

hyperparameter_tuning

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1 Classification of a Signal that Produces Higgs Boson Particles and background signals

2 Convolutional Neural Network

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[ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split, RandomizedSearchCV
from sklearn.metrics import classification_report, accuracy_score, roc_curve, auc, confusion_matrix
from sklearn.base import BaseEstimator, ClassifierMixin
from sklearn.decomposition import PCA
import torch
import torch.nn as nn
import torch.optim as optim
from torch.utils.data import DataLoader, TensorDataset, random_split
from tqdm import tqdm
from skorch import NeuralNetBinaryClassifier
from skorch import callbacks as cb
import optuna
name_dtype=np.array(['class_label', np.float32], ['jet_1_b-tag', np.float64],
                    ['jet_1_eta', np.float64], ['jet_1_phi', np.float64],
                    ['jet_1_pt', np.float64], ['jet_2_b-tag', np.float64],
                    ['jet_2_eta', np.float64], ['jet_2_phi', np.float64],
                    ['jet_2_pt', np.float64], ['jet_3_b-tag', np.float64],
                    ['jet_3_eta', np.float64], ['jet_3_phi', np.float64],
                    ['jet_3_pt', np.float64], ['jet_4_b-tag', np.float64],
                    ['jet_4_eta', np.float64], ['jet_4_phi', np.float64],
                    ['jet_4_pt', np.float64], ['lepton_eta', np.float64],
                    ['lepton_pT', np.float64], ['lepton_phi', np.float64],
                    ['m_bb', np.float64], ['m_jj', np.float64],
                    ['m_jjj', np.float64], ['m_jlv', np.float64],
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        ['m_lv', np.float64], ['m_wbb', np.float64],
        ['m_wvbb', np.float64], ['missing_energy_magnitude', np.float64],
        ['missing_energy_phi', np.float64]])
fullData=pd.read_csv('HIGGS.csv',header=None,names=name_dtype[:,0])
unscaled_X=fullData.drop(['class_label'],axis=1)
scaler=StandardScaler()
full_X=pd.DataFrame(scaler.fit_transform(unscaled_X.values),index=unscaled_X.
    ↪index,columns=unscaled_X.columns)
full_y=fullData['class_label']
X_train_df,X_test_df,y_train_df,y_test_df=train_test_split(full_X,full_y,test_size=0.
    ↪2,random_state=0)
pca = PCA(n_components=0.95) # retaining 95% of the variance
X_train_pca = pca.fit_transform(X_train_df)
X_test_pca = pca.transform(X_test_df)

# Checking the number of components selected and the amount of variance
    ↪explained
n_components = pca.n_components_
explained_variance = pca.explained_variance_ratio_.sum()

X_train=torch.tensor(X_train_pca).float()
X_test=torch.tensor(X_test_pca).float()
y_train=torch.tensor(y_train_df.values).float()
y_test=torch.tensor(y_test_df.values).float()
class DNN_NoDrop(nn.Module):
    def __init__(self, layer_sizes, activation):
        super(DNN_NoDrop, self).__init__()
        if layer_sizes == 'empty':
            layer_sizes = []
        else:
            layer_sizes = [int(size) for size in layer_sizes.split('_')]

        activation_functions = {'LeakyReLU': nn.LeakyReLU(), 'ReLU': nn.ReLU(),
    ↪'Tanh': nn.Tanh()}
        activation = activation_functions[activation]
        self.layers = nn.ModuleList()
        input_size = 23
        for hidden_size in layer_sizes:
            self.layers.append(nn.Linear(input_size, hidden_size))
            self.layers.append(activation)
            input_size = hidden_size

        self.layers.append(nn.Linear(input_size, 1))
        self.layers.append(nn.Sigmoid())

    def forward(self, x):
        for layer in self.layers:

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        x = layer(x)
        return x.squeeze()
device = 'cuda' if torch.cuda.is_available() else 'cpu'
X_train, y_train = X_train.to(device), y_train.to(device)
dataset = TensorDataset(X_train, y_train)

# Split dataset into training and validation sets
train_size = int(0.8 * len(dataset))
val_size = len(dataset) - train_size
train_dataset, val_dataset = random_split(dataset, [train_size, val_size])
import time
from torch.cuda.amp import autocast, GradScaler

def objective(trial):
    start_time = time.time()
    layer_sizes = trial.suggest_categorical('layer_sizes', ['empty', '16', '16_8', '64', '64_32', '64_32_16', '64_32_16_8', '128', '128_64', '128_64_32', '128_64_32_16', '128_64_32_16_8'])
    activation = trial.suggest_categorical('activation', ['LeakyReLU', 'ReLU', 'Tanh'])
    max_epochs = trial.suggest_categorical('max_epochs', [10, 25, 50, 100, 250])
    batch_size = 704
    lr = trial.suggest_float('lr', 1e-5, 1e-1, log=True)

    model = DNN_NoDrop(layer_sizes=layer_sizes, activation=activation)
    criterion = nn.BCEWithLogitsLoss()
    optimizer = optim.Adam(model.parameters(), lr=lr)
    model.to(device)
    scaler = GradScaler()

    train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True, num_workers=16)
    val_loader = DataLoader(val_dataset, batch_size=batch_size, shuffle=False, num_workers=16)

    for epoch in range(max_epochs):
        model.train()
        for batch, (input, target) in enumerate(train_loader):
            input, target = input.to(device), target.to(device)
            optimizer.zero_grad()

            with autocast():
                output = model(input)
                loss = criterion(output, target)
            scaler.scale(loss).backward()
            scaler.step(optimizer)
            scaler.update()

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model.eval()
val_loss = 0
correct = 0
with torch.no_grad(), autocast():
    for input, target in val_loader:
        input, target = input.to(device), target.to(device)
        output = model(input)
        val_loss += criterion(output, target).item()
        pred = torch.sigmoid(output).ge(0.5).view(-1)
        correct += pred.eq(target.view_as(pred)).sum().item()

val_loss /= len(val_loader.dataset)
accuracy = correct / len(val_loader.dataset)
trial.report(accuracy, epoch)

# Handle pruning based on the intermediate value
if trial.should_prune():
    raise optuna.exceptions.TrialPruned()

end_time = time.time()
duration = end_time - start_time
print(f"Trial {trial.number} took {duration:.2f} seconds.")

return accuracy

study = optuna.create_study(direction='maximize')
study.optimize(objective, n_trials=20)

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c:\Python39\lib\site-packages\tqdm\auto.py:21: TqdmWarning: IPProgress not found.
Please update jupyter and ipywidgets. See
https://ipywidgets.readthedocs.io/en/stable/user_install.html
from .autonotebook import tqdm as notebook_tqdm

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[ ]: best_trial = study.best_trial
best_params = best_trial.params

X_train, y_train = zip(*[data for data in train_dataset])
X_train, y_train = torch.stack(X_train), torch.tensor(y_train)
best_trial.fit(X_train, y_train)

```

Best trial parameters: {'layer_sizes': '112_56_28_14_7', 'activation': 'LeakyReLU', 'max_epochs': 250, 'batch_size': 200, 'lr': 0.00010022659726287869}

epoch	accuracy	train_loss	valid_acc	valid_loss	dur
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1	0.6577	0.6135	0.6895		
0.5854	33.3308				
2	0.7018	0.5711	0.7075		

0.5629	33.2186		
3	0.7131	0.5548	0.7168
0.5498	33.4217		
4	0.7202	0.5449	0.7216
0.5423	33.1873		
5	0.7244	0.5386	0.7253
0.5371	33.1697		
6	0.7279	0.5335	0.7281
0.5327	33.2076		
7	0.7309	0.5290	0.7307
0.5290	33.4589		
8	0.7335	0.5251	0.7329
0.5257	33.0833		
9	0.7355	0.5218	0.7350
0.5229	33.2814		
10	0.7372	0.5190	0.7361
0.5206	33.3182		
11	0.7390	0.5166	0.7375
0.5185	33.0533		
12	0.7404	0.5145	0.7388
0.5168	33.0976		
13	0.7418	0.5126	0.7395
0.5153	33.3232		
14	0.7429	0.5109	0.7405
0.5138	33.0845		
15	0.7438	0.5094	0.7412
0.5125	33.3679		
16	0.7446	0.5081	0.7422
0.5114	33.2192		
17	0.7454	0.5069	0.7428
0.5105	33.3223		
18	0.7460	0.5059	0.7436
0.5096	33.0912		
19	0.7464	0.5049	0.7441
0.5088	33.4119		
20	0.7471	0.5040	0.7446
0.5080	33.1403		
21	0.7476	0.5032	0.7448
0.5073	33.5221		
22	0.7480	0.5024	0.7453
0.5066	32.9688		
23	0.7485	0.5017	0.7455
0.5062	33.3719		
24	0.7490	0.5011	0.7458
0.5056	33.4424		
25	0.7494	0.5005	0.7461
0.5052	33.2093		
26	0.7497	0.4999	0.7465

0.5047	33.3468		
27	0.7500	0.4993	0.7469
0.5041	33.4303		
28	0.7504	0.4988	0.7473
0.5038	33.2018		
29	0.7508	0.4983	0.7476
0.5033	33.5672		
30	0.7509	0.4979	0.7478
0.5030	33.3906		
31	0.7514	0.4974	0.7479
0.5028	33.2646		
32	0.7516	0.4970	0.7480
0.5025	33.2547		
33	0.7519	0.4967	0.7481
0.5024	33.6310		
34	0.7522	0.4963	0.7483
0.5023	34.8152		
35	0.7524	0.4959	0.7487
0.5019	33.9008		
36	0.7526	0.4956	0.7488
0.5017	34.2353		
37	0.7528	0.4952	0.7488
0.5015	33.4609		
38	0.7531	0.4949	0.7491
0.5013	33.3208		
39	0.7533	0.4946	0.7492
0.5011	33.4447		
40	0.7535	0.4943	0.7494
0.5009	33.3671		
41	0.7538	0.4940	0.7497
0.5008	33.7089		
42	0.7539	0.4937	0.7498
0.5004	33.5423		
43	0.7541	0.4934	0.7499
0.5003	33.4545		
44	0.7543	0.4932	0.7501
0.5001	33.3203		
45	0.7545	0.4929	0.7502
0.4999	33.5867		
46	0.7546	0.4927	0.7502
0.4997	33.8412		
47	0.7548	0.4924	0.7504
0.4995	33.6954		
48	0.7550	0.4922	0.7506
0.4992	33.7670		
49	0.7553	0.4919	0.7506
0.4991	33.6493		
50	0.7554	0.4917	0.7510

0.4988	33.8705			
51	0.7555	0.4915	0.7512	
0.4986	33.7807			
52	0.7556	0.4913	0.7510	
0.4984	33.8562			
53	0.7558	0.4910	0.7512	
0.4983	33.8179			
54	0.7559	0.4908	0.7515	
0.4982	33.9671			
55	0.7560	0.4906	0.7517	
0.4978	34.7171			
56	0.7562	0.4904	0.7518	
0.4977	34.0144			
57	0.7563	0.4902	0.7520	
0.4976	34.0551			
58	0.7565	0.4900	0.7520	
0.4974	33.2860			
59	0.7566	0.4898	0.7520	
0.4974	33.2596			
60	0.7568	0.4896	0.7518	
0.4973	33.3637			
61	0.7568	0.4895	0.7519	
0.4971	33.9187			
62	0.7570	0.4893	0.7521	
0.4971	33.5873			
63	0.7572	0.4891	0.7521	
0.4970	34.0771			
64	0.7573	0.4889	0.7521	
0.4969	33.6861			
65	0.7573	0.4887	0.7522	
0.4967	33.5544			
66	0.7574	0.4886	0.7524	
0.4966	33.4595			
67	0.7575	0.4884	0.7523	
0.4964	33.7067			
68	0.7576	0.4882	0.7523	
0.4964	33.6068			
69	0.7577	0.4881	0.7523	
0.4963	33.7507			
70	0.7578	0.4879	0.7523	
0.4961	33.4291			
71	0.7579	0.4877	0.7524	0.4961
33.5620				
72	0.7581	0.4876	0.7527	
0.4960	32.8986			
73	0.7582	0.4874	0.7525	
0.4959	32.9185			
74	0.7583	0.4873	0.7526	

0.4959	33.2974				
	75	0.7584	0.4872	0.7528	
0.4958	33.1602				
	76	0.7585	0.4870	0.7528	
0.4957	32.7573				
	77	0.7587	0.4869	0.7529	
0.4957	33.2983				
	78	0.7587	0.4868	0.7527	
0.4957	33.6762				
	79	0.7589	0.4866	0.7530	
0.4955	33.0554				
	80	0.7590	0.4865	0.7529	
0.4954	33.4038				
	81	0.7590	0.4864	0.7528	
0.4953	33.0788				
	82	0.7591	0.4862	0.7529	
0.4952	33.1924				
	83	0.7592	0.4861	0.7530	
0.4951	33.2223				
	84	0.7592	0.4860	0.7530	0.4952
33.1058					
	85	0.7594	0.4858	0.7531	
0.4949	33.0031				
	86	0.7594	0.4857	0.7530	0.4950 33.1707
	87	0.7595	0.4856	0.7534	
0.4948	33.2052				
	88	0.7596	0.4854	0.7534	
0.4947	33.2935				
	89	0.7597	0.4853	0.7534	
0.4946	33.1718				
	90	0.7598	0.4852	0.7535	
0.4944	32.9304				
	91	0.7598	0.4851	0.7533	
0.4944	32.9716				
	92	0.7600	0.4849	0.7535	
0.4943	33.4579				
	93	0.7601	0.4848	0.7535	
0.4943	32.8633				
	94	0.7601	0.4847	0.7534	
0.4943	33.0351				
	95	0.7603	0.4846	0.7535	
0.4942	33.0710				
	96	0.7603	0.4845	0.7537	
0.4941	33.2753				
	97	0.7604	0.4843	0.7537	
0.4940	32.7136				
	98	0.7604	0.4842	0.7538	
0.4939	33.3731				

99	0.7606	0.4841	0.7538	
0.4938	32.7607			
100	0.7606	0.4840	0.7539	
0.4939	33.3309			
101	0.7607	0.4839	0.7539	
0.4937	33.0932			
102	0.7607	0.4838	0.7541	
0.4937	33.2370			
103	0.7609	0.4837	0.7542	
0.4936	33.5429			
104	0.7609	0.4836	0.7542	0.4935
33.3273				
105	0.7609	0.4835	0.7542	
0.4934	33.0516			
106	0.7611	0.4834	0.7541	0.4934
33.6356				
107	0.7610	0.4833	0.7541	0.4934 32.9384
108	0.7611	0.4832	0.7540	
0.4932	33.1146			
109	0.7612	0.4831	0.7543	
0.4930	33.2682			
110	0.7612	0.4830	0.7543	
0.4930	33.1630			
111	0.7613	0.4829	0.7543	
0.4929	33.2248			
112	0.7613	0.4828	0.7542	
0.4928	33.3828			
113	0.7614	0.4827	0.7543	0.4928
33.0257				
114	0.7615	0.4826	0.7544	
0.4927	33.2054			
115	0.7616	0.4825	0.7545	
0.4927	33.1183			
116	0.7617	0.4824	0.7545	
0.4926	32.9711			
117	0.7617	0.4823	0.7547	
0.4926	33.3141			
118	0.7617	0.4822	0.7548	
0.4924	33.2509			
119	0.7618	0.4821	0.7548	0.4926
33.1198				
120	0.7619	0.4820	0.7546	0.4925
33.2495				
121	0.7619	0.4820	0.7547	0.4924
33.1087				
122	0.7620	0.4819	0.7548	
0.4924	32.8823			
123	0.7620	0.4818	0.7547	0.4925

33.2705					
124	0.7621	0.4817	0.7549		
0.4923	33.1206				
125	0.7621	0.4816	0.7548	0.4923	
33.3332					
126	0.7621	0.4815	0.7549		
0.4922	33.1510				
127	0.7622	0.4814	0.7551		
0.4921	33.5232				
128	0.7623	0.4814	0.7552		
0.4921	32.9239				
129	0.7623	0.4813	0.7552	0.4921	
33.5846					
130	0.7625	0.4812	0.7554		
0.4919	33.0912				
131	0.7625	0.4811	0.7554		
0.4919	33.1003				
132	0.7626	0.4811	0.7556		
0.4918	33.0466				
133	0.7626	0.4810	0.7554	0.4918	
33.2930					
134	0.7627	0.4809	0.7558		
0.4916	32.9677				
135	0.7627	0.4808	0.7557		
0.4916	33.5999				
136	0.7626	0.4808	0.7556	0.4917	33.0545
137	0.7627	0.4807	0.7557		
0.4915	33.3725				
138	0.7628	0.4806	0.7558		
0.4914	33.4217				
139	0.7628	0.4805	0.7558	0.4915	
33.1685					
140	0.7629	0.4805	0.7559		
0.4915	33.1362				
141	0.7629	0.4804	0.7560		
0.4914	33.5220				
142	0.7630	0.4803	0.7558		
0.4914	33.0402				
143	0.7630	0.4803	0.7559		
0.4913	33.6703				
144	0.7630	0.4802	0.7559	0.4913	
33.4445					
145	0.7631	0.4801	0.7560		
0.4913	33.4219				
146	0.7631	0.4801	0.7560		
0.4912	33.3156				
147	0.7632	0.4800	0.7559		
0.4911	33.1754				

148	0.7632	0.4799	0.7561		
0.4911	33.4129				
149	0.7633	0.4799	0.7560	0.4912	
33.5381					
150	0.7633	0.4798	0.7562		
0.4911	33.1344				
151	0.7633	0.4797	0.7562		
0.4910	33.2029				
152	0.7633	0.4797	0.7561	0.4910	
33.0906					
153	0.7634	0.4796	0.7561		
0.4910	33.1365				
154	0.7635	0.4796	0.7560		
0.4910	33.5121				
155	0.7635	0.4795	0.7561	0.4910	33.3896
156	0.7635	0.4795	0.7562	0.4910	
33.0809					
157	0.7636	0.4794	0.7561	0.4910	
33.2163					
158	0.7636	0.4794	0.7561	0.4910	
33.3013					
159	0.7636	0.4793	0.7561	0.4909	
33.0286					
160	0.7637	0.4792	0.7561		
0.4908	33.5756				
161	0.7637	0.4792	0.7562		
0.4909	33.0609				
162	0.7638	0.4791	0.7561		
0.4908	34.2981				
163	0.7639	0.4791	0.7562		
0.4907	34.2185				
164	0.7638	0.4790	0.7562	0.4908	34.2539
165	0.7639	0.4790	0.7562	0.4908	
34.7693					
166	0.7639	0.4789	0.7561	0.4908	
36.0364					
167	0.7640	0.4789	0.7562	0.4908	
36.1572					
168	0.7641	0.4788	0.7560	0.4908	
36.1592					
169	0.7641	0.4788	0.7561		
0.4907	35.9536				
170	0.7640	0.4787	0.7559	0.4908	35.9349
171	0.7641	0.4786	0.7561	0.4907	
36.0666					
172	0.7642	0.4786	0.7560	0.4907	
35.9308					
173	0.7642	0.4785	0.7561		

0.4907	36.2332				
174	0.7643	0.4785	0.7560		
0.4906	36.2272				
175	0.7643	0.4784	0.7561	0.4906	35.8607
176	0.7643	0.4784	0.7562	0.4906	
36.3572					
177	0.7644	0.4783	0.7561		
0.4906	36.0230				
178	0.7644	0.4783	0.7561	0.4905	
36.1796					
179	0.7644	0.4782	0.7562	0.4906	
36.1551					
180	0.7644	0.4782	0.7562	0.4905	35.8282
181	0.7644	0.4781	0.7562	0.4906	
36.2939					
182	0.7645	0.4781	0.7561		
0.4905	36.0766				
183	0.7645	0.4781	0.7562	0.4905	
36.0645					
184	0.7645	0.4780	0.7561		
0.4903	36.1515				
185	0.7645	0.4780	0.7561	0.4904	
35.7540					
186	0.7646	0.4779	0.7562		
0.4903	36.0516				
187	0.7646	0.4779	0.7563	0.4903	
36.1428					
188	0.7646	0.4778	0.7563		
0.4903	35.7642				
189	0.7647	0.4778	0.7561	0.4903	
36.3163					
190	0.7648	0.4778	0.7562	0.4903	
36.0699					
191	0.7648	0.4777	0.7561	0.4903	36.0346
192	0.7648	0.4777	0.7562	0.4904	
36.2238					
193	0.7648	0.4776	0.7562	0.4904	
35.9624					
194	0.7649	0.4776	0.7561	0.4905	
36.2237					
195	0.7648	0.4775	0.7561	0.4905	36.0540
196	0.7648	0.4775	0.7560	0.4905	35.8101
197	0.7648	0.4775	0.7563	0.4904	36.1640
198	0.7649	0.4774	0.7562	0.4904	
36.0417					
199	0.7649	0.4774	0.7562	0.4905	
36.1296					
200	0.7650	0.4773	0.7562	0.4904	

36.2157					
201	0.7649	0.4773	0.7562	0.4903	35.7578
202	0.7650	0.4773	0.7563	0.4903	
36.1990					
203	0.7651	0.4772	0.7562	0.4903	
36.2754					
204	0.7651	0.4772	0.7562	0.4904	
36.0889					
205	0.7650	0.4771	0.7561	0.4904	36.1859
206	0.7650	0.4771	0.7562	0.4904	35.9681
207	0.7651	0.4771	0.7561	0.4904	
36.1567					
208	0.7652	0.4770	0.7561	0.4904	
36.4758					
209	0.7652	0.4770	0.7562	0.4903	
35.7312					
210	0.7652	0.4770	0.7562	0.4904	36.1003
211	0.7652	0.4769	0.7563	0.4903	
35.9962					
212	0.7651	0.4769	0.7563		
0.4902	36.0532				
213	0.7652	0.4769	0.7562	0.4903	
36.3398					
214	0.7653	0.4768	0.7562	0.4903	
35.8434					
215	0.7654	0.4768	0.7562	0.4903	
36.0920					
216	0.7654	0.4768	0.7563	0.4903	
36.3955					
217	0.7654	0.4767	0.7562	0.4903	
34.9830					
218	0.7655	0.4767	0.7562	0.4904	
35.1405					
219	0.7654	0.4767	0.7563	0.4903	34.1841
220	0.7654	0.4766	0.7562	0.4903	33.9491
221	0.7655	0.4766	0.7563		
0.4902	34.6490				
222	0.7656	0.4765	0.7563		
0.4902	34.6645				
223	0.7655	0.4765	0.7563	0.4902	34.7960
224	0.7656	0.4765	0.7564		
0.4902	34.8672				
225	0.7656	0.4764	0.7562	0.4902	34.8064
226	0.7655	0.4764	0.7562	0.4902	34.7409
227	0.7656	0.4764	0.7562	0.4903	
34.7281					
228	0.7656	0.4764	0.7562	0.4902	
34.6703					

229	0.7657	0.4763	0.7562	0.4902	
34.6803					
230	0.7656	0.4763	0.7563	0.4902	34.2622
231	0.7657	0.4763	0.7563	0.4902	34.9269
232	0.7657	0.4762	0.7564		
0.4902	33.9108				
233	0.7656	0.4762	0.7564		
0.4901	34.8315				
234	0.7657	0.4762	0.7565		
0.4901	34.8357				
235	0.7657	0.4761	0.7563	0.4902	
34.4304					
236	0.7657	0.4761	0.7566		
0.4900	34.5308				
237	0.7657	0.4761	0.7566		
0.4899	34.7119				
238	0.7657	0.4760	0.7566		
0.4899	34.9732				
239	0.7657	0.4760	0.7565	0.4900	34.6083
240	0.7658	0.4760	0.7564	0.4900	
34.7902					
241	0.7657	0.4760	0.7567	0.4900	
34.6877					
242	0.7659	0.4759	0.7567		
0.4899	34.5644				
243	0.7659	0.4759	0.7567		
0.4899	34.0632				
244	0.7660	0.4759	0.7566	0.4899	
33.8716					
245	0.7659	0.4758	0.7568	0.4899	
34.5290					
246	0.7659	0.4758	0.7569		
0.4898	35.1448				
247	0.7659	0.4758	0.7567	0.4898	
34.6044					
248	0.7659	0.4758	0.7567	0.4897	
34.4777					
249	0.7660	0.4757	0.7569		
0.4897	34.8220				
250	0.7661	0.4757	0.7570		
0.4897	34.9404				

```

-----
TypeError                                Traceback (most recent call last)
c:\Users\matth\Desktop\COMP 4531\final.ipynb Cell 12 line <cell line: 23>()
    <a href='vscode-notebook-cell:/c%3A/Users/matth/Desktop/COMP%204531/final.
↪ipynb#X24sZmlsZQ%3D%3D?line=20'>21</a> X_train, y_train = torch.
↪stack(X_train), torch.tensor(y_train)

```

```

    <a href='vscode-notebook-cell:/c%3A/Users/matth/Desktop/COMP%204531/final.
    ↪ipynb#X24sZmlsZQ%3D%3D?line=21'>22</a> best_model.fit(X_train, y_train)
--> <a href='vscode-notebook-cell:/c%3A/Users/matth/Desktop/COMP%204531/final.
    ↪ipynb#X24sZmlsZQ%3D%3D?line=22'>23</a> best_model.save_params(f='best_model.
    ↪pth')

```

File c:\Python39\lib\site-packages\skorch\net.py:2488, in NeuralNet.

```

    ↪save_params(self, f_params, f_optimizer, f_criterion, f_history,
    ↪use_safetensors, **kwargs)
    2485     def _save_state_dict(state_dict, f_name):
    2486         torch.save(module.state_dict(), f_name)
-> 2488 kwargs_module, kwargs_other = _check_f_arguments(
    2489     'save_params',
    2490     f_params=f_params,
    2491     f_optimizer=f_optimizer,
    2492     f_criterion=f_criterion,
    2493     f_history=f_history,
    2494     **kwargs)
    2496 if not kwargs_module and not kwargs_other:
    2497     if self.verbose:

```

File c:\Python39\lib\site-packages\skorch\utils.py:755, in

```

    ↪_check_f_arguments(caller_name, **kwargs)
    753 for key, val in kwargs.items():
    754     if not key.startswith('f_'):
--> 755         raise TypeError(
    756             "{name} got an unexpected argument '{key}', did you mean
    ↪'f_{key}'?"
    757             .format(name=caller_name, key=key))
    759 if val is None:
    760     continue

```

TypeError: save_params got an unexpected argument 'f', did you mean 'f_f'?

```

[ ]: # Above error was fixed below, didn't want to re-run code above as it to 20
    ↪hours.
import pickle

with open("study.pkl", "wb") as f:
    pickle.dump(study, f)
#best_model.save_params(f='best_model.pth')

```