

predict sin

In [1]:

```
import sys, os
sys.path.append(os.pardir) # 親ディレクトリのファイルをインポートするための設定
import numpy as np
from common import functions
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

np.random.seed(0)

# sin曲線
round_num = 10 # 10πだと、5周。
div_num = 500 # 分割数 5周を500分割だと、1周100分割。1分割3.6度。
ts = np.linspace(0, round_num * np.pi, div_num)
f = np.sin(ts)

# tanhの導関数
def d_tanh(x):
    return 1/(np.cosh(x)**2 + 1e-4)

# ひとつの時系列データの長さ
maxlen = 2

# sin波予測の入力データ
test_head = [[f[k]] for k in range(0, maxlen)] # 配列fのindexの0~k-1番目の値を抽出する。

# print(f)
print(test_head)
```

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In [2]:

```
data = []
target = []

for i in range(div_num - maxlen): # 500-2
    data.append(f[i: i + maxlen]) # 0~1, 1~2, 2~3,
    target.append(f[i + maxlen]) # 2, 3, 4,

X = np.array(data).reshape(len(data), maxlen, 1)
D = np.array(target).reshape(len(data), 1)
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In [3]:

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print(X)
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In [5]:

```
# データ設定  
N_train = int(len(data) * 0.8)  
N_validation = len(data) - N_train  
print(N_train, N_validation, len(data))
```

```
398 100 498
```

In [6]:

```
x_train, x_test, d_train, d_test = train_test_split(X, D, test_size=N_validation)
```

In [7]:

```
x_train
```

Out[7]:

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[[ 0.18156486],
 [ 0.11933469]],

[[-0.98611478],
 [-0.99460929]]])

```

In [8]:

```

input_layer_size = 1#入力層サイズ
hidden_layer_size = 5#隠れ層サイズ
output_layer_size = 1#出力層サイズ

weight_init_std = 0.01
learning_rate = 0.1#学習率

iters_num = 500#反復回数

```

In [9]:

```

# ウェイト初期化 (バイアスは簡単のため省略)
W_in = weight_init_std * np.random.randn(input_layer_size, hidden_layer_size)#係数×標準正規分布
W_out = weight_init_std * np.random.randn(hidden_layer_size, output_layer_size)
W = weight_init_std * np.random.randn(hidden_layer_size, hidden_layer_size)

```

In [10]:

```
print(W_in)
```

```
[[-0.00596314 -0.00052567 -0.0193628  0.00188779  0.00523891]]
```

In [11]:

```
# 勾配
```

```
W_in_grad = np.zeros_like(W_in)
```

```
W_out_grad = np.zeros_like(W_out)
```

```
W_grad = np.zeros_like(W)
```

In [12]:

```
print(W_in_grad)
```

```
[[0.  0.  0.  0.  0.]]
```

In [13]:

```
us = []
```

```
zs = []
```

```
u = np.zeros(hidden_layer_size)
```

```
z = np.zeros(hidden_layer_size)
```

```
y = np.zeros(output_layer_size)
```

In [14]:

```
print(u, z, y)
```

```
[0.  0.  0.  0.  0.] [0.  0.  0.  0.  0.] [0.]
```

In [15]:

```
delta_out = np.zeros(output_layer_size)
```

```
delta = np.zeros(hidden_layer_size)
```

In [16]:

```
print(delta_out, delta)
```

```
print(x_train.shape[0])
```

```
[0.] [0.  0.  0.  0.  0.]
```

```
398
```

In [17]:

```

losses = []

# トレーニング
for i in range(iters_num): #500
    for s in range(x_train.shape[0]): ##398行
        us.clear()
        zs.clear()
        z *= 0

        # sにおける正解データ
        d = d_train[s]

        xs = x_train[s]

        # 時系列ループ
        for t in range(maxlen): #t=0, t=1.

            # 入力値
            x = xs[t]
            u = np.dot(x, W_in) + np.dot(z, W)
            us.append(u)
            z = np.tanh(u)
            zs.append(z)

        y = np.dot(z, W_out) # z1 * W_out

        #誤差
        loss = functions.mean_squared_error(d, y)

        delta_out = functions.d_mean_squared_error(d, y)

        delta *= 0

        for t in range(maxlen)[::-1]: #range(1, -1, -1), t=1, t=0.

            #t=1の場合 delta = 0 + np.dot(delta_out, W_out.T) * d_tanh(us[t])
            #t=0の場合 delta = (np.dot(delta, W.T) + np.dot(delta_out, W_out.T)) * d_tanh(us[t])
            delta = (np.dot(delta, W.T) + np.dot(delta_out, W_out.T)) * d_tanh(us[t])

            # 勾配更新
            W_grad += np.dot(zs[t].reshape(-1, 1), delta.reshape(1, -1))
            W_in_grad += np.dot(xs[t], delta.reshape(1, -1))
            W_out_grad = np.dot(z.reshape(-1, 1), delta_out)

        # 勾配適用
        W -= learning_rate * W_grad
        W_in -= learning_rate * W_in_grad
        W_out -= learning_rate * W_out_grad.reshape(-1, 1)

        #勾配初期化
        W_in_grad *= 0
        W_out_grad *= 0
        W_grad *= 0

```

In [18]:

```
# テスト
for s in range(x_test.shape[0]):
    z *= 0

    # sにおける正解データ
    d = d_test[s]

    xs = x_test[s]

    # 時系列ループ
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)

    # 誤差
    loss = functions.mean_squared_error(d, y)
    print('loss:', loss, 'd:', d, 'y:', y)

original = np.full(maxlen, None)
pred_num = 200

xs = test_head

# sin波予測
for s in range(0, pred_num):
    z *= 0
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)
    original = np.append(original, y)
    xs = np.delete(xs, 0)
    xs = np.append(xs, y)

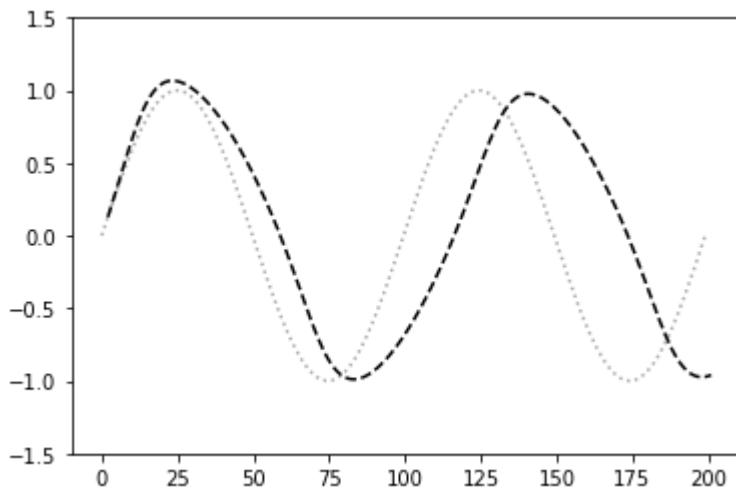
plt.figure()
plt.ylim([-1.5, 1.5])
plt.plot(np.sin(np.linspace(0, round_num* pred_num / div_num * np.pi, pred_num)), linestyle='dotted', color='#aaaaaa')
plt.plot(original, linestyle='dashed', color='black')
plt.show()
```

loss: 5.111112056304763e-07	d: [-0.47157024]	y: [-0.47258129]
loss: 1.4153640452728792e-06	d: [-0.39789889]	y: [-0.39958137]
loss: 2.559902991488718e-06	d: [-0.78740743]	y: [-0.78967013]
loss: 7.216263290783507e-09	d: [0.25526991]	y: [0.25514977]
loss: 1.692063852303632e-06	d: [0.6529121]	y: [0.6547517]
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loss: 1.4812813327813093e-06	d: [-0.3863158]	y: [-0.38803701]
loss: 1.5504762308512516e-06	d: [0.63363256]	y: [0.63539351]
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loss: 1.6215636803484033e-06	d: [0.64332332]	y: [0.64512419]
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loss: 9.92544090722427e-08	d: [-0.98611478]	y: [-0.98566924]
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loss: 1.8816335271670955e-06	d: [-0.94789551]	y: [-0.94983543]

```

loss: 1. 2190012629922352e-06    d: [-0.97512765]    y: [-0.97668906]
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loss: 1. 8748123442560268e-06    d: [0.22471249]    y: [0.22664888]
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loss: 6. 585761031379055e-07    d: [-0.99085292]   y: [-0.99200059]
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loss: 1. 6566551435475068e-06    d: [0.13806466]    y: [0.13988491]
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loss: 2. 09155034432836e-06    d: [0.70821885]    y: [0.71026411]
loss: 1. 7631822152185102e-06    d: [0.95374324]    y: [0.9556211]
loss: 2. 4114224765198005e-07    d: [0.97512765]    y: [0.97443319]
loss: 2. 067490890143368e-06    d: [0.93739898]    y: [0.93943244]
loss: 8. 485394806789291e-07    d: [0.98611478]    y: [0.9874175]
loss: 1. 72986325908338e-07    d: [0.98039956]    y: [0.97981136]
loss: 3. 9139136578767244e-07    d: [-0.55262221]   y: [-0.55350696]
loss: 5. 144245939051554e-07    d: [0.08175375]    y: [0.08073943]
loss: 1. 5434925122990647e-06    d: [-0.37467145]   y: [-0.37642843]
loss: 8. 64601116962286e-08    d: [0.98714074]    y: [0.9867249]
loss: 1. 7821870821634795e-06    d: [0.17537017]    y: [0.17725812]
loss: 1. 7227716039716527e-06    d: [-0.95561698]   y: [-0.9574732]
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loss: 1. 4861572178323642e-07    d: [-0.38050117]   y: [-0.38104635]
loss: 8. 149646046184465e-07    d: [-0.48814053]   y: [-0.48941721]
loss: 9. 154701847834834e-07    d: [0.54208448]    y: [0.5434376]
loss: 8. 19187220025201e-09    d: [-0.69021707]   y: [-0.69008907]
loss: 1. 1946819048400479e-07    d: [0.60896952]    y: [0.60945833]

```



[try]

- `iters_num`を100にしよう

In [19]:

```
import sys, os
sys.path.append(os.pardir) # 親ディレクトリのファイルをインポートするための設定
import numpy as np
from common import functions
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

np.random.seed(0)

# sin曲線
round_num = 10
div_num = 500
ts = np.linspace(0, round_num * np.pi, div_num)
f = np.sin(ts)

def d_tanh(x):
    return 1/(np.cosh(x)**2 + 1e-4)

# ひとつの時系列データの長さ
maxlen = 2

# sin波予測の入力データ
test_head = [[f[k]] for k in range(0, maxlen)]

data = []
target = []

for i in range(div_num - maxlen):
    data.append(f[i: i + maxlen])
    target.append(f[i + maxlen])

X = np.array(data).reshape(len(data), maxlen, 1)
D = np.array(target).reshape(len(data), 1)

# データ設定
N_train = int(len(data) * 0.8)
N_validation = len(data) - N_train

x_train, x_test, d_train, d_test = train_test_split(X, D, test_size=N_validation)

input_layer_size = 1
hidden_layer_size = 5
output_layer_size = 1

weight_init_std = 0.01
learning_rate = 0.1

iters_num = 100

# ウェイト初期化 (バイアスは簡単のため省略)
W_in = weight_init_std * np.random.randn(input_layer_size, hidden_layer_size)
W_out = weight_init_std * np.random.randn(hidden_layer_size, output_layer_size)
W = weight_init_std * np.random.randn(hidden_layer_size, hidden_layer_size)

# 勾配
W_in_grad = np.zeros_like(W_in)
W_out_grad = np.zeros_like(W_out)
W_grad = np.zeros_like(W)
```



```
us = []
zs = []

u = np.zeros(hidden_layer_size)
z = np.zeros(hidden_layer_size)
y = np.zeros(output_layer_size)

delta_out = np.zeros(output_layer_size)
delta = np.zeros(hidden_layer_size)

losses = []

# トレーニング
for i in range(iters_num):
    for s in range(x_train.shape[0]):
        us.clear()
        zs.clear()
        z *= 0

        # sにおける正解データ
        d = d_train[s]

        xs = x_train[s]

        # 時系列ループ
        for t in range(maxlen):

            # 入力値
            x = xs[t]
            u = np.dot(x, W_in) + np.dot(z, W)
            us.append(u)
            z = np.tanh(u)
            zs.append(z)

        y = np.dot(z, W_out)

        #誤差
        loss = functions.mean_squared_error(d, y)

        delta_out = functions.d_mean_squared_error(d, y)

        delta *= 0
        for t in range(maxlen)[::-1]:

            delta = (np.dot(delta, W.T) + np.dot(delta_out, W_out.T)) * d_tanh(us[t])

            # 勾配更新
            W_grad += np.dot(zs[t].reshape(-1, 1), delta.reshape(1, -1))
            W_in_grad += np.dot(xs[t], delta.reshape(1, -1))
            W_out_grad = np.dot(z.reshape(-1, 1), delta_out)

            # 勾配適用
            W -= learning_rate * W_grad
            W_in -= learning_rate * W_in_grad
            W_out -= learning_rate * W_out_grad.reshape(-1, 1)

            W_in_grad *= 0
            W_out_grad *= 0
            W_grad *= 0

# テスト
```

```
for s in range(x_test.shape[0]):
    z *= 0

    # sにおける正解データ
    d = d_test[s]

    xs = x_test[s]

    # 時系列ループ
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)

    #誤差
    loss = functions.mean_squared_error(d, y)
    print('loss:', loss, 'd:', d, 'y:', y)

original = np.full(maxlen, None)
pred_num = 200

xs = test_head

# sin波予測
for s in range(0, pred_num):
    z *= 0
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)
    original = np.append(original, y)
    xs = np.delete(xs, 0)
    xs = np.append(xs, y)

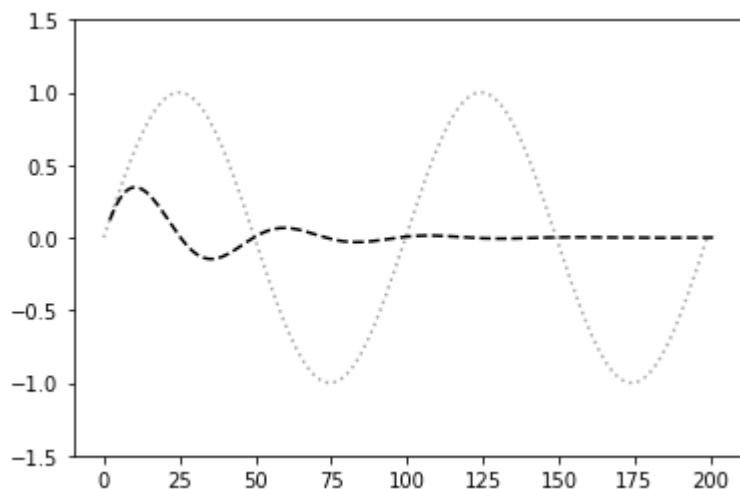
plt.figure()
plt.ylim([-1.5, 1.5])
plt.plot(np.sin(np.linspace(0, round_num* pred_num / div_num * np.pi, pred_num)), linestyle='dotted', color='aaaaaa')
plt.plot(original, linestyle='dashed', color='black')
plt.show()
```

loss: 1.0018211688363152e-06	d: [-0.47157024]	y: [-0.47298574]
loss: 3.831628581739869e-05	d: [-0.39789889]	y: [-0.38914489]
loss: 1.1955502192448862e-05	d: [-0.78740743]	y: [-0.79229732]
loss: 5.688169957032234e-07	d: [0.25526991]	y: [0.25633651]
loss: 5.042766131766361e-06	d: [0.6529121]	y: [0.65608787]
loss: 2.323535484703895e-05	d: [0.8773359]	y: [0.87051895]
loss: 1.2851597420607724e-05	d: [0.92114593]	y: [0.9160761]
loss: 2.9577805540623868e-06	d: [-0.58880346]	y: [-0.59123566]
loss: 5.3033087017894245e-05	d: [-0.6529121]	y: [-0.64261325]
loss: 3.068291439526736e-05	d: [-0.32751865]	y: [-0.31968501]
loss: 5.638683232237903e-07	d: [0.39789889]	y: [0.39896084]
loss: 1.4135523848939483e-05	d: [-0.15674537]	y: [-0.15142832]
loss: 1.3233060812018978e-05	d: [0.97076771]	y: [0.97591224]
loss: 1.3588146407522809e-06	d: [-0.50453668]	y: [-0.50618521]
loss: 1.352410323795142e-07	d: [-0.99460929]	y: [-0.99512937]
loss: 1.4962796333390856e-05	d: [0.95745284]	y: [0.96292327]
loss: 1.1148085202439717e-05	d: [-0.92833248]	y: [-0.92361059]
loss: 3.3266626628950615e-05	d: [-0.35120641]	y: [-0.34304962]
loss: 3.707863108292281e-05	d: [-0.3863158]	y: [-0.37770434]
loss: 4.326570868918968e-06	d: [0.63363256]	y: [0.63657418]
loss: 4.3539234693184196e-05	d: [0.44921588]	y: [0.4398843]
loss: 2.533672093849227e-05	d: [-0.86811636]	y: [-0.86099783]
loss: 8.639496897928521e-07	d: [0.45483173]	y: [0.45614623]
loss: 5.311920595825067e-05	d: [0.59893397]	y: [0.58862677]
loss: 1.0055949688712286e-06	d: [0.18156486]	y: [0.18298303]
loss: 4.676740710063253e-06	d: [0.64332332]	y: [0.64638167]
loss: 1.59930658073789e-08	d: [0.99247351]	y: [0.99265236]
loss: 1.8474236177344764e-05	d: [-0.89762559]	y: [-0.89154706]
loss: 1.8802261802955873e-06	d: [-0.11308158]	y: [-0.11502077]
loss: 1.2182950786145393e-05	d: [0.79127273]	y: [0.79620891]
loss: 4.857808115858404e-06	d: [-0.64813056]	y: [-0.65124754]
loss: 5.0259896375981215e-06	d: [0.95561698]	y: [0.95244649]
loss: 5.84751441954738e-06	d: [0.99690497]	y: [1.00032477]
loss: 5.842918948781338e-07	d: [-0.4036669]	y: [-0.40474791]
loss: 2.0465335767793966e-06	d: [-0.99987614]	y: [-1.00189927]
loss: 5.237713037590953e-05	d: [-0.57343317]	y: [-0.56319822]
loss: 4.550293795779566e-05	d: [0.75537465]	y: [0.74583495]
loss: 4.716398434628641e-06	d: [0.00629574]	y: [0.00936702]
loss: 2.065453201950312e-05	d: [-0.23084276]	y: [-0.22441554]
loss: 2.3468307818580476e-06	d: [0.56307233]	y: [0.56523881]
loss: 1.8459664559315096e-06	d: [0.99975723]	y: [1.00167867]
loss: 1.7754700800083484e-06	d: [0.11933469]	y: [0.12121909]
loss: 2.1247195911347073e-05	d: [0.23696388]	y: [0.2304451]
loss: 2.0004302281892026e-07	d: [0.99524241]	y: [0.99587493]
loss: 1.1411501423016343e-05	d: [-0.98039956]	y: [-0.9851769]
loss: 1.6706877736517124e-05	d: [0.18775236]	y: [0.18197189]
loss: 3.446312694027539e-05	d: [0.82463104]	y: [0.81632886]
loss: 6.2409520302279e-06	d: [-0.94988243]	y: [-0.94634946]
loss: 1.1705129757380487e-05	d: [0.92597363]	y: [0.92113522]
loss: 2.3677213141816805e-05	d: [0.26135201]	y: [0.25447055]
loss: 4.49811295488685e-05	d: [-0.75948523]	y: [-0.75000038]
loss: 2.028533646768841e-07	d: [-0.98611478]	y: [-0.98547783]
loss: 7.499350947699529e-07	d: [-0.43793098]	y: [-0.43915567]
loss: 6.702616914592727e-07	d: [0.23084276]	y: [0.23200057]
loss: 1.4664769817573572e-05	d: [0.83516734]	y: [0.84058301]
loss: 4.333798205022689e-05	d: [0.77163571]	y: [0.76232571]
loss: 1.2737717225792436e-05	d: [0.97371292]	y: [0.97876023]
loss: 3.4112375431131424e-06	d: [-0.99975723]	y: [-1.00236922]
loss: 6.429288289821193e-06	d: [0.99583607]	y: [0.99942196]
loss: 7.85436168297347e-07	d: [0.44358222]	y: [0.44483557]
loss: 1.5805387662329316e-05	d: [-0.94789551]	y: [-0.95351786]

```

loss: 1. 2481961197274553e-05    d: [-0.97512765]    y: [-0.98012404]
loss: 6. 866202285129303e-06    d: [-0.694759]    y: [-0.69846473]
loss: 2. 1846135041325457e-05    d: [-0.2430756]    y: [-0.23646559]
loss: 2. 00684638533191e-05      d: [0.22471249]    y: [0.21837712]
loss: 2. 1074553733212728e-06    d: [-0.10056216]   y: [-0.10261519]
loss: 5. 174173943527327e-05    d: [0.68564779]    y: [0.6754751]
loss: 2. 74601418124138e-05      d: [0.29761864]    y: [0.29020782]
loss: 2. 7697525031755335e-07    d: [-0.99583607]   y: [-0.99658035]
loss: 8. 510577037380914e-06     d: [-0.99085292]   y: [-0.99497859]
loss: 4. 6342324398754784e-07    d: [0.35120641]    y: [0.35216914]
loss: 2. 2541089852561257e-05    d: [-0.88033969]   y: [-0.87362536]
loss: 5. 198366295972673e-05     d: [-0.68105132]   y: [-0.67085488]
loss: 1. 5650869216024103e-05    d: [0.85534252]    y: [0.86093731]
loss: 9. 106005186775855e-06     d: [-0.98907524]   y: [-0.99334279]
loss: 4. 9537135434721655e-05    d: [-0.71705202]   y: [-0.70709842]
loss: 2. 674810502869443e-05     d: [-0.86179776]   y: [-0.85448365]
loss: 1. 2704010990953187e-05    d: [0.13806466]    y: [0.13302402]
loss: 1. 108184027478091e-06     d: [-0.48263615]   y: [-0.4841249]
loss: 1. 3322624649906846e-06    d: [-0.15052435]   y: [-0.15215669]
loss: 1. 4631672720098082e-05    d: [0.16296018]    y: [0.15755062]
loss: 7. 5277718530109865e-06    d: [0.70821885]    y: [0.712099]
loss: 1. 532365497732746e-05     d: [0.95374324]    y: [0.95927924]
loss: 1. 4698452123511701e-06    d: [0.97512765]    y: [0.9734131]
loss: 1. 6440458209345824e-05    d: [0.93739898]    y: [0.94313317]
loss: 9. 989092604746525e-06     d: [0.98611478]    y: [0.99058448]
loss: 7. 597360561243469e-07     d: [0.98039956]    y: [0.97916689]
loss: 5. 144766400804239e-05     d: [-0.55262221]   y: [-0.54247848]
loss: 2. 49562656838687e-06      d: [0.08175375]    y: [0.08398787]
loss: 3. 582176538836955e-05     d: [-0.37467145]   y: [-0.3662072]
loss: 1. 3465686259774947e-07     d: [0.98714074]    y: [0.98662178]
loss: 1. 5651516544468314e-05    d: [0.17537017]    y: [0.16977526]
loss: 1. 5147090412232574e-05    d: [-0.95561698]   y: [-0.96112099]
loss: 1. 3648771002149447e-05     d: [0.15052435]    y: [0.14529965]
loss: 1. 695825274903518e-05     d: [-0.92114593]   y: [-0.92696972]
loss: 5. 135842537369791e-07     d: [-0.38050117]   y: [-0.38151466]
loss: 4. 704410899615774e-05     d: [-0.48814053]   y: [-0.47844062]
loss: 1. 933106793590669e-06     d: [0.54208448]    y: [0.54405075]
loss: 5. 148111233523464e-05     d: [-0.69021707]   y: [-0.68007004]
loss: 5. 3281078139314694e-05    d: [0.60896952]    y: [0.59864663]

```



[try]

- `iters_num`を3000(※時間がかかる)にしよう

In [20]:

```
import sys, os
sys.path.append(os.pardir) # 親ディレクトリのファイルをインポートするための設定
import numpy as np
from common import functions
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

np.random.seed(0)

# sin曲線
round_num = 10
div_num = 500
ts = np.linspace(0, round_num * np.pi, div_num)
f = np.sin(ts)

def d_tanh(x):
    return 1/(np.cosh(x)**2 + 1e-4)

# ひとつの時系列データの長さ
maxlen = 2

# sin波予測の入力データ
test_head = [[f[k]] for k in range(0, maxlen)]

data = []
target = []

for i in range(div_num - maxlen):
    data.append(f[i: i + maxlen])
    target.append(f[i + maxlen])

X = np.array(data).reshape(len(data), maxlen, 1)
D = np.array(target).reshape(len(data), 1)

# データ設定
N_train = int(len(data) * 0.8)
N_validation = len(data) - N_train

x_train, x_test, d_train, d_test = train_test_split(X, D, test_size=N_validation)

input_layer_size = 1
hidden_layer_size = 5
output_layer_size = 1

weight_init_std = 0.01
learning_rate = 0.1

iters_num = 3000

# ウェイト初期化 (バイアスは簡単のため省略)
W_in = weight_init_std * np.random.randn(input_layer_size, hidden_layer_size)
W_out = weight_init_std * np.random.randn(hidden_layer_size, output_layer_size)
W = weight_init_std * np.random.randn(hidden_layer_size, hidden_layer_size)

# 勾配
W_in_grad = np.zeros_like(W_in)
W_out_grad = np.zeros_like(W_out)
W_grad = np.zeros_like(W)
```

```

us = []
zs = []

u = np.zeros(hidden_layer_size)
z = np.zeros(hidden_layer_size)
y = np.zeros(output_layer_size)

delta_out = np.zeros(output_layer_size)
delta = np.zeros(hidden_layer_size)

losses = []

# トレーニング
for i in range(iters_num):
    for s in range(x_train.shape[0]):
        us.clear()
        zs.clear()
        z *= 0

        # sにおける正解データ
        d = d_train[s]

        xs = x_train[s]

        # 時系列ループ
        for t in range(maxlen):

            # 入力値
            x = xs[t]
            u = np.dot(x, W_in) + np.dot(z, W)
            us.append(u)
            z = np.tanh(u)
            zs.append(z)

        y = np.dot(z, W_out)

        #誤差
        loss = functions.mean_squared_error(d, y)

        delta_out = functions.d_mean_squared_error(d, y)

        delta *= 0
        for t in range(maxlen)[::-1]:

            delta = (np.dot(delta, W.T) + np.dot(delta_out, W_out.T)) * d_tanh(us[t])

            # 勾配更新
            W_grad += np.dot(zs[t].reshape(-1, 1), delta.reshape(1, -1))
            W_in_grad += np.dot(xs[t], delta.reshape(1, -1))
            W_out_grad = np.dot(z.reshape(-1, 1), delta_out)

            # 勾配適用
            W -= learning_rate * W_grad
            W_in -= learning_rate * W_in_grad
            W_out -= learning_rate * W_out_grad.reshape(-1, 1)

            W_in_grad *= 0
            W_out_grad *= 0
            W_grad *= 0

# テスト

```

```

for s in range(x_test.shape[0]):
    z *= 0

    # sにおける正解データ
    d = d_test[s]

    xs = x_test[s]

    # 時系列ループ
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)

    #誤差
    loss = functions.mean_squared_error(d, y)
    print('loss:', loss, 'd:', d, 'y:', y)

original = np.full(maxlen, None)
pred_num = 200

xs = test_head

# sin波予測
for s in range(0, pred_num):
    z *= 0
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)
    original = np.append(original, y)
    xs = np.delete(xs, 0)
    xs = np.append(xs, y)

plt.figure()
plt.ylim([-1.5, 1.5])
plt.plot(np.sin(np.linspace(0, round_num* pred_num / div_num * np.pi, pred_num)), linestyle='dotted', color='aaaaaa')
plt.plot(original, linestyle='dashed', color='black')
plt.show()

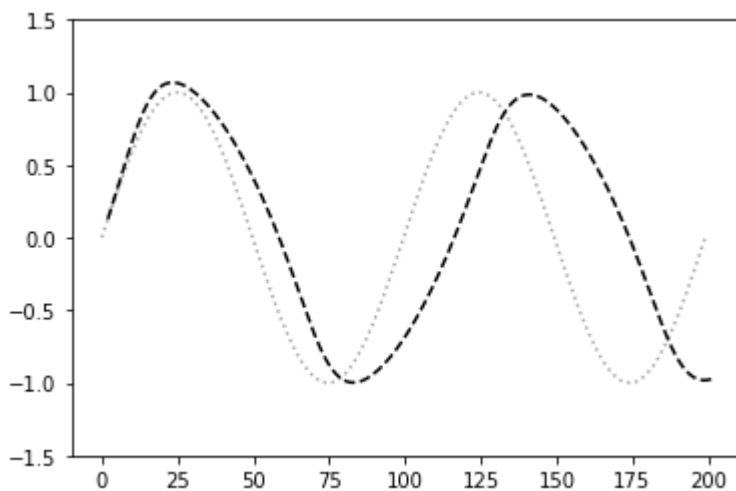
```

loss: 3.912343122231309e-07	d: [-0.47157024]	y: [-0.47245481]
loss: 1.3052648322202188e-06	d: [-0.39789889]	y: [-0.39951461]
loss: 2.4568661324794487e-06	d: [-0.78740743]	y: [-0.78962413]
loss: 1.836933553802922e-08	d: [0.25526991]	y: [0.25507824]
loss: 1.5234519740797024e-06	d: [0.6529121]	y: [0.65465764]
loss: 5.187974276348698e-07	d: [0.8773359]	y: [0.87631727]
loss: 5.147787696349783e-07	d: [0.92114593]	y: [0.92013126]
loss: 1.0581260161824844e-06	d: [-0.58880346]	y: [-0.5902582]
loss: 3.273298436623086e-08	d: [-0.6529121]	y: [-0.65316796]
loss: 1.577987966024491e-06	d: [-0.32751865]	y: [-0.32929515]
loss: 1.336466170586074e-07	d: [0.39789889]	y: [0.3984159]
loss: 1.536853774722165e-06	d: [-0.15674537]	y: [-0.15849857]
loss: 1.3150157772974342e-06	d: [0.97076771]	y: [0.97238945]
loss: 5.488011547346638e-07	d: [-0.50453668]	y: [-0.50558435]
loss: 3.4729876467868902e-09	d: [-0.99460929]	y: [-0.99452595]
loss: 1.6448993535149984e-06	d: [0.95745284]	y: [0.95926662]
loss: 4.937438724118824e-07	d: [-0.92833248]	y: [-0.92733875]
loss: 1.5034914183015375e-06	d: [-0.35120641]	y: [-0.35294047]
loss: 1.3602120481600566e-06	d: [-0.3863158]	y: [-0.38796517]
loss: 1.3788941646447499e-06	d: [0.63363256]	y: [0.63529321]
loss: 1.026676639415935e-06	d: [0.44921588]	y: [0.45064883]
loss: 4.989099370011925e-07	d: [-0.86811636]	y: [-0.86711745]
loss: 3.2105982617360104e-07	d: [0.45483173]	y: [0.45563306]
loss: 1.9274300037523416e-07	d: [0.59893397]	y: [0.59955485]
loss: 1.4944014901163935e-07	d: [0.18156486]	y: [0.18101816]
loss: 1.45121458712902e-06	d: [0.64332332]	y: [0.64502698]
loss: 1.4921537554827992e-08	d: [0.99247351]	y: [0.99230076]
loss: 5.402630094375798e-07	d: [-0.89762559]	y: [-0.8965861]
loss: 3.707778123070018e-07	d: [-0.11308158]	y: [-0.11222045]
loss: 2.475796598999519e-06	d: [0.79127273]	y: [0.79349794]
loss: 1.4873575430756097e-06	d: [-0.64813056]	y: [-0.64985529]
loss: 3.474647728818484e-07	d: [0.95561698]	y: [0.95478335]
loss: 3.6223161724552973e-07	d: [0.99690497]	y: [0.99775613]
loss: 1.489766397919691e-07	d: [-0.4036669]	y: [-0.40421275]
loss: 6.468374433663414e-08	d: [-0.99987614]	y: [-1.00023581]
loss: 3.0702144887264835e-07	d: [-0.57343317]	y: [-0.57421678]
loss: 1.000399708517381e-07	d: [0.75537465]	y: [0.75492734]
loss: 8.444804874432303e-07	d: [0.00629574]	y: [0.00499613]
loss: 1.675728118498778e-06	d: [-0.23084276]	y: [-0.23267345]
loss: 8.878245718019855e-07	d: [0.56307233]	y: [0.56440486]
loss: 5.348947467931103e-08	d: [0.99975723]	y: [1.00008431]
loss: 3.4716720303622646e-07	d: [0.11933469]	y: [0.11850143]
loss: 1.6794114068719324e-06	d: [0.23696388]	y: [0.23879658]
loss: 1.4042006221287043e-09	d: [0.99524241]	y: [0.99518941]
loss: 1.0268531315089448e-06	d: [-0.98039956]	y: [-0.98183263]
loss: 1.6149340072836495e-06	d: [0.18775236]	y: [0.18954954]
loss: 3.5352983082116036e-07	d: [0.82463104]	y: [0.82379017]
loss: 3.873534373161674e-07	d: [-0.94988243]	y: [-0.94900226]
loss: 5.013989623342182e-07	d: [0.92597363]	y: [0.92497223]
loss: 1.6811027696115039e-06	d: [0.26135201]	y: [0.26318564]
loss: 1.121968297083886e-07	d: [-0.75948523]	y: [-0.75901153]
loss: 6.853245118558047e-08	d: [-0.98611478]	y: [-0.98574456]
loss: 2.57089195761961e-07	d: [-0.43793098]	y: [-0.43864804]
loss: 4.827092212115843e-08	d: [0.23084276]	y: [0.23053204]
loss: 2.6191637340832623e-06	d: [0.83516734]	y: [0.83745608]
loss: 1.5126301136728962e-07	d: [0.77163571]	y: [0.77108568]
loss: 1.2319066819462866e-06	d: [0.97371292]	y: [0.97528257]
loss: 1.5486378230634107e-07	d: [-0.99975723]	y: [-1.00031377]
loss: 4.192727163127846e-07	d: [0.99583607]	y: [0.99675179]
loss: 2.7770361244151876e-07	d: [0.44358222]	y: [0.44432748]
loss: 1.841913853315806e-06	d: [-0.94789551]	y: [-0.94981484]


```

loss: 1. 1904984010087743e-06    d: [-0.97512765]    y: [-0.9766707]
loss: 1. 8413639121995479e-06    d: [-0.694759]     y: [-0.69667805]
loss: 1. 6817990829983107e-06    d: [-0.2430756]    y: [-0.24490961]
loss: 1. 670761128353524e-06    d: [0.22471249]    y: [0.22654047]
loss: 4. 198304467446638e-07    d: [-0.10056216]   y: [-0.09964583]
loss: 3. 7564423210744303e-10    d: [0.68564779]    y: [0.6856752]
loss: 1. 6441452134331962e-06    d: [0.29761864]    y: [0.29943201]
loss: 2. 5076515871196765e-10    d: [-0.99583607]   y: [-0.99581368]
loss: 6. 461401578331407e-07    d: [-0.99085292]   y: [-0.99198971]
loss: 4. 021287785915039e-08    d: [0.35120641]    y: [0.35149]
loss: 5. 240456662428166e-07    d: [-0.88033969]   y: [-0.87931592]
loss: 1. 7735508691111668e-09    d: [-0.68105132]   y: [-0.68111088]
loss: 2. 6272940036240856e-06    d: [0.85534252]    y: [0.85763481]
loss: 7. 177811412733664e-07    d: [-0.98907524]   y: [-0.99027339]
loss: 1. 8205365057779595e-08    d: [-0.71705202]   y: [-0.71686121]
loss: 4. 823656298617088e-07    d: [-0.86179776]   y: [-0.86081555]
loss: 1. 4778770546303914e-06    d: [0.13806466]    y: [0.13978389]
loss: 4. 412968820138988e-07    d: [-0.48263615]   y: [-0.48357562]
loss: 2. 3912968440796433e-07    d: [-0.15052435]   y: [-0.14983279]
loss: 1. 5545796530955029e-06    d: [0.16296018]    y: [0.16472346]
loss: 1. 9422431386775e-06      d: [0.70821885]    y: [0.71018976]
loss: 1. 7249794746621768e-06    d: [0.95374324]    y: [0.95560065]
loss: 1. 7742172036673962e-07    d: [0.97512765]    y: [0.97453197]
loss: 2. 0256007449752334e-06    d: [0.93739898]    y: [0.93941174]
loss: 8. 299315679013527e-07    d: [0.98611478]    y: [0.98740314]
loss: 1. 2494659733137965e-07    d: [0.98039956]    y: [0.97989966]
loss: 4. 142129095805074e-07    d: [-0.55262221]   y: [-0.55353239]
loss: 4. 976219112443149e-07    d: [0.08175375]    y: [0.08075614]
loss: 1. 4117892688671288e-06    d: [-0.37467145]   y: [-0.3763518]
loss: 5. 8851642607912266e-08    d: [0.98714074]    y: [0.98679766]
loss: 1. 5869287610358092e-06    d: [0.17537017]    y: [0.1771517]
loss: 1. 6851238184123744e-06    d: [-0.95561698]   y: [-0.9574528]
loss: 1. 5181379496026437e-06    d: [0.15052435]    y: [0.15226684]
loss: 2. 251389961868721e-06    d: [-0.92114593]   y: [-0.92326791]
loss: 9. 244665177336762e-08    d: [-0.38050117]   y: [-0.38093116]
loss: 7. 925032084200643e-07    d: [-0.48814053]   y: [-0.4893995]
loss: 7. 580591186093183e-07    d: [0.54208448]    y: [0.54331578]
loss: 1. 0255350134194105e-11    d: [-0.69021707]   y: [-0.69021254]
loss: 1. 5392344980496395e-07    d: [0.60896952]    y: [0.60952436]

```



[try]

- maxlenを5にしよう

In []:

```
import sys, os
sys.path.append(os.pardir) # 親ディレクトリのファイルをインポートするための設定
import numpy as np
from common import functions
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

np.random.seed(0)

# sin曲線
round_num = 10
div_num = 500
ts = np.linspace(0, round_num * np.pi, div_num)
f = np.sin(ts)

def d_tanh(x):
    return 1/(np.cosh(x)**2 + 1e-4)

# ひとつの時系列データの長さ
maxlen = 5

# sin波予測の入力データ
test_head = [[f[k]] for k in range(0, maxlen)]

data = []
target = []

for i in range(div_num - maxlen):
    data.append(f[i: i + maxlen])
    target.append(f[i + maxlen])

X = np.array(data).reshape(len(data), maxlen, 1)
D = np.array(target).reshape(len(data), 1)

# データ設定
N_train = int(len(data) * 0.8)
N_validation = len(data) - N_train

x_train, x_test, d_train, d_test = train_test_split(X, D, test_size=N_validation)

input_layer_size = 1
hidden_layer_size = 5
output_layer_size = 1

weight_init_std = 0.01
learning_rate = 0.1

iters_num = 500

# ウェイト初期化 (バイアスは簡単のため省略)
W_in = weight_init_std * np.random.randn(input_layer_size, hidden_layer_size)
W_out = weight_init_std * np.random.randn(hidden_layer_size, output_layer_size)
W = weight_init_std * np.random.randn(hidden_layer_size, hidden_layer_size)

# 勾配
W_in_grad = np.zeros_like(W_in)
W_out_grad = np.zeros_like(W_out)
W_grad = np.zeros_like(W)
```

```
us = []
zs = []

u = np.zeros(hidden_layer_size)
z = np.zeros(hidden_layer_size)
y = np.zeros(output_layer_size)

delta_out = np.zeros(output_layer_size)
delta = np.zeros(hidden_layer_size)

losses = []

# トレーニング
for i in range(iters_num):
    for s in range(x_train.shape[0]):
        us.clear()
        zs.clear()
        z *= 0

        # sにおける正解データ
        d = d_train[s]

        xs = x_train[s]

        # 時系列ループ
        for t in range(maxlen):

            # 入力値
            x = xs[t]
            u = np.dot(x, W_in) + np.dot(z, W)
            us.append(u)
            z = np.tanh(u)
            zs.append(z)

        y = np.dot(z, W_out)

        #誤差
        loss = functions.mean_squared_error(d, y)

        delta_out = functions.d_mean_squared_error(d, y)

        delta *= 0
        for t in range(maxlen)[::-1]:

            delta = (np.dot(delta, W.T) + np.dot(delta_out, W_out.T)) * d_tanh(us[t])

            # 勾配更新
            W_grad += np.dot(zs[t].reshape(-1, 1), delta.reshape(1, -1))
            W_in_grad += np.dot(xs[t], delta.reshape(1, -1))
            W_out_grad = np.dot(z.reshape(-1, 1), delta_out)

            # 勾配適用
            W -= learning_rate * W_grad
            W_in -= learning_rate * W_in_grad
            W_out -= learning_rate * W_out_grad.reshape(-1, 1)

            W_in_grad *= 0
            W_out_grad *= 0
            W_grad *= 0

# テスト
```

```

for s in range(x_test.shape[0]):
    z *= 0

    # sにおける正解データ
    d = d_test[s]

    xs = x_test[s]

    # 時系列ループ
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)

    #誤差
    loss = functions.mean_squared_error(d, y)
    print('loss:', loss, 'd:', d, 'y:', y)

original = np.full(maxlen, None)
pred_num = 200

xs = test_head

# sin波予測
for s in range(0, pred_num):
    z *= 0
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)
    original = np.append(original, y)
    xs = np.delete(xs, 0)
    xs = np.append(xs, y)

plt.figure()
plt.ylim([-1.5, 1.5])
plt.plot(np.sin(np.linspace(0, round_num* pred_num / div_num * np.pi, pred_num)), linestyle='dotted', color='aaaaaa')
plt.plot(original, linestyle='dashed', color='black')
plt.show()

```

In []:

```

# predict sin

-----

### [try]
- maxlenを5, iters_numを500, 3000 (※時間がかかる)にしよう

```

In []:

```
import sys, os
sys.path.append(os.pardir) # 親ディレクトリのファイルをインポートするための設定
import numpy as np
from common import functions
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

np.random.seed(0)

# sin曲線
round_num = 10
div_num = 500
ts = np.linspace(0, round_num * np.pi, div_num)
f = np.sin(ts)

def d_tanh(x):
    return 1/(np.cosh(x)**2 + 1e-4)

# ひとつの時系列データの長さ
maxlen = 5

# sin波予測の入力データ
test_head = [[f[k]] for k in range(0, maxlen)]

data = []
target = []

for i in range(div_num - maxlen):
    data.append(f[i: i + maxlen])
    target.append(f[i + maxlen])

X = np.array(data).reshape(len(data), maxlen, 1)
D = np.array(target).reshape(len(data), 1)

# データ設定
N_train = int(len(data) * 0.8)
N_validation = len(data) - N_train

x_train, x_test, d_train, d_test = train_test_split(X, D, test_size=N_validation)

input_layer_size = 1
hidden_layer_size = 5
output_layer_size = 1

weight_init_std = 0.01
learning_rate = 0.1

iters_num = 3000

# ウェイト初期化 (バイアスは簡単のため省略)
W_in = weight_init_std * np.random.randn(input_layer_size, hidden_layer_size)
W_out = weight_init_std * np.random.randn(hidden_layer_size, output_layer_size)
W = weight_init_std * np.random.randn(hidden_layer_size, hidden_layer_size)

# 勾配
W_in_grad = np.zeros_like(W_in)
W_out_grad = np.zeros_like(W_out)
W_grad = np.zeros_like(W)
```

```

us = []
zs = []

u = np.zeros(hidden_layer_size)
z = np.zeros(hidden_layer_size)
y = np.zeros(output_layer_size)

delta_out = np.zeros(output_layer_size)
delta = np.zeros(hidden_layer_size)

losses = []

# トレーニング
for i in range(iters_num):
    for s in range(x_train.shape[0]):
        us.clear()
        zs.clear()
        z *= 0

        # sにおける正解データ
        d = d_train[s]

        xs = x_train[s]

        # 時系列ループ
        for t in range(maxlen):

            # 入力値
            x = xs[t]
            u = np.dot(x, W_in) + np.dot(z, W)
            us.append(u)
            z = np.tanh(u)
            zs.append(z)

        y = np.dot(z, W_out)

        #誤差
        loss = functions.mean_squared_error(d, y)

        delta_out = functions.d_mean_squared_error(d, y)

        delta *= 0
        for t in range(maxlen)[::-1]:

            delta = (np.dot(delta, W.T) + np.dot(delta_out, W_out.T)) * d_tanh(us[t])

            # 勾配更新
            W_grad += np.dot(zs[t].reshape(-1, 1), delta.reshape(1, -1))
            W_in_grad += np.dot(xs[t], delta.reshape(1, -1))
            W_out_grad = np.dot(z.reshape(-1, 1), delta_out)

            # 勾配適用
            W -= learning_rate * W_grad
            W_in -= learning_rate * W_in_grad
            W_out -= learning_rate * W_out_grad.reshape(-1, 1)

        W_in_grad *= 0
        W_out_grad *= 0
        W_grad *= 0

```

```
# テスト
for s in range(x_test.shape[0]):
    z *= 0

    # sにおける正解データ
    d = d_test[s]

    xs = x_test[s]

    # 時系列ループ
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)

    #誤差
    loss = functions.mean_squared_error(d, y)
    print('loss:', loss, 'd:', d, 'y:', y)

original = np.full(maxlen, None)
pred_num = 200

xs = test_head

# sin波予測
for s in range(0, pred_num):
    z *= 0
    for t in range(maxlen):

        # 入力値
        x = xs[t]
        u = np.dot(x, W_in) + np.dot(z, W)
        z = np.tanh(u)

    y = np.dot(z, W_out)
    original = np.append(original, y)
    xs = np.delete(xs, 0)
    xs = np.append(xs, y)

plt.figure()
plt.ylim([-1.5, 1.5])
plt.plot(np.sin(np.linspace(0, round_num* pred_num / div_num * np.pi, pred_num)), linestyle='dotted', color='aaaaaa')
plt.plot(original, linestyle='dashed', color='black')
plt.show()
```