Table 1

Model	Train jup notebook	Score(s)	Dataset
./state_dict15c.pt	c-15-predict- normalized- varience-not- randomness	Valid: 0.780438 Test: MSE loss: 0.57919404 MAE loss: 0.55981259 KLDiv loss: -0.06383617	BATCH_SIZE = 32 SEQUENCE_SIZE = 24 TARGETs_NAMEs = [] dataset=CryptoDataset(predict_delta=1, batch_size=BATCH_SIZE, sequence_size=SEQUENCE_SIZE)
./state_dict15c2.pt	c2-15- predict- normalized- varience-not- randomness	Valid: 0.786? Test: MSE loss: 0.62601354 MAE loss: 0.58926985 KLDiv loss: -0.09573120	BATCH_SIZE = 32 SEQUENCE_SIZE = 24 TARGETs_NAMEs = [] dataset=CryptoDataset(predict_delta=1,
./state_dict15c5.pt	c5-15- predict- normalized- varience-not- randomness- True	Valid: 0.635867 Test: MSE loss: 0.42487826 MAE loss: 0.52023978 KLDiv loss: -0.11454240	BATCH_SIZE = 32 SEQUENCE_SIZE = 24 TARGETS_NAMEs = [] dataset=CryptoDataset(predict_delta=1, batch_size=BATCH_SIZE, sequence_size=SEQUENCE_SIZE) clipped

Model	Model arch	Model forward
./state_dict15c.pt	<pre>SentimentNet((liniar): Linear(in_features=1829, out_features=3658, bias=True) (normalization_pre_lstm): BatchNorm1d(24, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True) (lstm): LSTM(3658, 1829, num_layers=4, batch_first=True, dropout=0.4) (fc): Sequential((0): Linear(in_features=1829, out_features=914, bias=True) (1): BatchNorm1d(914, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True) (2): LeakyReLU(negative_slope=0.01) (3): Dropout(p=0.4, inplace=False) (4): Linear(in_features=914, out_features=1, bias=True)) }</pre>	# Initialize hidden state with zeros h0 = torch.zeros(self.n_layers, x.size(0), self.hidden_dim).requires_grad_().to(self.device) # Initialize cell state c0 = torch.zeros(self.n_layers, x.size(0), self.hidden_dim).requires_grad_().to(self.device) out = self.liniar(x) out = self.normalization_pre_lstm(out) out = torch.tanh(out) out, (hn, cn) = self.lstm(out, (h0, c0)) out = torch.add(out, x) # Index hidden state of last time step out = self.fc(out[:, -1, :])
./state_dict15c2.pt	SentimentNet((pre_lstm): Sequential((0): Linear(in_features=1829, out_features=3658, bias=True) (1): BatchNorm1d(24, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True) (2): Tanh() (3): Linear(in_features=3658, out_features=1829, bias=True) (4): BatchNorm1d(24, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True) (5): Tanh()) (Istm): LSTM(1829, 1829, num_layers=4, batch_first=True, dropout=0.6) (fc): Sequential((0): Linear(in_features=1829, out_features=914, bias=True) (1): BatchNorm1d(914, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True) (2): LeakyReLU(negative_slope=0.01) (3): Dropout(p=0.6, inplace=False) (4): Linear(in_features=914, out_features=1829, bias=True) (5): BatchNorm1d(1829, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True) (6): LeakyReLU(negative_slope=0.01) (7): Dropout(p=0.6, inplace=False) (8): Linear(in_features=1829, out_features=1, bias=True)))	# Initialize hidden state with zeros h0 = torch.zeros(self.n_layers, x.size(0), self.hidden_dim).requires_grad_().to(self.device) # Initialize cell state c0 = torch.zeros(self.n_layers, x.size(0), self.hidden_dim).requires_grad_().to(self.device) out = self.pre_lstm(x) lstm_out, (_, _) = self.lstm(out, (h0, c0)) # resblock out = torch.add(lstm_out, out) # Index hidden state of last time step out = self.fc(out[:, -1, :])
./state_dict15c5.pt	<pre>SentimentNet((liniar): Linear(in_features=1829, out_features=3658, bias=True) (normalization_pre_lstm): BatchNormld(24, eps=le-05, momentum=0.1, affine=True, track_running_stats=True) (lstm): LSTM(3658, 1829, num_layers=4, batch_first=True, dropout=0.4) (fc): Sequential((0): Linear(in_features=1829, out_features=914, bias=True) (1): BatchNormld(914, eps=le-05, momentum=0.1, affine=True, track_running_stats=True) (2): LeakyReLU(negative_slope=0.01) (3): Dropout(p=0.4, inplace=False) (4): Linear(in_features=914, out_features=1, bias=True)) }</pre>	# Initialize hidden state with zeros h0 = torch.zeros(self.n_layers, x.size(0), self.hidden_dim).requires_grad_().to(self.device) # Initialize cell state c0 = torch.zeros(self.n_layers, x.size(0), self.hidden_dim).requires_grad_().to(self.device) out = self.liniar(x) out = self.normalization_pre_lstm(out) out = torch.tanh(out) out, (hn, cn) = self.lstm(out, (h0, c0)) out = torch.add(out, x) # Index hidden state of last time step out = self.fc(out[:, -1, :])

Model	Test visualizer
./state_dict15c.pt	390 -50 -300 -300 270 280 290 300 310
./state_dict15c2.pt	200 150 100 50 100 1
./state_dict15c5.pt	1.5 - 1.6 - 1.79 200 210 220 233 240