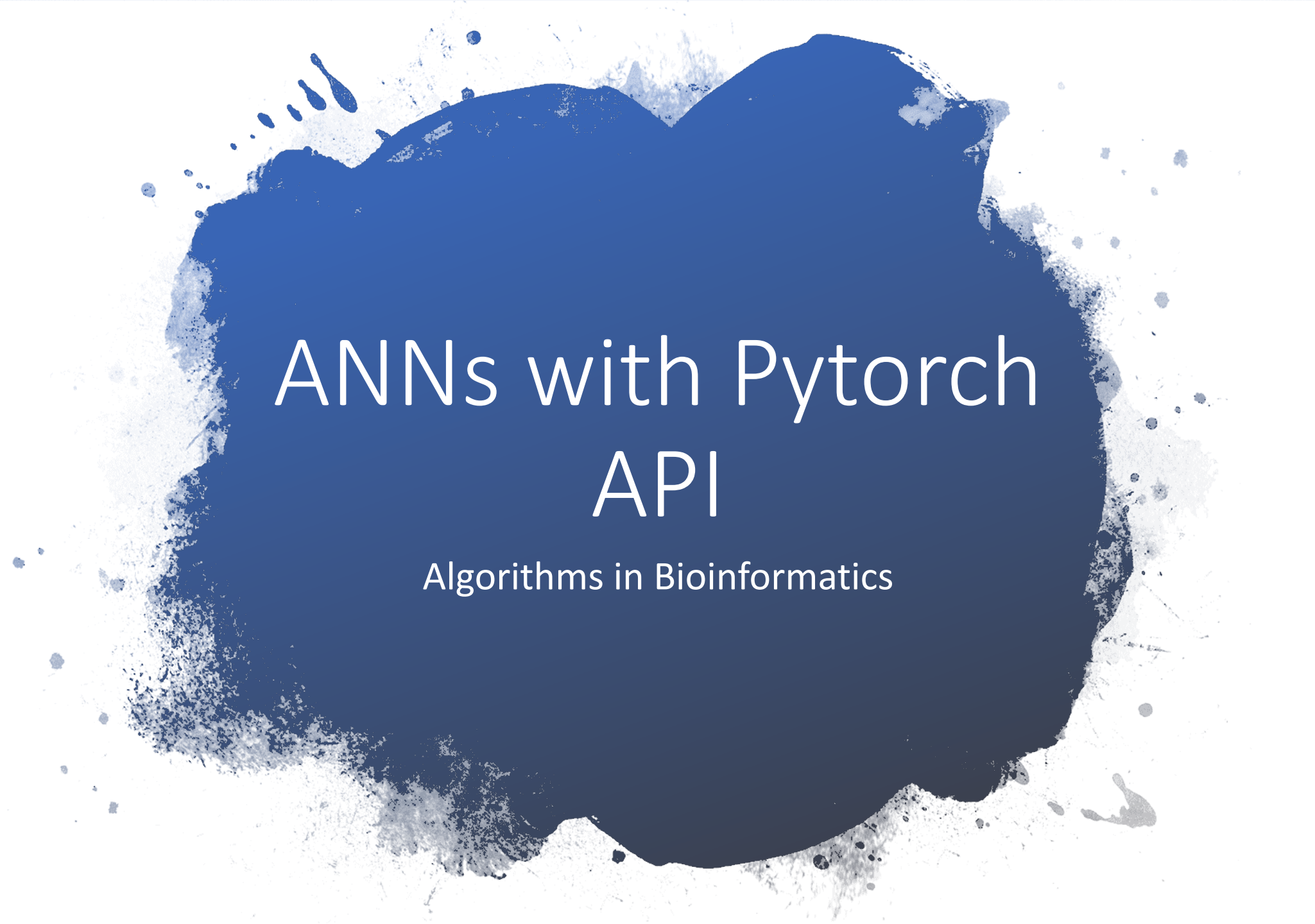




ANNs with Pytorch API

Algorithms in Bioinformatics

Who am I?



PhD student with Morten



Graduated in March 2019



Warning! I am no expert

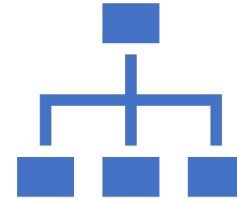
What's up for today?



Play with Pytorch



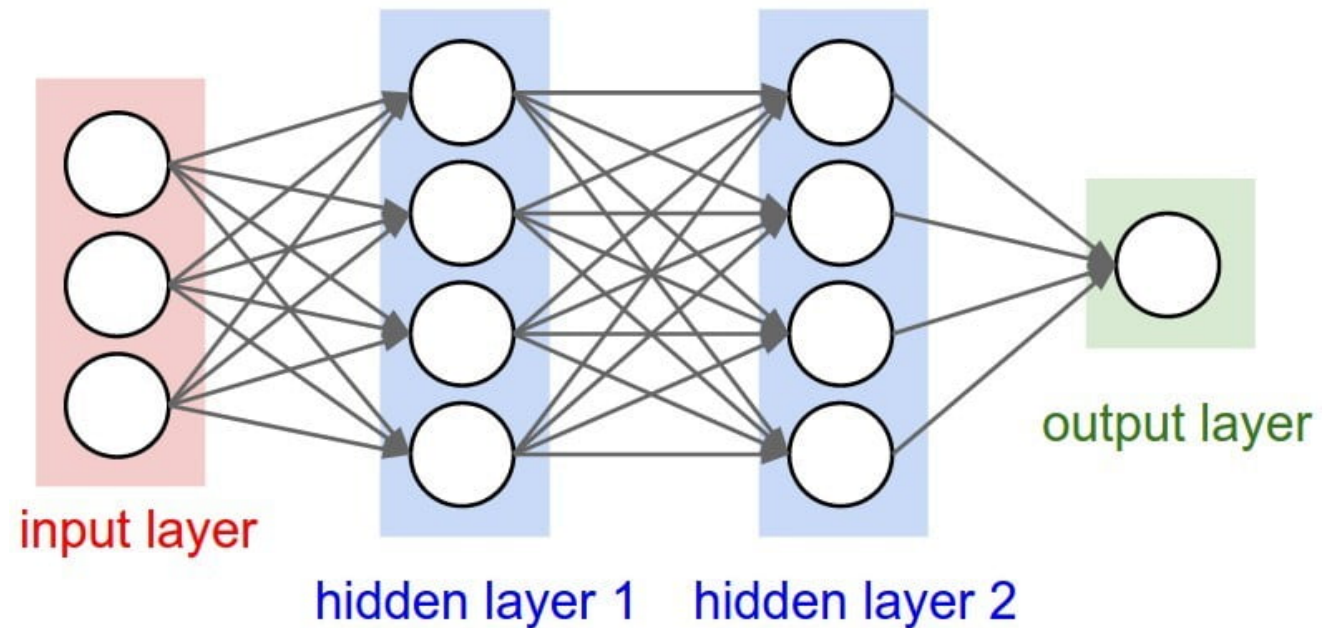
Why is Pytorch smart?



Other frameworks?

How to build an Artificial Neural Network?

- Invariant parameters
 - Input dimension
 - Output dimension
- Semi-variable
 - Loss-function
- Variable
 - # hidden layers
 - # hidden neurons
 - Activation functions
 - GD optimizations (learning rate, momentum, batch)
- Add-ons
 - Normalization
 - Regularization
 - Dropout



How do I build the best network?



You most
likely won't

Hyper-parameter search

Methods

- Manual Search
- Grid Search

```
from sklearn.model_selection import GridSearchCV
```

- Random Grid Search

```
from sklearn.model_selection import RandomizedSearchCV
```

Evaluation

- Evolutionary Algorithms
- Bayesian Hyper-parameter Evaluation

<https://github.com/HIPS/Spearmint>



Getting
started

Load Data

Build Model

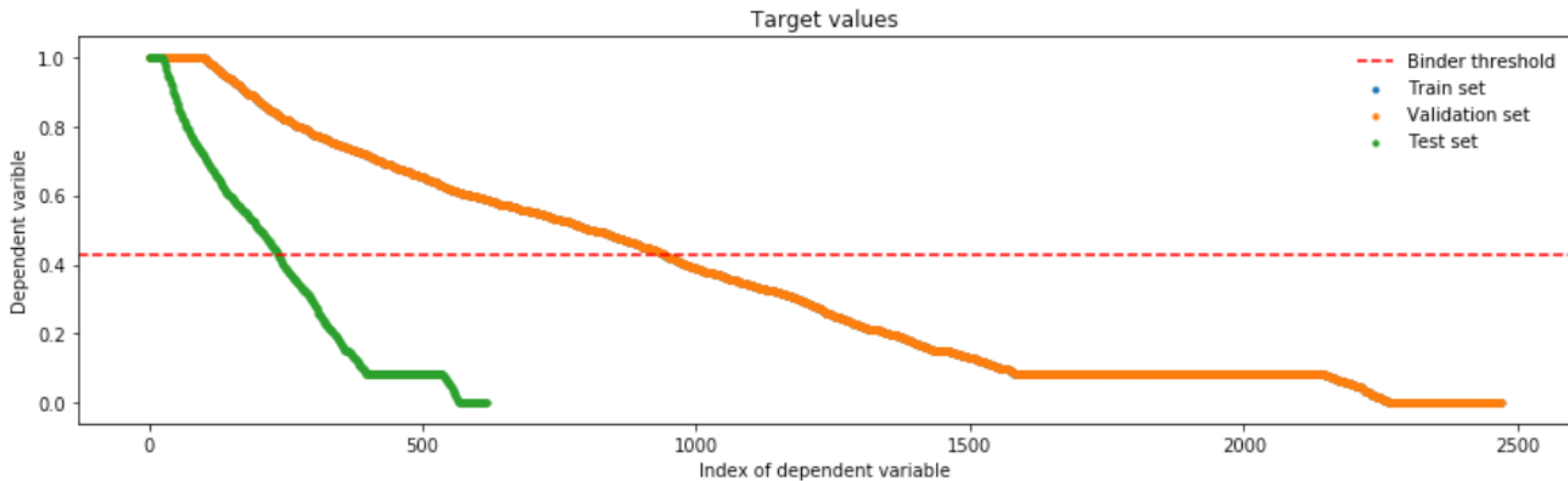
Select Hyper-parameters

Compile Model

Train Model

Evaluate Model

Load data



Build model

```
class Net(nn.Module):  
  
    def __init__(self, n_features, n_l1):  
        super(Net, self).__init__()  
        self.fc1 = nn.Linear(n_features, n_l1)  
        self.fc2 = nn.Linear(n_l1, 1)  
  
    def forward(self, x):  
        x = F.relu(self.fc1(x))  
        x = self.fc2(x)  
        return x
```

Compile model

```
net = Net(n_features, N_HIDDEN_NEURONS)  
  
optimizer = optim.SGD(net.parameters(),  
                        lr=LEARNING_RATE)  
criterion = nn.MSELoss()
```

Train Model

```
def train():
    train_loss, valid_loss = [], []

    early_stopping =
        EarlyStopping(patience=PATIENCE)

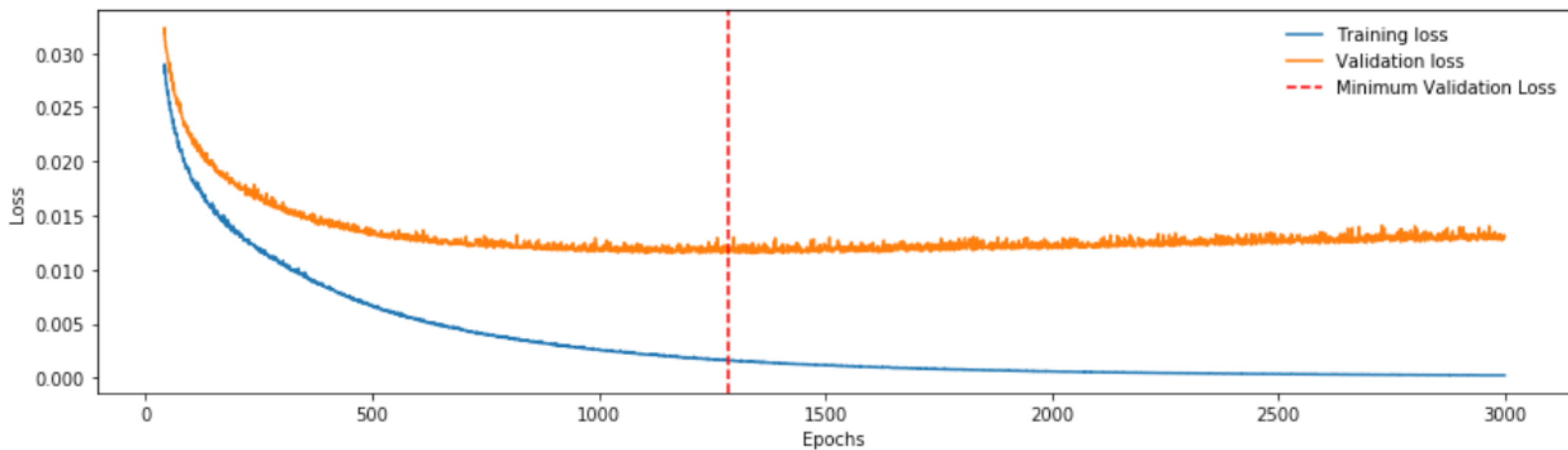
    for epoch in range(EPOCHS):
        net.train()
        pred = net(x_train)
        loss = criterion(pred, y_train)
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()
        train_loss.append(loss.data)

        batch_loss = 0
        net.eval()
        for x_valid, y_valid in valid_loader:
            pred = net(x_valid)
            loss = criterion(pred, y_valid)
            batch_loss += loss.data
        valid_loss.append(batch_loss /
                           len(valid_loader))

        if invoke(early_stopping,
                   valid_loss[-1], net,
                   implement=False):
            load_checkpoints()
            break

    return net, train_loss, valid_loss
```

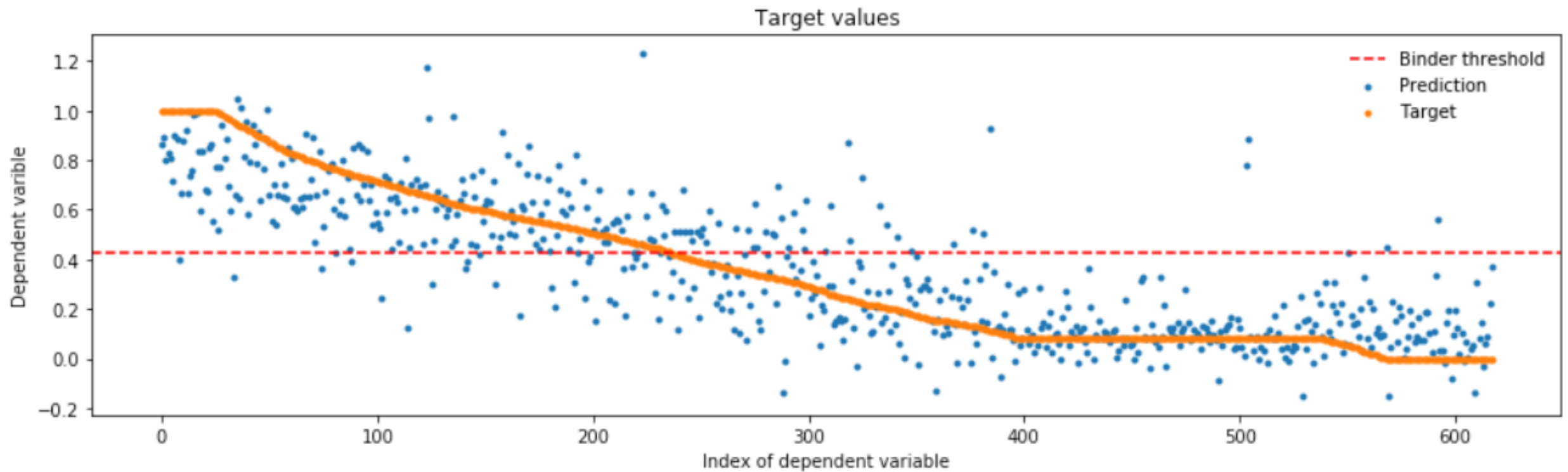
Load data



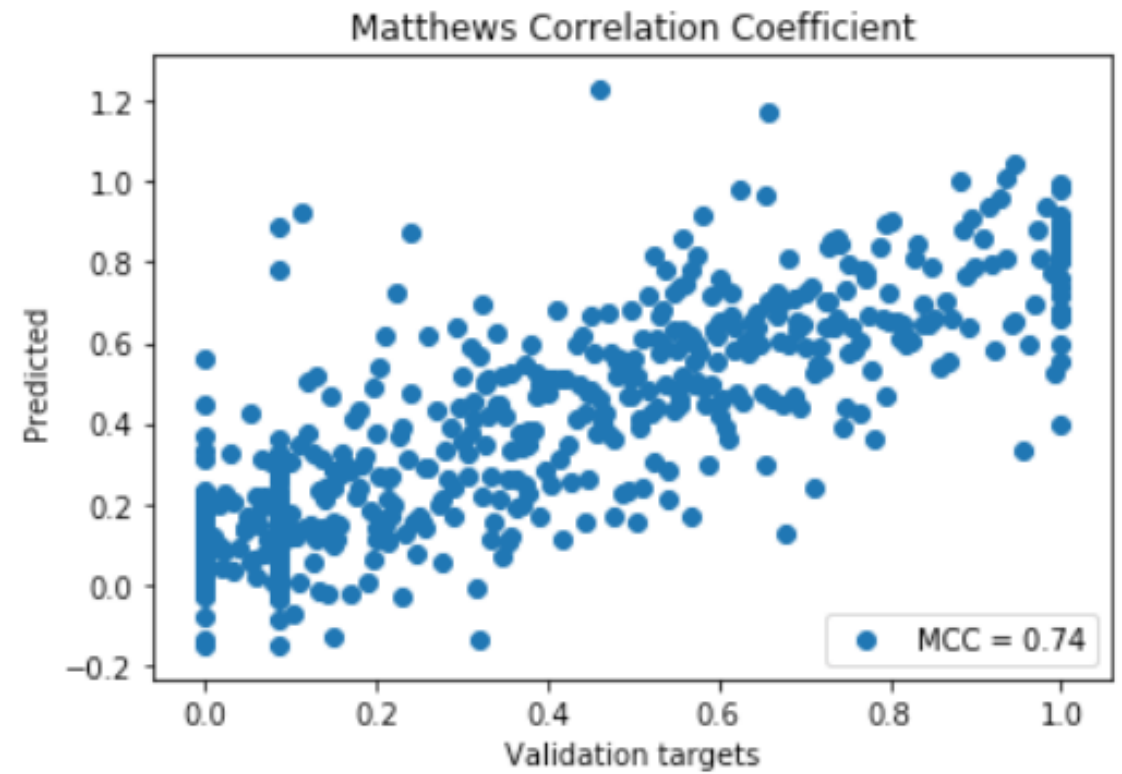
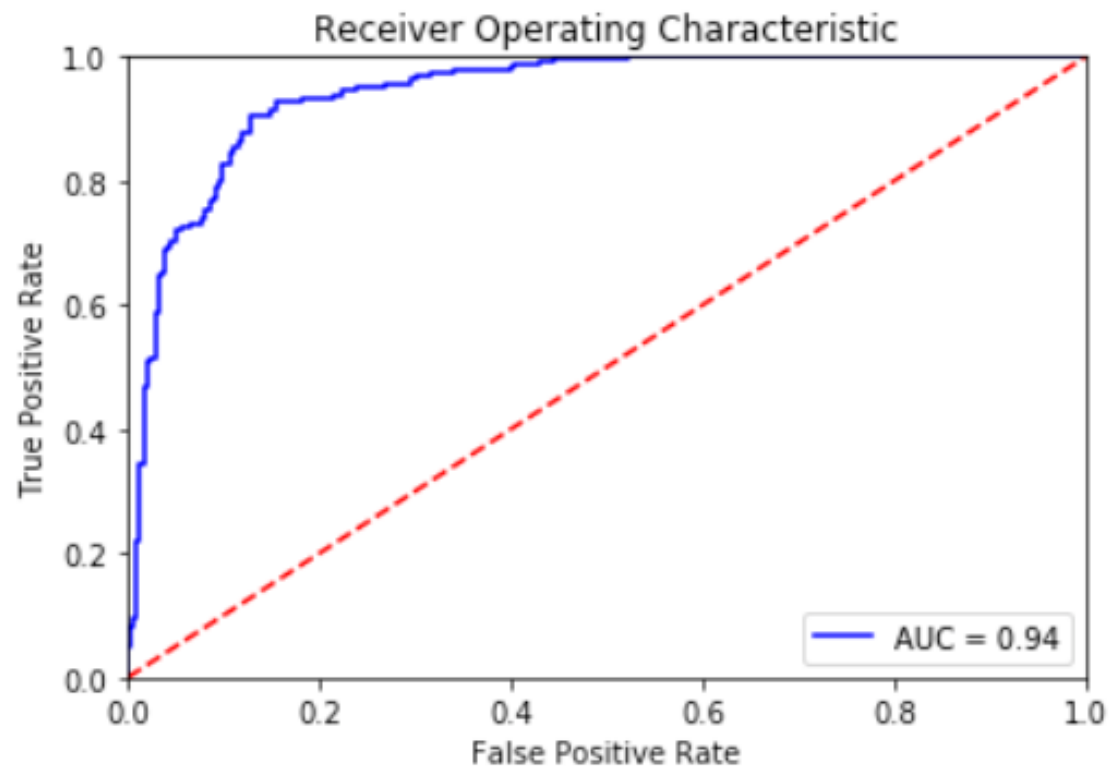
Evaluate model

```
net.eval()  
pred = net(x_test)  
loss = criterion(pred, y_test)
```

Evaluate model



Evaluate model



Avoid overfitting

- L2 Regularization

```
optim.SGD(net.parameters(), lr=LEARNING_RATE, weight_decay=0)
```

- L1 Regularization

Manual implementation using nn.L1Loss

- Dropout

```
Def __init__(): self.drop_layer = nn.Dropout(p=p)
```

```
Def Forward(): x = self.drop_layer(x)
```

- Batch normalization

```
Def __init__(): self.batch_norm = nn.BatchNorm1d(n_neurons)
```

```
Def Forward(): x = self.batch_norm(x)
```

The slide features a dark gray horizontal band across the middle. Above and below this band are decorative elements consisting of three overlapping semi-circles. The semi-circles are in two shades of blue: a lighter, semi-transparent blue and a darker, more solid blue. The overlapping areas create a pattern of varying blue tones.

Give it a go!