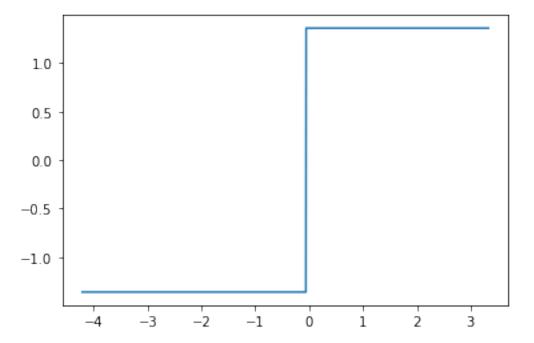
generate

March 27, 2022

[1]: %load_ext autoreload

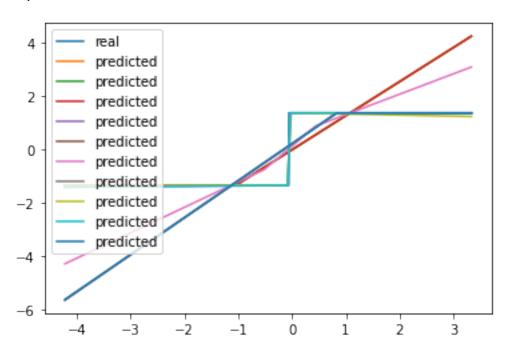
```
%autoreload 2
 [2]: import numpy as np
      import torch
      import torch.nn as nn
      import torch.nn.functional as F
      from torch.utils.data import TensorDataset, DataLoader
      import matplotlib.pyplot as plt
      import generate as generate
      from single_layer import *
     0.1 Experiment 1
        • Data Generation: R1 -> R1, with only one activation unit
        • model
            - hidden dim: 2
            - lr: 0.01
 [3]: # Constants
      d = 1
      N = 3\#int(np.exp(d))
      M = d
      num = 1
      T = 2000
      lr = 0.01
      hidden dim = 2
[10]: (an, bn) = generate_generate_activations(d, N)
      (In, thetan) = generate_generate_single_layer(N, M, d, num, an, bn)
      (X, Y) = generate_generate_single_data(T, an, bn, In, thetan)
      print(X.shape)
      print(Y.shape)
     (1, 2000, 1)
     (1, 2000)
```

```
[11]: plt.plot(*zip(*sorted(zip(X[0], Y[0]))))
   plt.show()
```



epochs: 366, validation loss: 0.6367917060852051 epochs: 400, validation loss: 0.6367545127868652 epochs: 392, validation loss: 0.6367880702018738 epochs: 1000, validation loss: 0.0249573215842247 epochs: 541, validation loss: 0.6444958448410034 epochs: 309, validation loss: 0.531389057636261

epochs: 414, validation loss: 0.6451221704483032 epochs: 1000, validation loss: 0.030945193022489548 epochs: 1000, validation loss: 0.024709703400731087 epochs: 310, validation loss: 0.6443562507629395



0.2 Experiment 2

- Data Generation: R1 -> R1, with only one activation unit
- model
 - -hidden_dim: 3
 - lr: 0.01

```
[13]: # Constants
d = 1
N = 3#int(np.exp(d))
M = d
num = 1
T = 2000

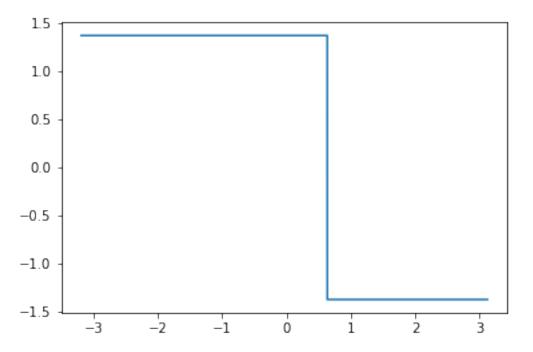
lr = 0.01
hidden_dim = 3
```

```
[14]: (an, bn) = generate.generate_activations(d, N)
   (In, thetan) = generate.generate_single_layer(N, M, d, num, an, bn)
   (X, Y) = generate.generate_single_data(T, an, bn, In, thetan)
   print(X.shape)
```

```
print(Y.shape)

(1, 2000, 1)
   (1, 2000)

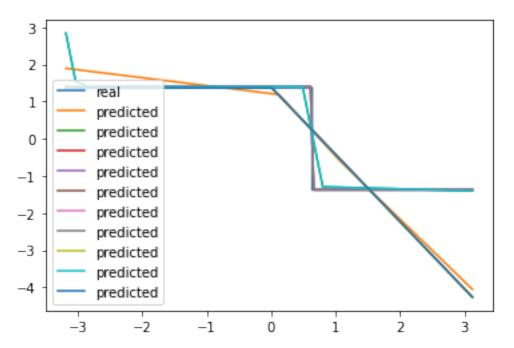
[15]: plt.plot(*zip(*sorted(zip(X[0], Y[0]))))
   plt.show()
```



```
[16]: num_experiments = 10
      input = X[0]
      plt.plot(*zip(*sorted(zip(X[0], Y[0]))))
      for i in range(num_experiments):
          (model, epoch_number, best_vloss) = train_one_model(
              hidden_dim, X[0], Y[0],
              val_ratio=0.2,
              lr=lr,
              patience=100,
              epochs=4000,
          )
          print(f"epochs: {epoch_number}, validation loss: {best_vloss}")
          predicted = model(torch.Tensor(X[0])).detach().numpy()
          plt.plot(*zip(*sorted(zip(X[0], predicted))))
      plt.legend(["real"] + ["predicted"] * num_experiments)
      plt.show()
```

epochs: 189, validation loss: 0.33814457058906555

```
epochs: 4000, validation loss: 0.04905490577220917 epochs: 120, validation loss: 0.3504645824432373 epochs: 4000, validation loss: 0.005515805445611477 epochs: 4000, validation loss: 0.005026805214583874 epochs: 3107, validation loss: 0.006231104023754597 epochs: 4000, validation loss: 0.0037900456227362156 epochs: 137, validation loss: 0.35054540634155273 epochs: 4000, validation loss: 0.04905082285404205 epochs: 118, validation loss: 0.3504469394683838
```



0.3 Experiment 3

- Data Generation: R1 -> R1, with only one activation unit
- model
 - hidden dim: 4
 - lr: 0.01

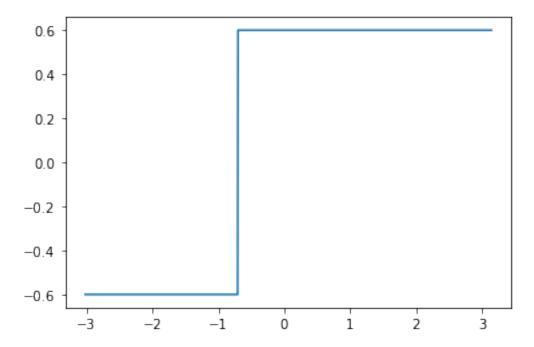
```
[17]: # Constants
d = 1
N = 3#int(np.exp(d))
M = d
num = 1
T = 2000

lr = 0.01
hidden_dim = 4
```

```
[18]: (an, bn) = generate.generate_activations(d, N)
    (In, thetan) = generate.generate_single_layer(N, M, d, num, an, bn)
    (X, Y) = generate.generate_single_data(T, an, bn, In, thetan)
    print(X.shape)
    print(Y.shape)

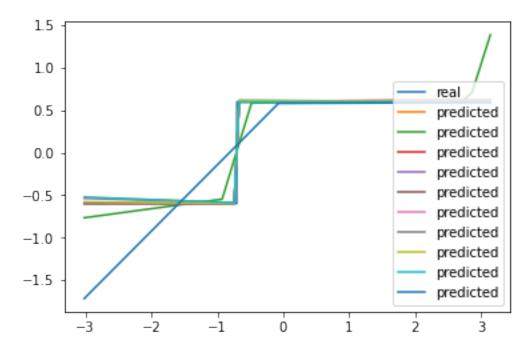
(1, 2000, 1)
    (1, 2000)

[19]: plt.plot(*zip(*sorted(zip(X[0], Y[0]))))
    plt.show()
```



```
plt.legend(["real"] + ["predicted"] * num_experiments)
plt.show()
```

```
epochs: 1000, validation loss: 0.0015389877371490002 epochs: 1000, validation loss: 0.015289440751075745 epochs: 1000, validation loss: 0.0015104215126484632 epochs: 1000, validation loss: 0.0012099889572709799 epochs: 1000, validation loss: 0.001457865466363728 epochs: 1000, validation loss: 0.001496882294304669 epochs: 1000, validation loss: 0.0015868310583755374 epochs: 1000, validation loss: 0.0013930064160376787 epochs: 1000, validation loss: 0.0015373507048934698 epochs: 153, validation loss: 0.05847661942243576
```



0.4 Experiment 4

- Data Generation: R1 -> R1, with two activation units
- model
 - -hidden_dim: 4
 - lr: 0.01

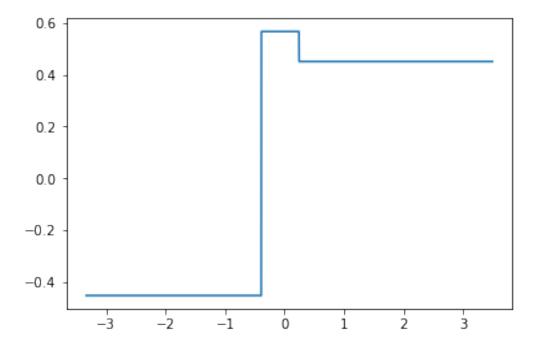
```
[21]: # Constants
d = 1
N = 3#int(np.exp(d))
M = 2
num = 1
```

```
T = 2000
lr = 0.01
hidden_dim = 4
```

```
[27]: (an, bn) = generate.generate_activations(d, N)
   (In, thetan) = generate.generate_single_layer(N, M, d, num, an, bn)
   (X, Y) = generate.generate_single_data(T, an, bn, In, thetan)
   print(X.shape)
   print(Y.shape)
```

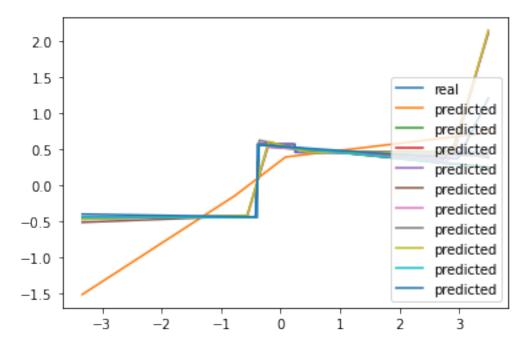
(1, 2000, 1) (1, 2000)

[28]: plt.plot(*zip(*sorted(zip(X[0], Y[0])))) plt.show()



```
epochs=2000,
)
print(f"epochs: {epoch_number}, validation loss: {best_vloss}")
predicted = model(torch.Tensor(X[0])).detach().numpy()
plt.plot(*zip(*sorted(zip(X[0], predicted))))
plt.legend(["real"] + ["predicted"] * num_experiments)
plt.show()
```

```
epochs: 108, validation loss: 0.061039477586746216 epochs: 2000, validation loss: 0.0026688140351325274 epochs: 2000, validation loss: 0.002557485830038786 epochs: 2000, validation loss: 0.001267323736101389 epochs: 1226, validation loss: 0.025260601192712784 epochs: 1423, validation loss: 0.02478194423019886 epochs: 2000, validation loss: 0.0017017138889059424 epochs: 1293, validation loss: 0.024264130741357803 epochs: 2000, validation loss: 0.002609127899631858 epochs: 2000, validation loss: 0.0044997879303991795
```



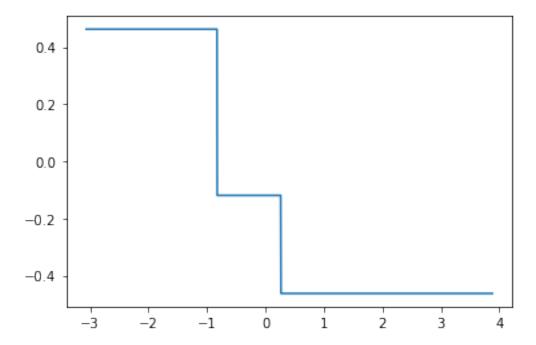
0.5 Experiment 5

- Data Generation: R1 -> R1, with two activation units
- model
 - hidden dim: 8
 - lr: 0.01

```
[30]: # Constants
d = 1
N = 3#int(np.exp(d))
M = 2
num = 1
T = 2000
lr = 0.01
hidden_dim = 8
[33]: (an, bn) = generate.generate_activations(d, N)
(In, thetan) = generate.generate_single_layer(N, M, d, num, an, bn)
(X, Y) = generate.generate_single_data(T, an, bn, In, thetan)
print(X.shape)
print(Y.shape)
```

(1, 2000, 1) (1, 2000)

```
[34]: plt.plot(*zip(*sorted(zip(X[0], Y[0]))))
plt.show()
```



```
[35]: num_experiments = 10
input = X[0]
plt.plot(*zip(*sorted(zip(X[0], Y[0]))))
for i in range(num_experiments):
```

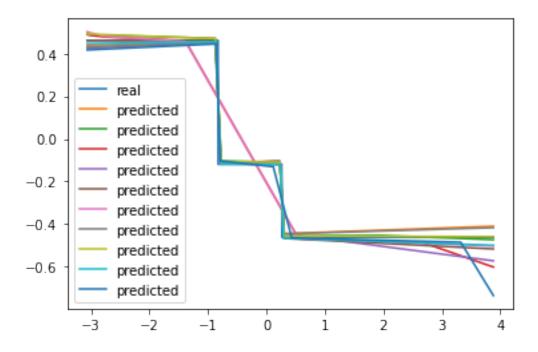
```
(model, epoch_number, best_vloss) = train_one_model(
     hidden_dim, X[0], Y[0],
     val_ratio=0.2,
     lr=lr,
     patience=100,
     epochs=2000,
)

print(f"epochs: {epoch_number}, validation loss: {best_vloss}")
     predicted = model(torch.Tensor(X[0])).detach().numpy()
     plt.plot(*zip(*sorted(zip(X[0], predicted))))

plt.legend(["real"] + ["predicted"] * num_experiments)

plt.show()
```

epochs: 1662, validation loss: 0.0006806793389841914 epochs: 1607, validation loss: 0.0006721923127770424 epochs: 259, validation loss: 0.012223339639604092 epochs: 2000, validation loss: 0.000542911235243082 epochs: 1738, validation loss: 0.0006583314971067011 epochs: 248, validation loss: 0.012197546660900116 epochs: 994, validation loss: 0.0008134039817377925 epochs: 1927, validation loss: 0.0006438980344682932 epochs: 1981, validation loss: 0.0005194177501834929 epochs: 2000, validation loss: 0.0017564221052452922



[]: