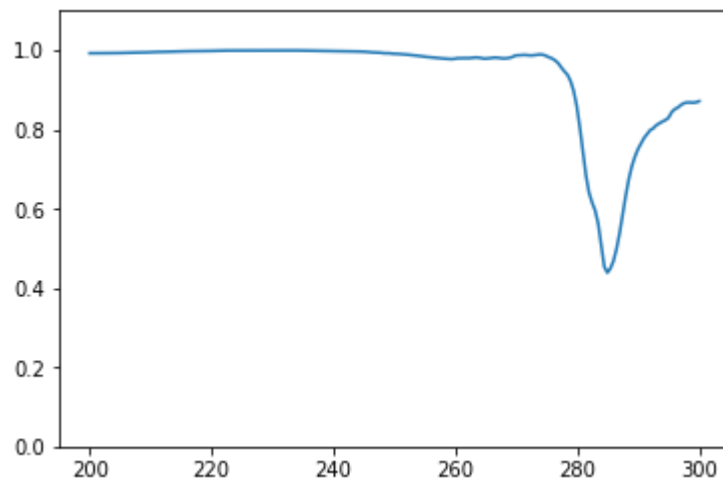


Test 59

True spectrum:

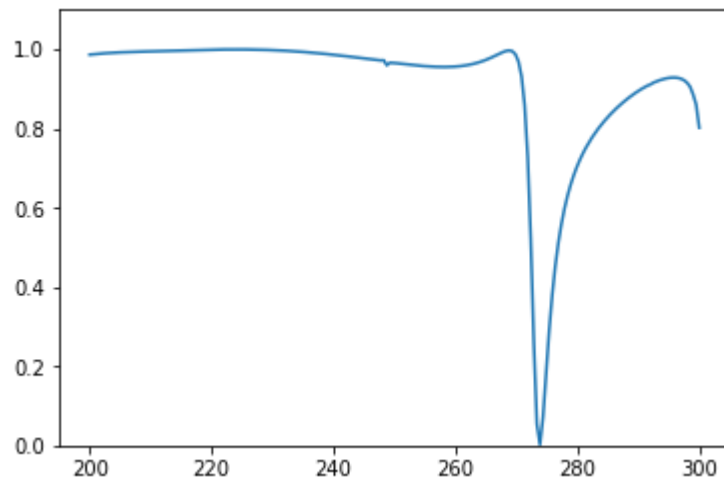


Predicted spectrum:

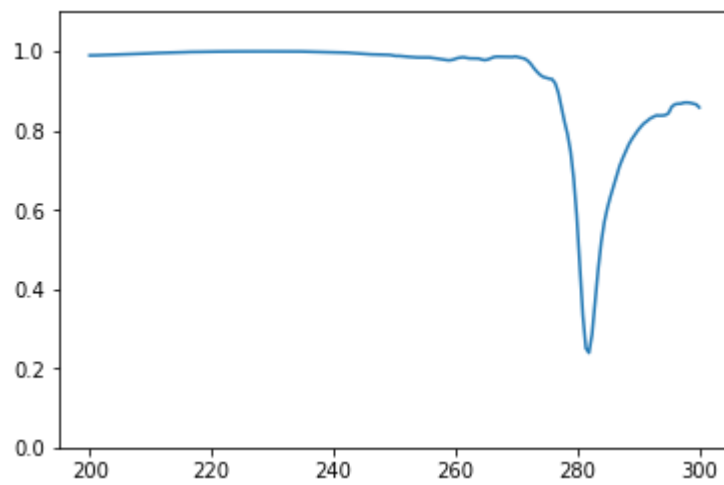


Test 60

True spectrum:

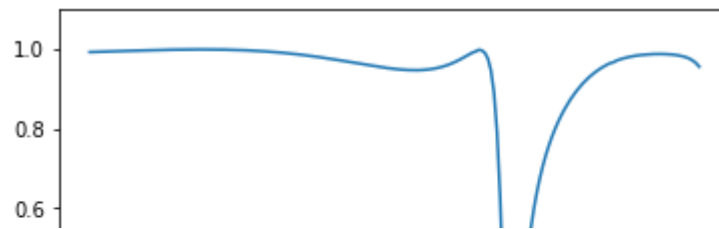


Predicted spectrum:

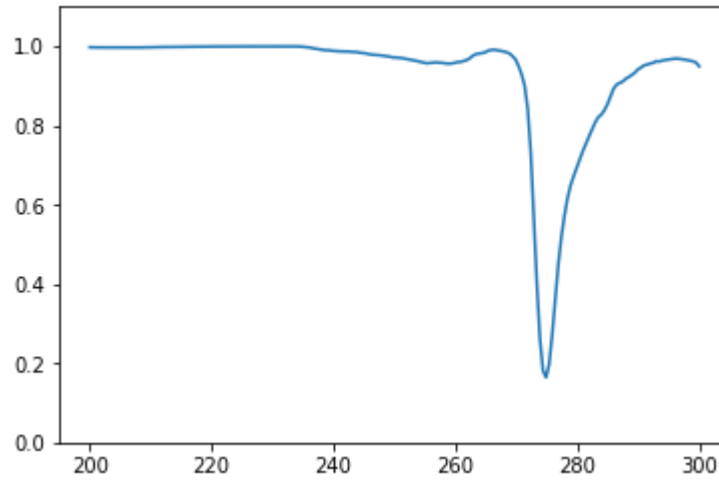


Test 61

True spectrum:

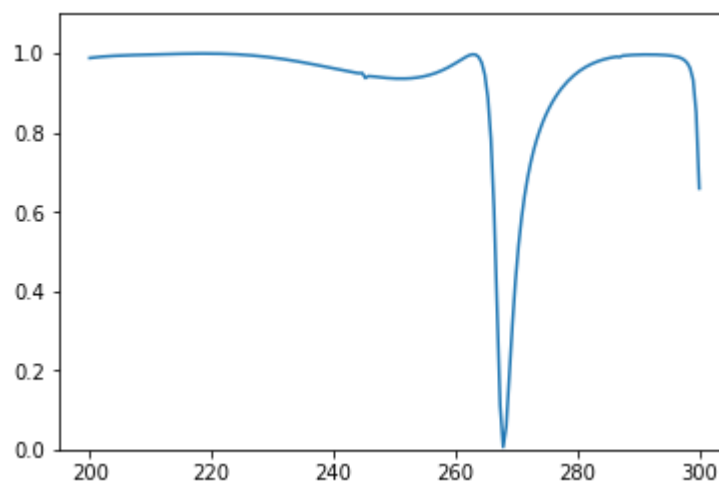


Predicted spectrum:

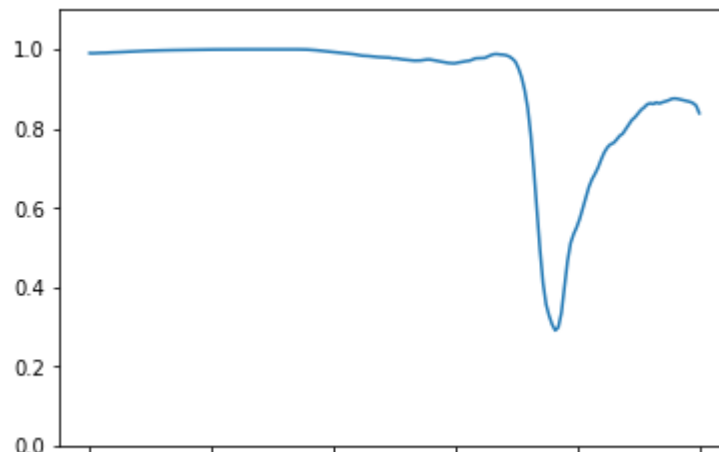


Test 62

True spectrum:

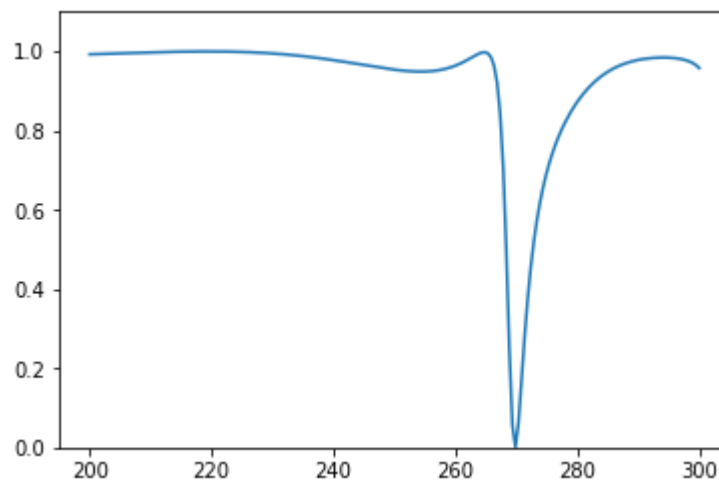


Predicted spectrum:



Test 63

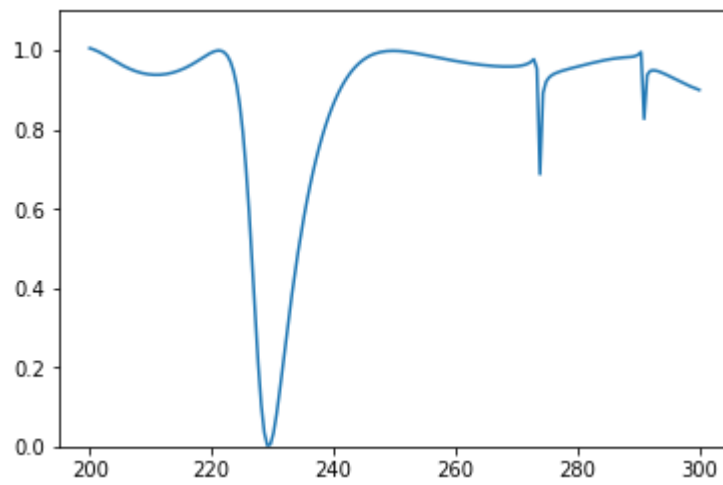
True spectrum:



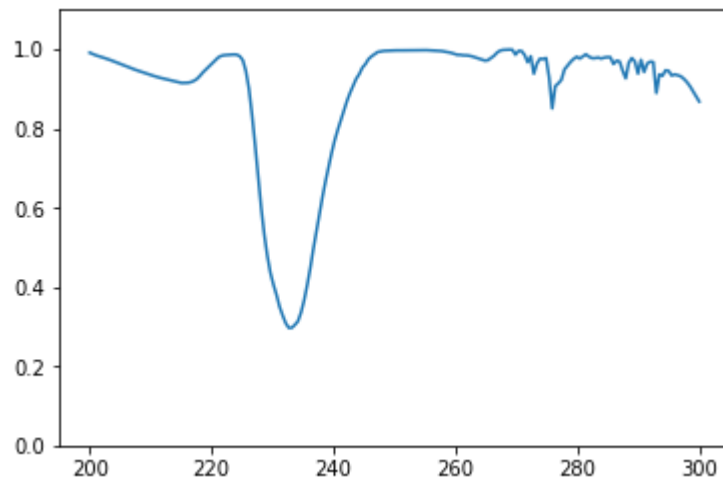
Predicted spectrum:

Test 64

True spectrum:

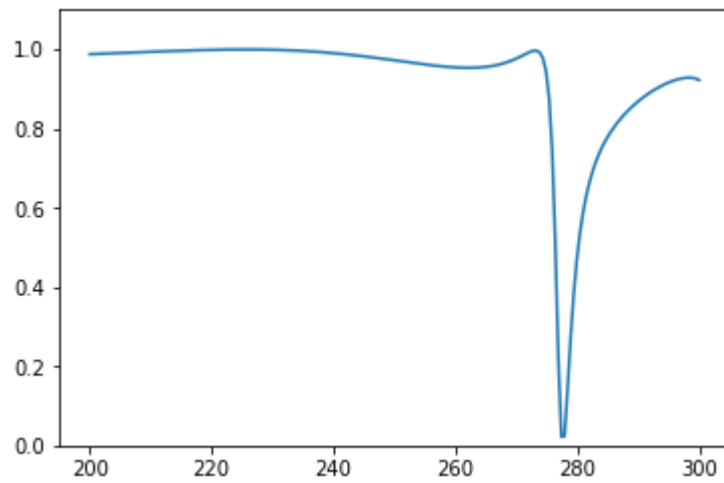


Predicted spectrum:

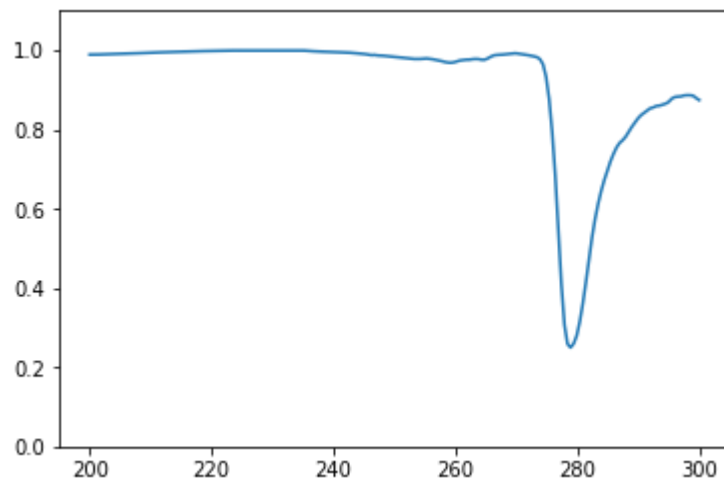


Test 65

True spectrum:

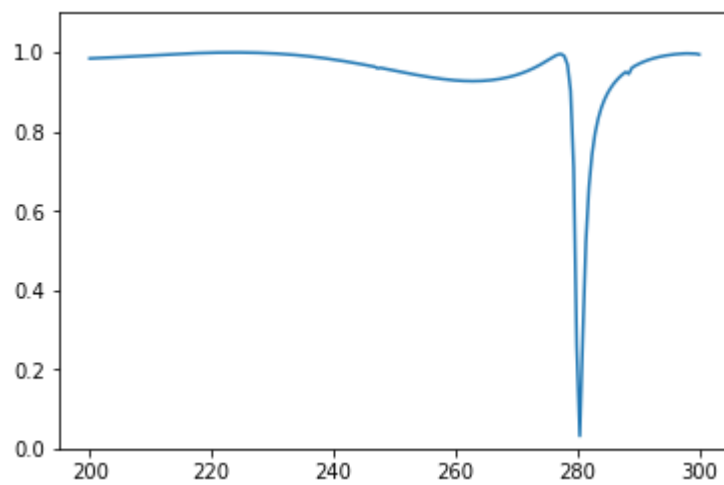


Predicted spectrum:

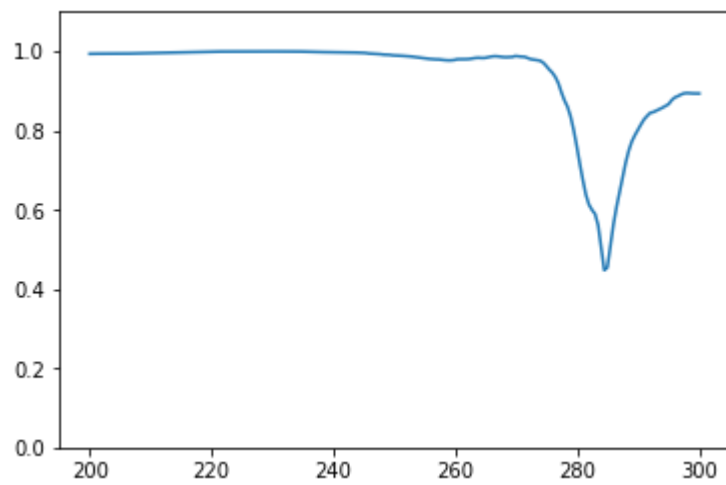


Test 66

True spectrum:

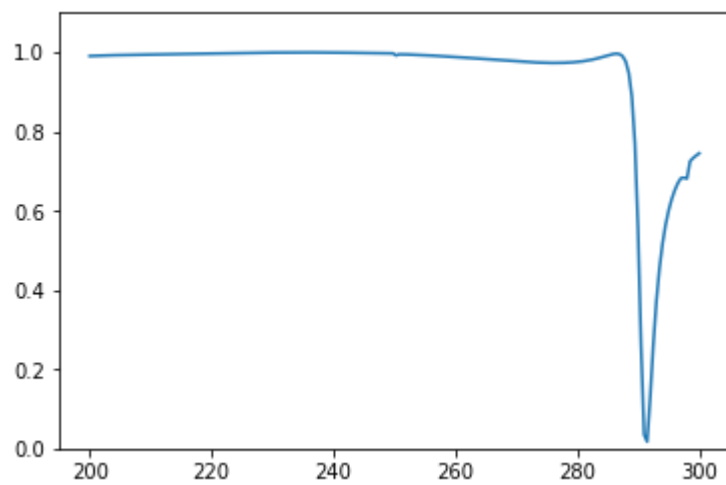


Predicted spectrum:

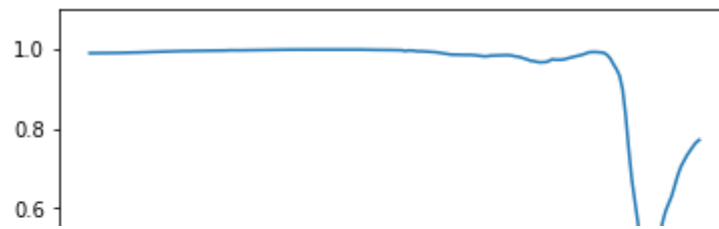


Test 67

True spectrum:

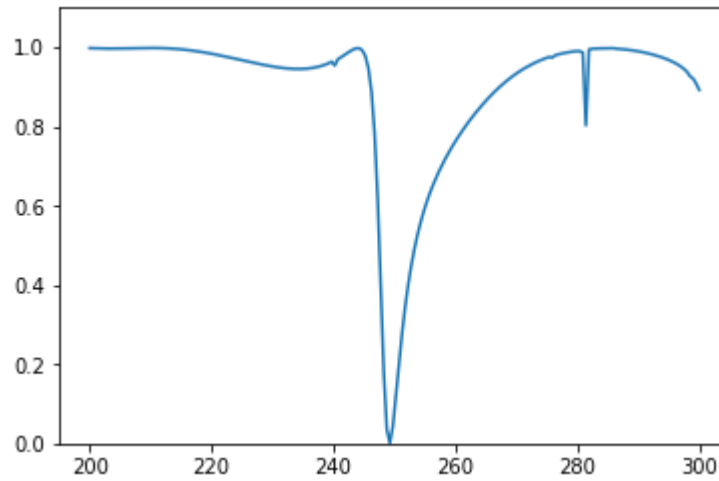


Predicted spectrum:

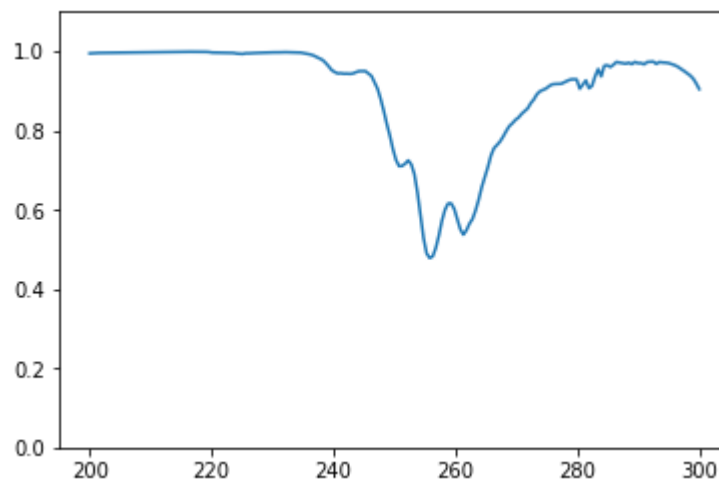


Test 68

True spectrum:

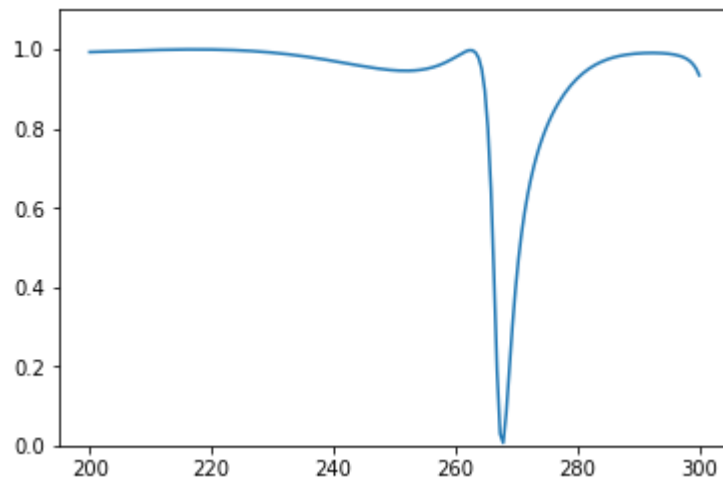


Predicted spectrum:

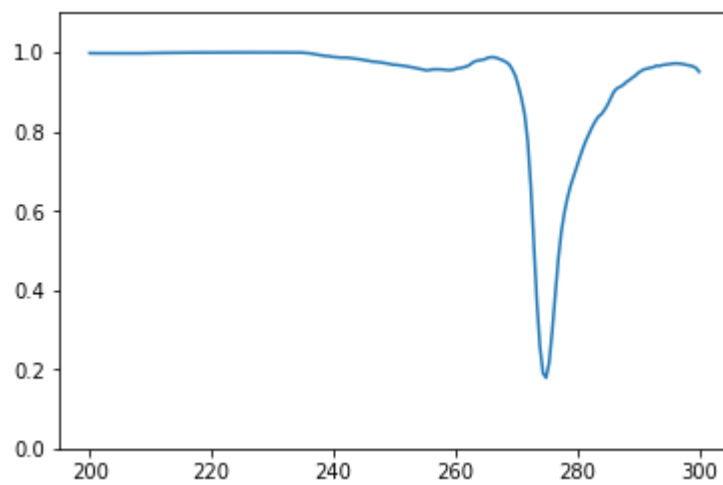


Test 69

True spectrum:

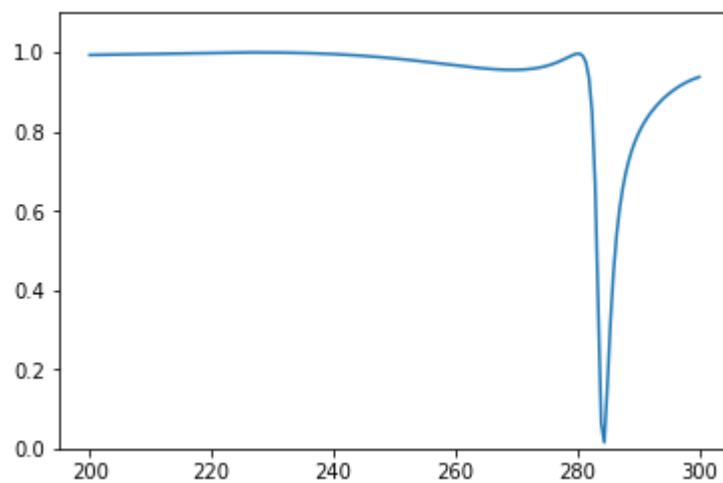


Predicted spectrum:

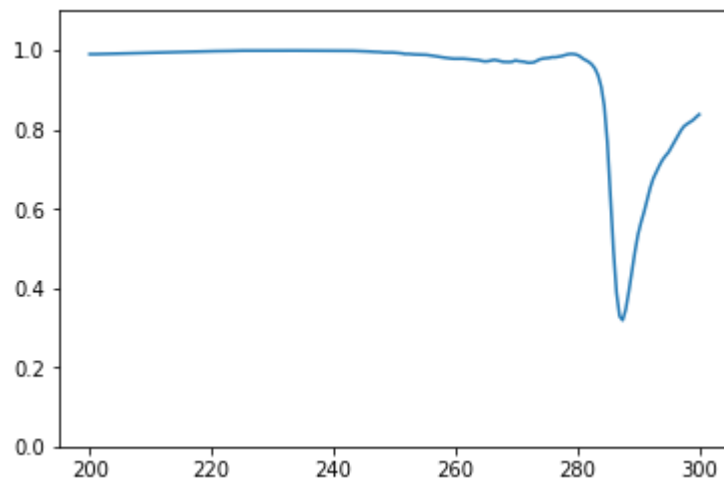


Test 70

True spectrum:

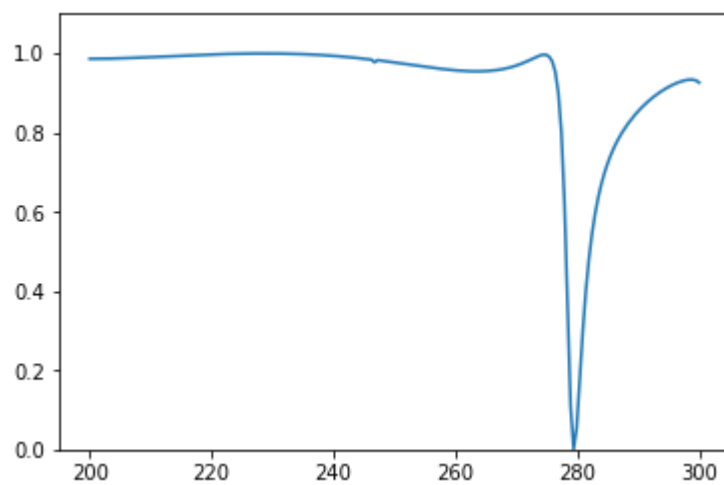


Predicted spectrum:

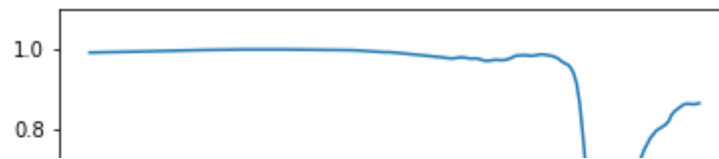


Test 71

True spectrum:

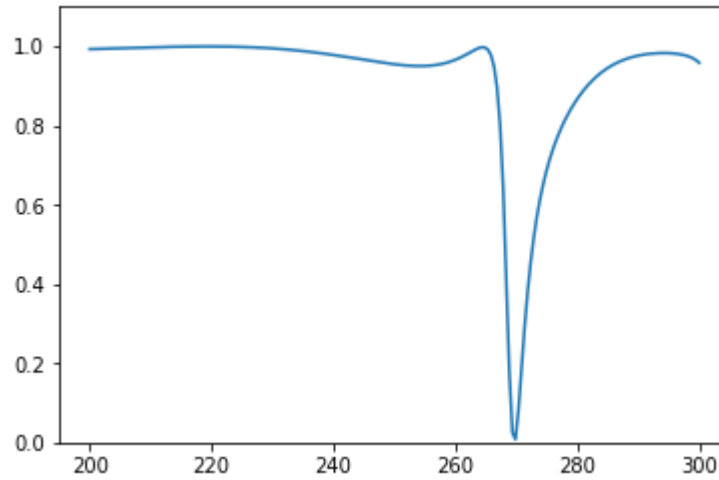


Predicted spectrum:

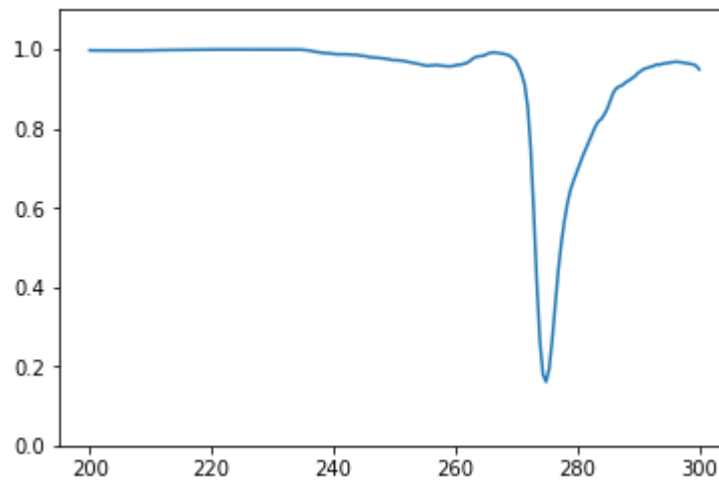


Test 72

True spectrum:

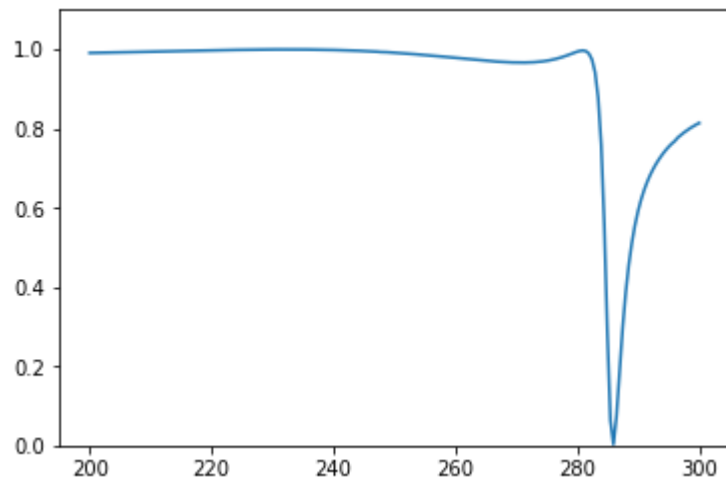


Predicted spectrum:

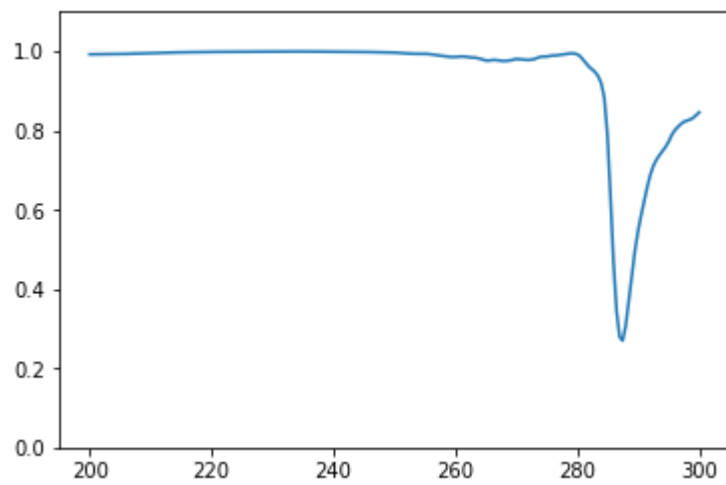


Test 73

True spectrum:



Predicted spectrum:

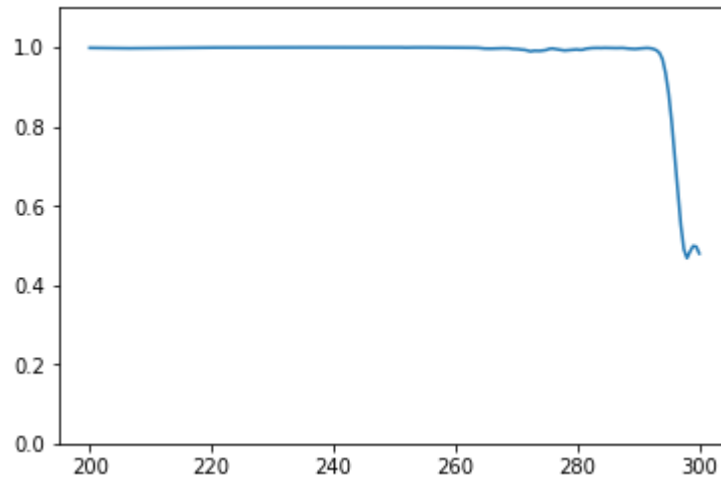


Test 74

True spectrum:

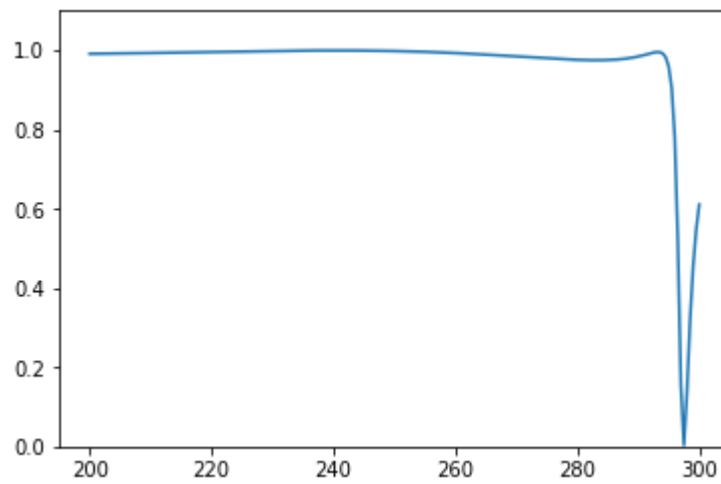


Predicted spectrum:

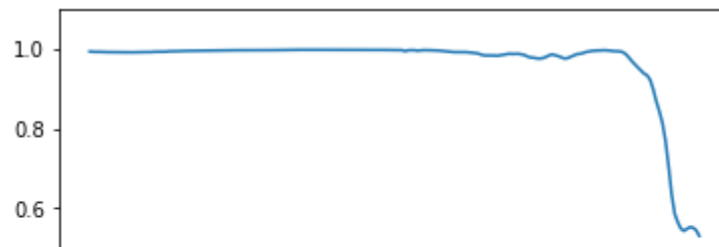


Test 75

True spectrum:

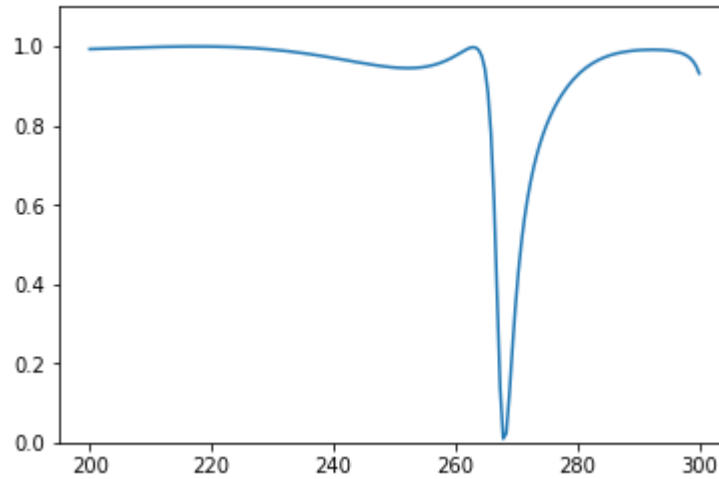


Predicted spectrum:

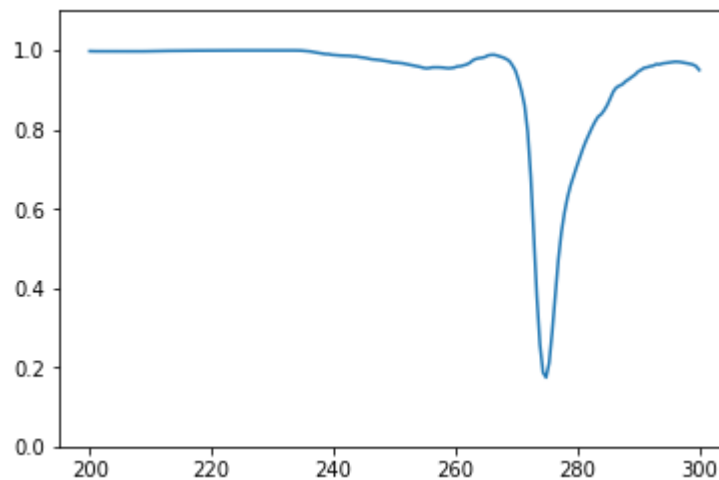


Test 76

True spectrum:

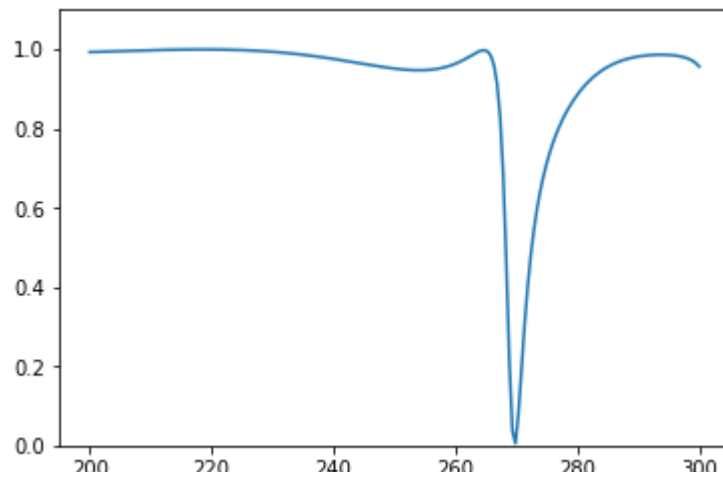


Predicted spectrum:

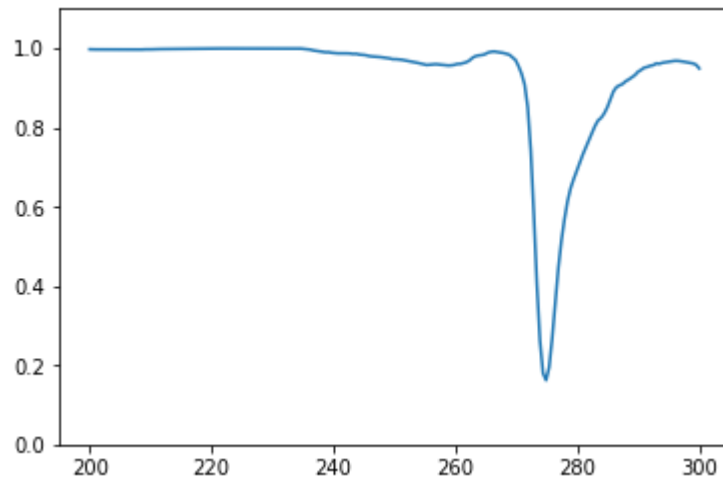


Test 77

True spectrum:

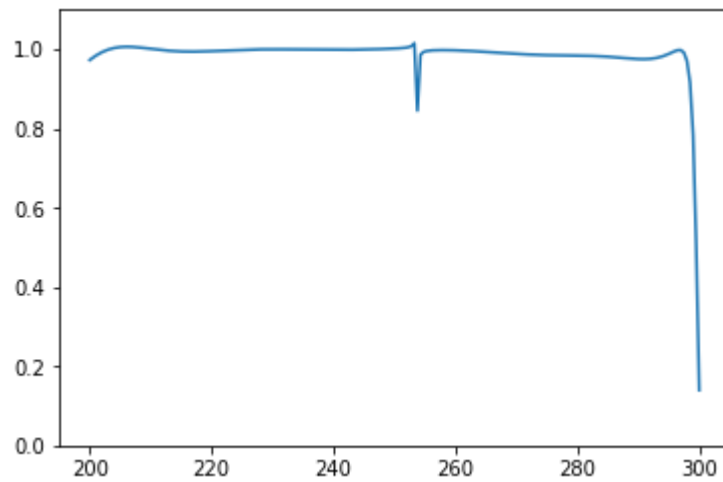


Predicted spectrum:

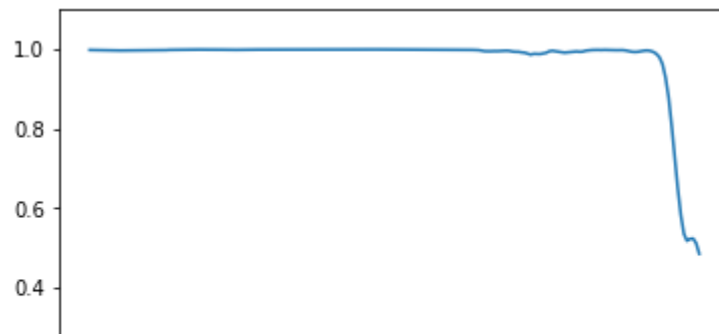


Test 78

True spectrum:

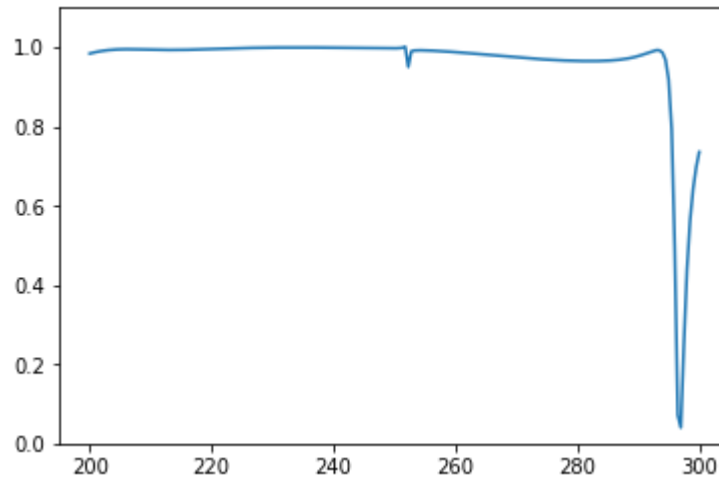


Predicted spectrum:

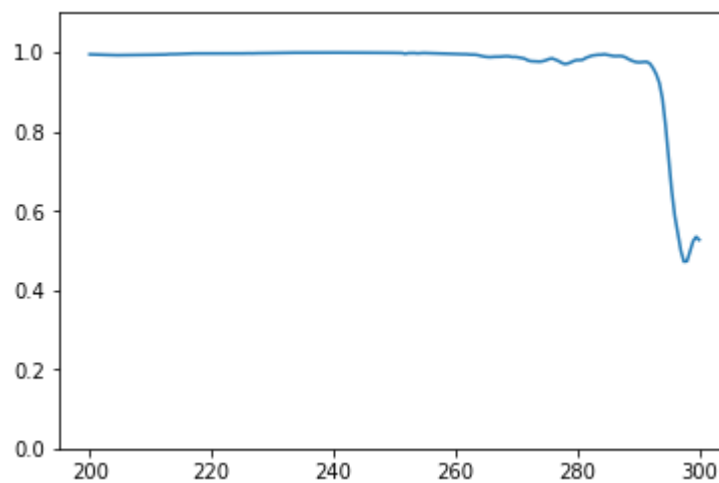


Test 79

True spectrum:

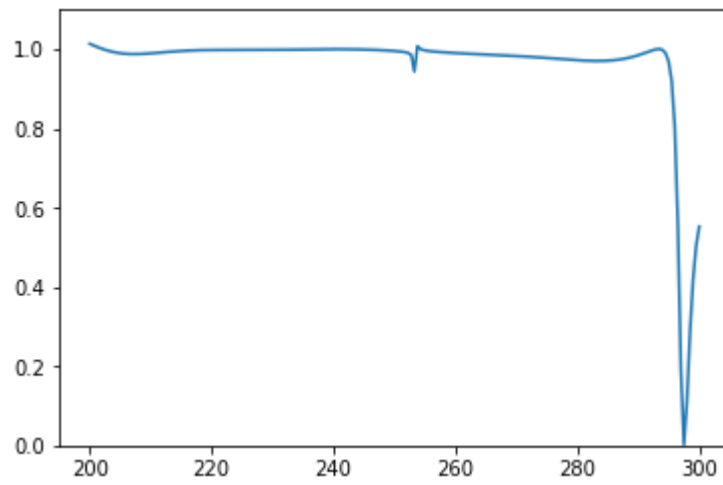


Predicted spectrum:

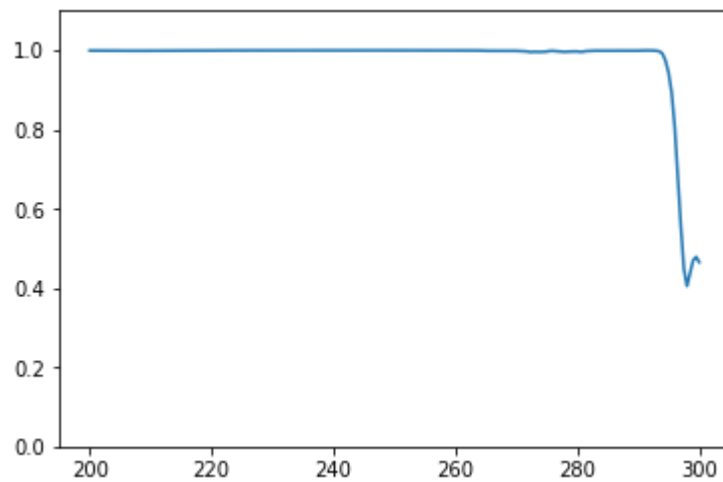


Test 80

True spectrum:

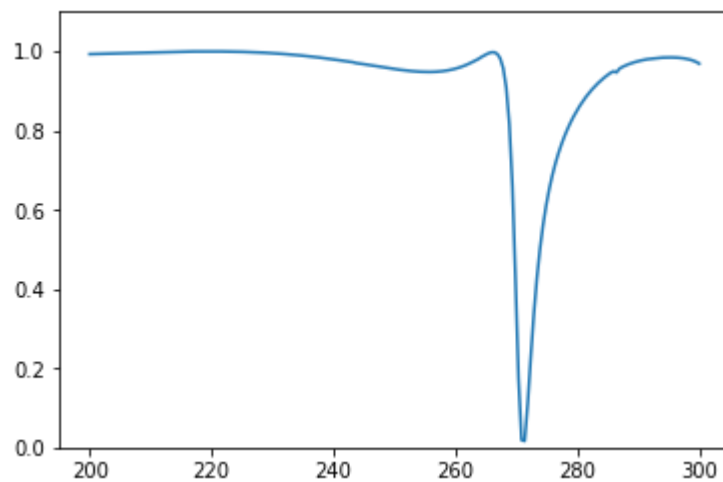


Predicted spectrum:

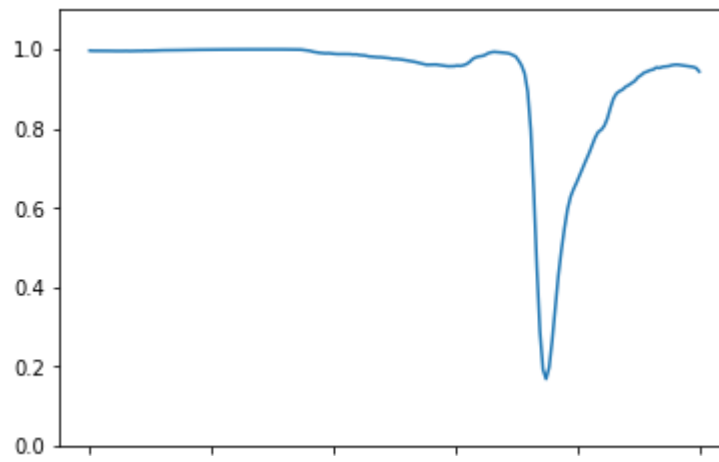


Test 81

True spectrum:

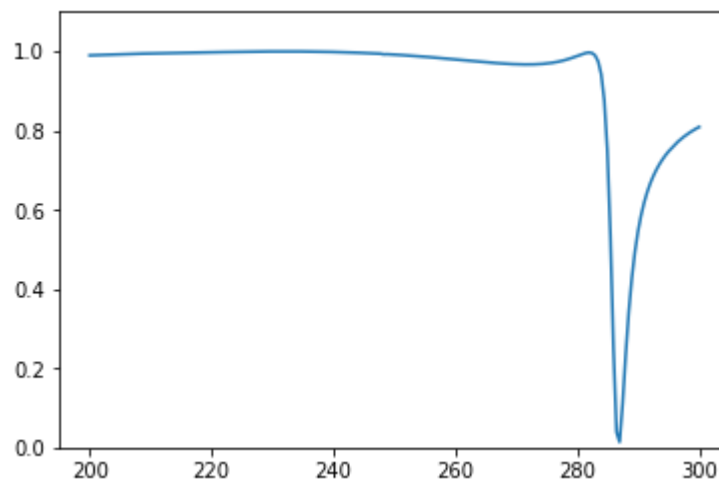


Predicted spectrum:

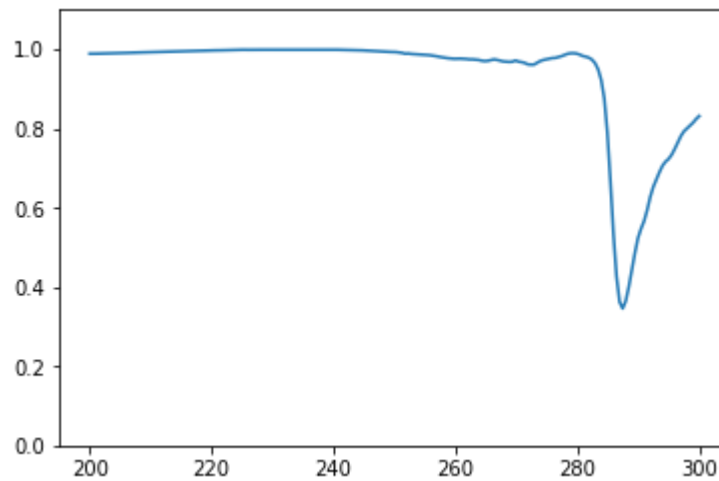


Test 82

True spectrum:

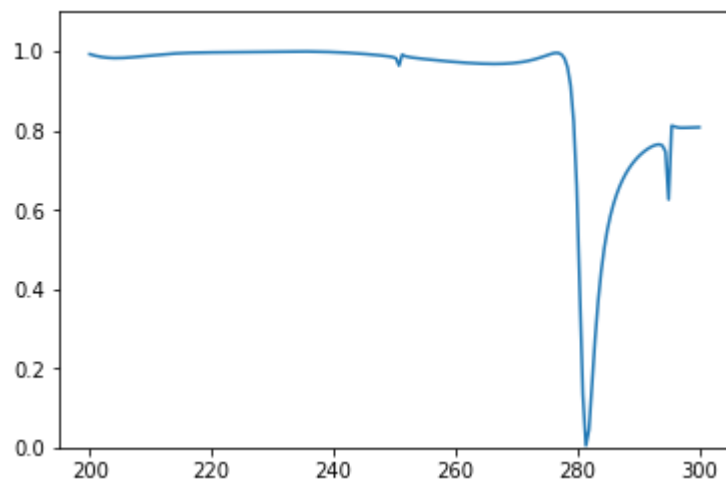


Predicted spectrum:

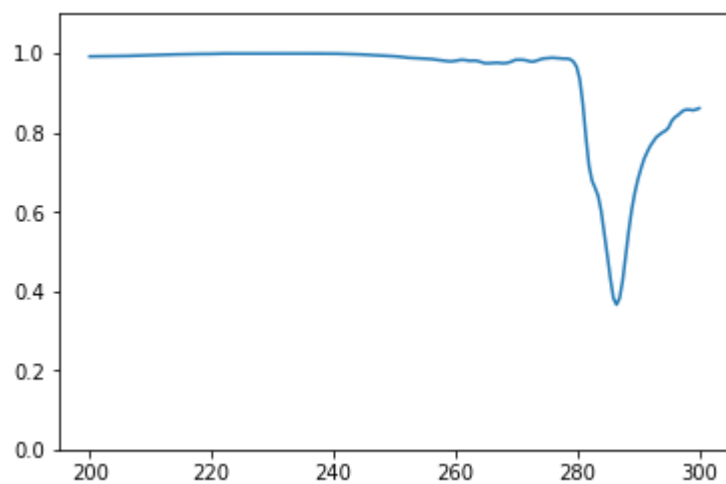


Test 83

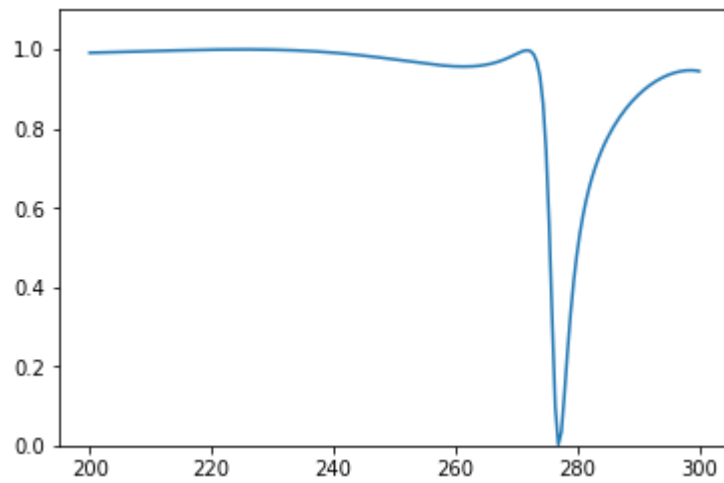
True spectrum:



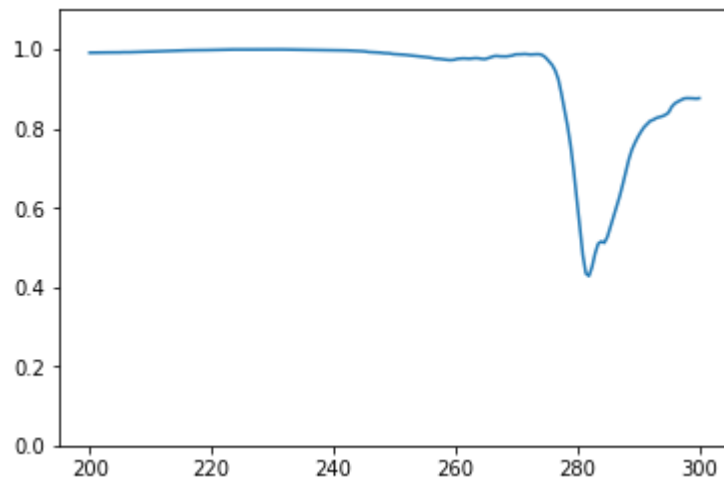
Predicted spectrum:



Test 84
True spectrum:

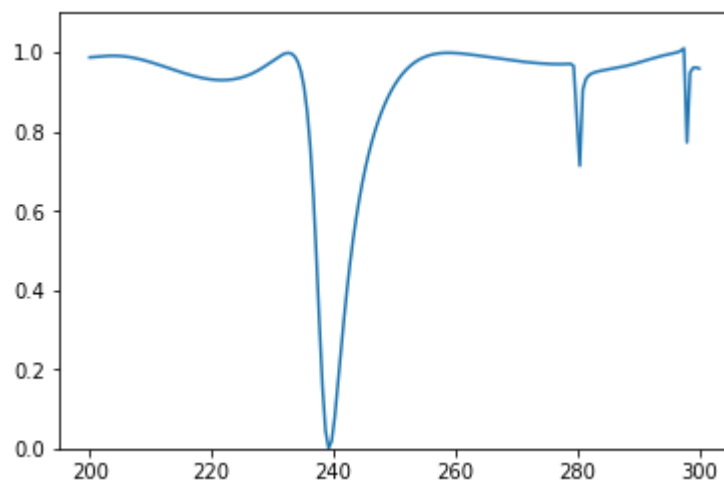


Predicted spectrum:



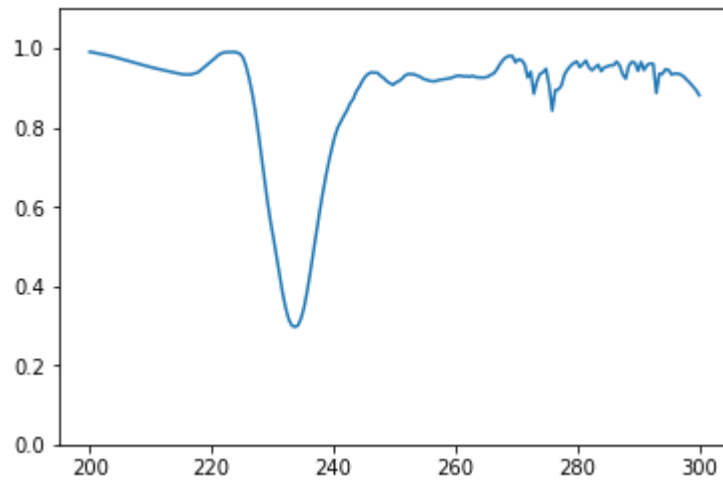
Test 85

True spectrum:



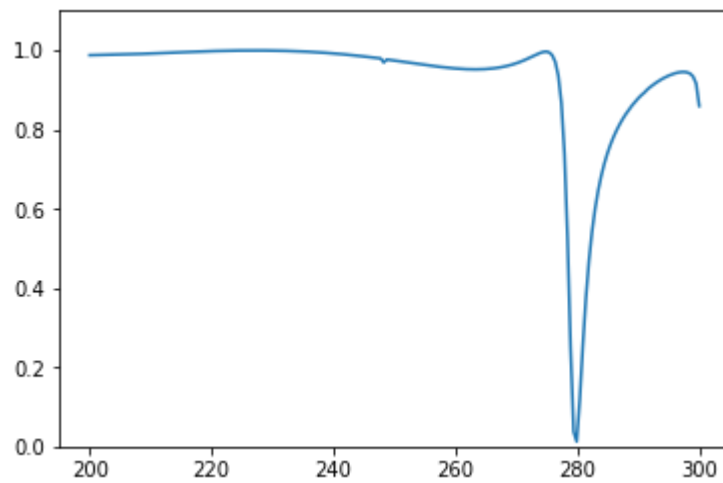


Predicted spectrum:

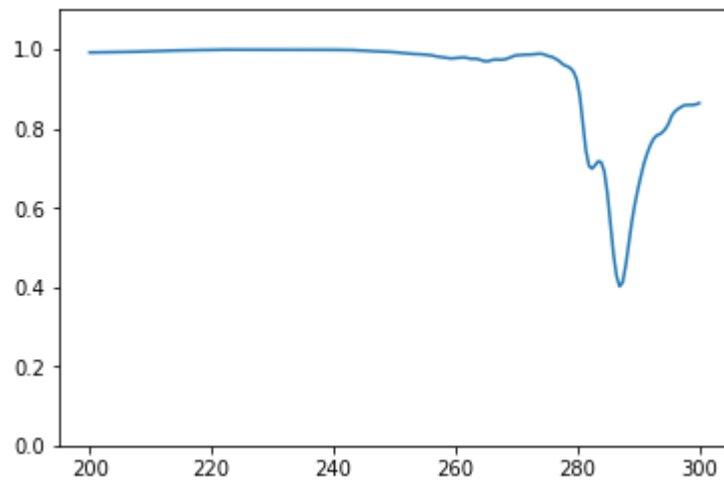


Test 86

True spectrum:

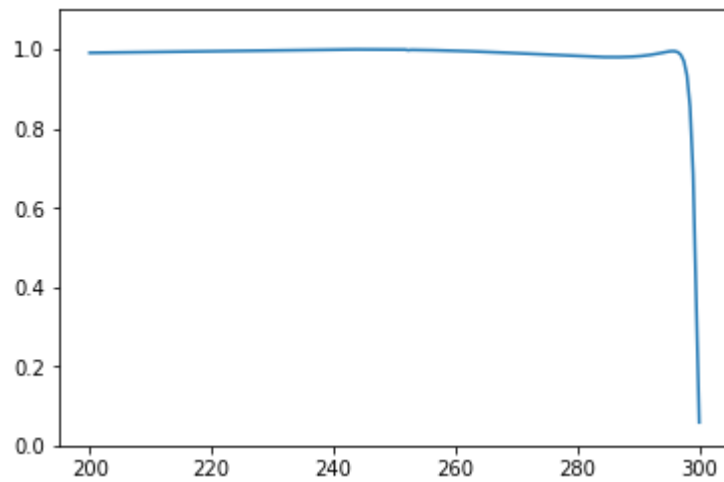


Predicted spectrum:

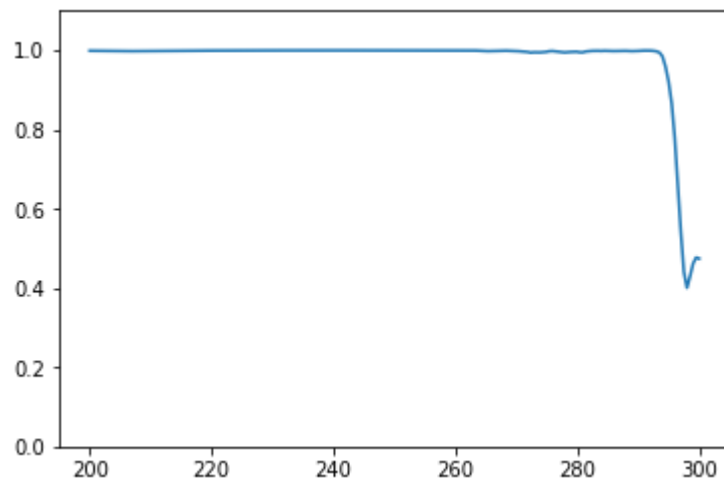


Test 87

True spectrum:

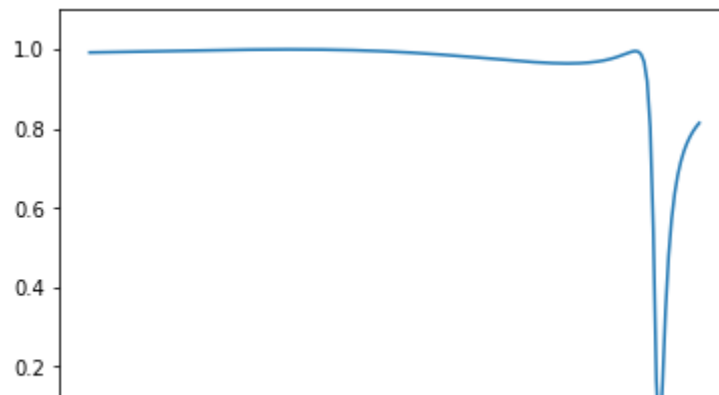


Predicted spectrum:

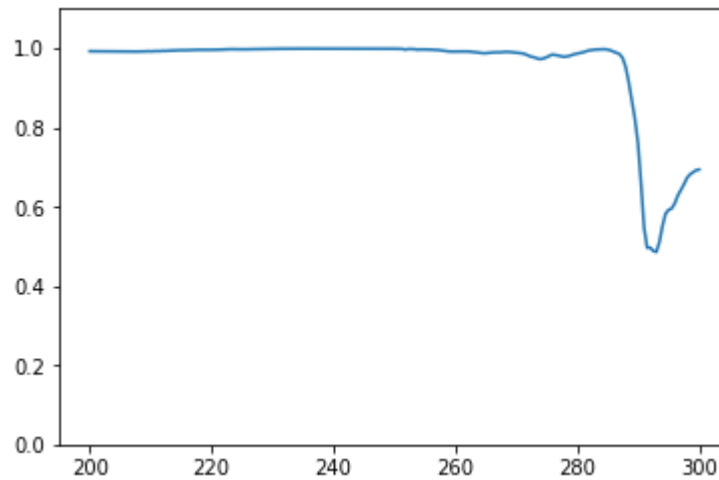


Test 88

True spectrum:

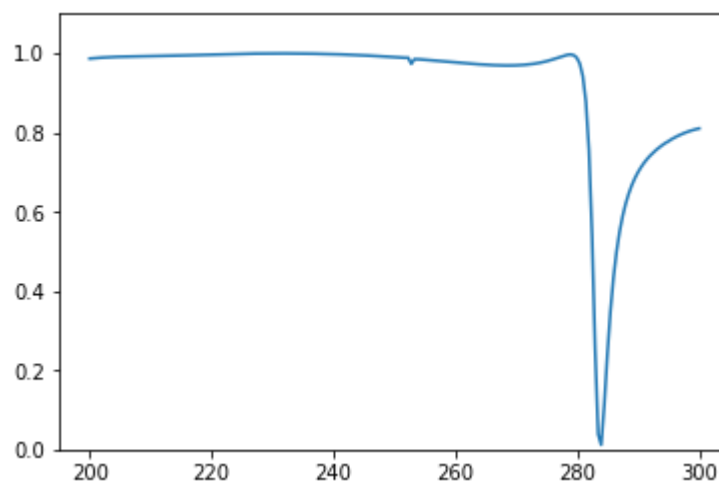


Predicted spectrum:

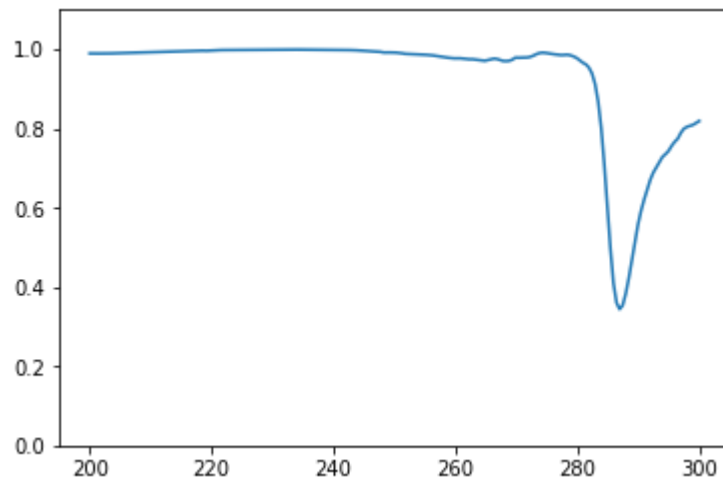


Test 89

True spectrum:

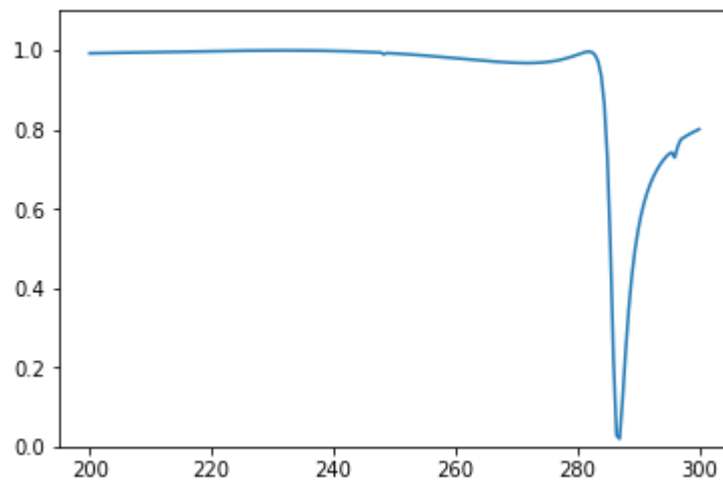


Predicted spectrum:

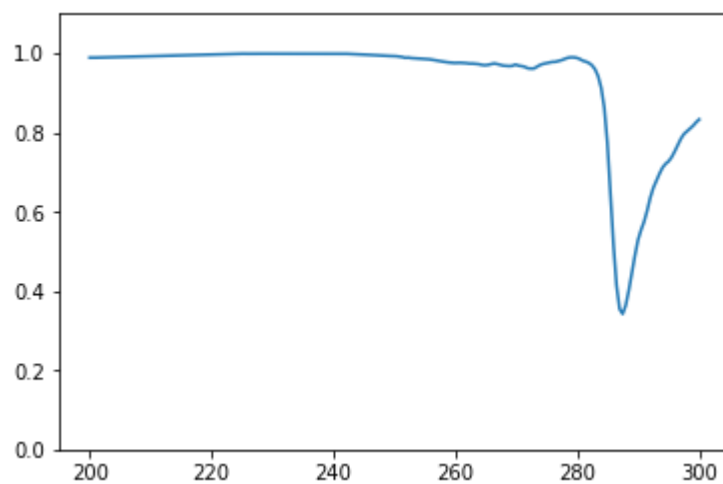


Test 90

True spectrum:

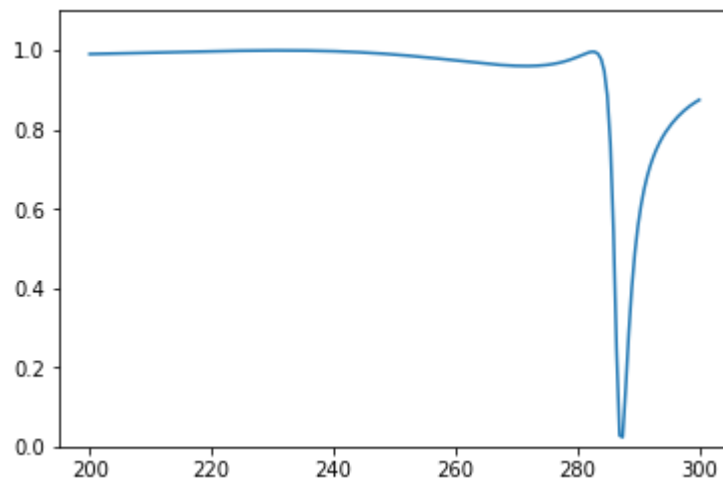


Predicted spectrum:

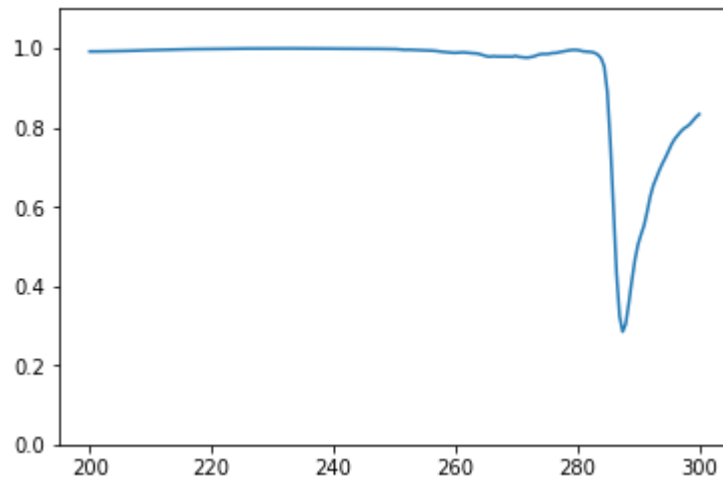


Test 91

True spectrum:

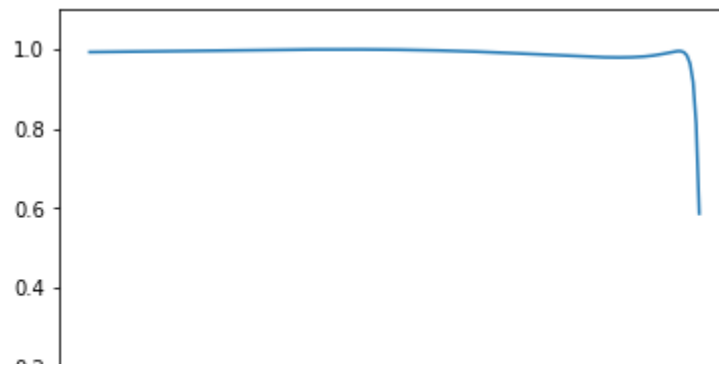


Predicted spectrum:



Test 92

True spectrum:



Predicted spectrum:

