

Batch Normalization in PyTorch

```
class Net(nn.Module):
```

```
    def __init__(self, use_batch_norm, in_size = 784,  
                  hid_dim = 256, out_size = 10):
```

```
        super(Net, self).__init__()
```

usual
stuff

```
        self.in_size = in_size  
        self.hid_dim = hid_dim  
        self.out_size = out_size
```

```
        self.use_batch_norm = use_batch_norm → True / False
```

first
hidden
layer
with
optional
b-norm

```
        if use_batch_norm:
```

```
            self.fc1 = nn.Linear(in_size, hid_dim * 2, bias = False)
```

```
            self.batch_norm1 = nn.BatchNorm1d(hid_dim * 2)
```

```
        else:
```

```
            self.fc1 = nn.Linear(in_size, hid_dim * 2)
```

second
hidden
layer
with
optional
b-norm

```
        if use_batch_norm:
```

```
            self.fc2 = nn.Linear(hid_dim * 2, hid_dim, bias = False)
```

```
            self.batch_norm2 = nn.BatchNorm1d(hid_dim)
```

```
        else:
```

```
            self.fc2 = nn.Linear(hid_dim * 2, hid_dim)
```


`self.fc3 = nn.Linear(hid_dim, outsize)` → final FC

`def forward(self, x):`

`x = x.view(-1, 28*28)` → flatten

first
layer

`x = self.fc1(x)`

`if self.use_batch_norm:`

`x = self.batch_norm1(x)`

`x = F.relu(x)`

second
layer

`x = self.fc2(x)`

`if self.use_batch_norm:`

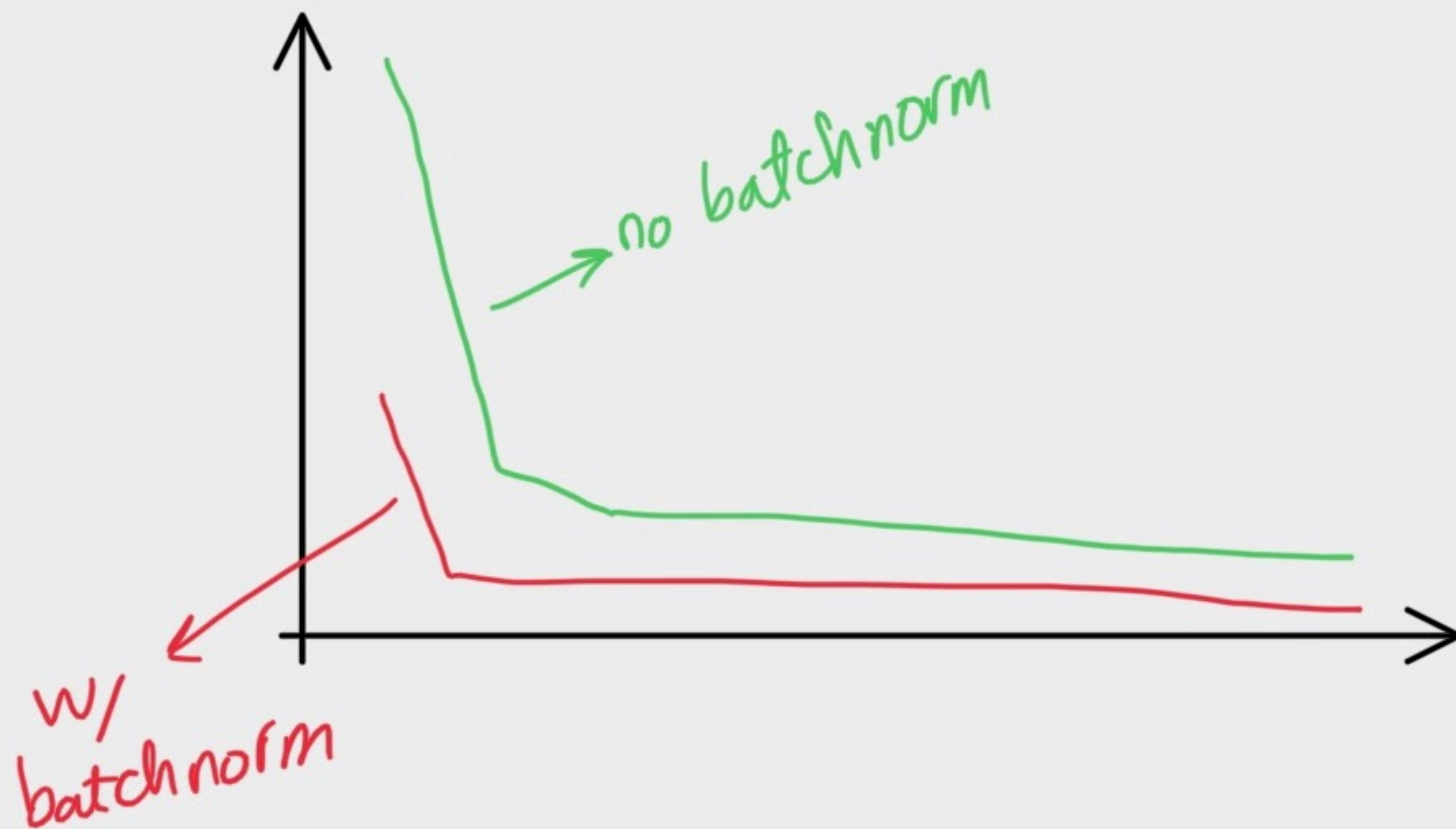
`x = self.batch_norm2(x)`

`x = F.relu(x)`

`x = self.fc3(x)` → final FC, no batch norm or activation

`return x`

How does it affect the training?



Summary

- Layers with batchnorm: bias = False
- BatchNorm1d, BatchNorm2d in PyTorch
- batchnorm layer: before activation function