bnn code

December 20, 2023

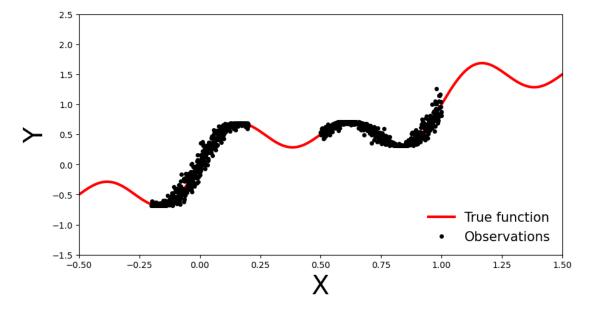
```
[]: import torch
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import logging
     import seaborn as sns
     import matplotlib.pyplot as plt
[]: import pyro
     import pyro.distributions as dist
     from pyro.nn import PyroModule, PyroSample
     import torch.nn as nn
     from pyro.infer.autoguide import AutoDiagonalNormal
     from pyro.infer import SVI, Trace_ELBO, Predictive
     from pyro.infer import MCMC, HMC, NUTS
     from tqdm.auto import trange
[]: import GPy
```

1 Generate noisy observations from sinusoidal function

```
ylims = [-1.5, 2.5]

# Create plot
fig, ax = plt.subplots(figsize=(10, 5))
ax.plot(x_true, y_true, 'r-', linewidth=3, label="True function")
ax.plot(x_obs, y_obs, 'ko', markersize=4, label="Observations")
ax.set_xlim(xlims)
ax.set_ylim(ylims)
ax.set_ylim(ylims)
ax.set_ylabel("X", fontsize=30)
ax.set_ylabel("Y", fontsize=30)
ax.legend(loc=4, fontsize=15, frameon=False)

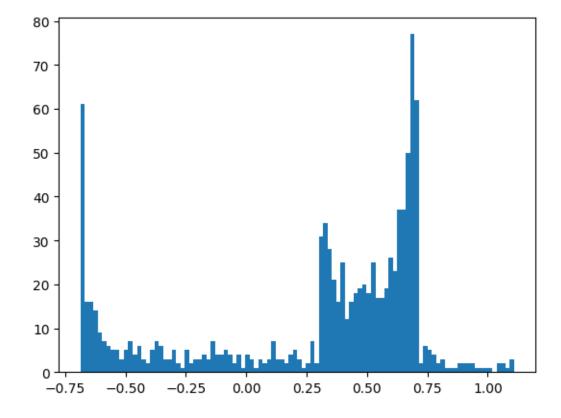
plt.show()
```



```
[]: plt.hist(y_obs, bins=100)
[]: (array([61., 16., 16., 14., 9., 7., 6., 5., 5.,
                                                      3.,
                                                           5.,
                                                               7.,
            6., 3., 2., 5., 7., 6.,
                                        3., 3.,
                                                 5.,
                                                      2.,
            3., 3., 4., 3., 7., 4., 4., 5., 4., 2.,
            3., 1., 3., 2., 3., 7., 3., 3., 2., 4., 5., 3.,
            2., 7., 2., 31., 34., 28., 21., 16., 25., 12., 16., 18., 19.,
            20., 18., 25., 17., 17., 19., 26., 23., 37., 37., 50., 77., 62.,
            2., 6., 5., 4., 2., 3., 1., 1., 1., 2., 2., 2., 2.,
            1., 1., 1., 1., 0., 2., 2., 1., 3.]),
     array([-0.68630667, -0.66835209, -0.65039751, -0.63244293, -0.61448834,
            -0.59653376, -0.57857918, -0.5606246, -0.54267002, -0.52471544,
           -0.50676085, -0.48880627, -0.47085169, -0.45289711, -0.43494253,
```

```
-0.41698795, -0.39903336, -0.38107878, -0.3631242, -0.34516962,
-0.32721504, -0.30926046, -0.29130587, -0.27335129, -0.25539671,
-0.23744213, -0.21948755, -0.20153297, -0.18357838, -0.1656238,
-0.14766922, -0.12971464, -0.11176006, -0.09380548, -0.07585089,
-0.05789631, -0.03994173, -0.02198715, -0.00403257,
                                                     0.01392202,
 0.0318766 , 0.04983118, 0.06778576,
                                        0.08574034,
                                                      0.10369492,
 0.12164951,
              0.13960409,
                           0.15755867,
                                        0.17551325,
                                                      0.19346783,
 0.21142241,
              0.229377 ,
                           0.24733158,
                                        0.26528616,
                                                      0.28324074,
                                        0.35505907,
 0.30119532,
              0.3191499 ,
                           0.33710449,
                                                      0.37301365,
 0.39096823,
              0.40892281,
                           0.42687739,
                                        0.44483198,
                                                      0.46278656,
 0.48074114,
              0.49869572,
                           0.5166503 ,
                                        0.53460488,
                                                      0.55255947,
 0.57051405,
                           0.60642321,
              0.58846863,
                                        0.62437779,
                                                      0.64233237,
 0.66028696,
              0.67824154,
                           0.69619612,
                                        0.7141507 ,
                                                      0.73210528,
              0.76801445,
                                        0.80392361,
 0.75005986,
                           0.78596903,
                                                      0.82187819,
                                        0.89369652,
 0.83983277,
              0.85778735,
                           0.87574194,
                                                      0.9116511 ,
 0.92960568,
              0.94756026,
                           0.96551485,
                                        0.98346943,
                                                      1.00142401,
 1.01937859,
              1.03733317,
                                        1.07324234,
                           1.05528775,
                                                      1.09119692,
 1.1091515]),
```

<BarContainer object of 100 artists>)



Utility function to plot predictions, prediction mean, true function and the observed data

```
[]: def plot_predictions(preds):
         y_pred = preds['obs'].T.detach().numpy().mean(axis=1)
         y_std = preds['obs'].T.detach().numpy().std(axis=1)
         fig, ax = plt.subplots(figsize=(10, 5))
         xlims = [-0.5, 1.5]
         ylims = [-1.5, 2.5]
         plt.xlim(xlims)
         plt.ylim(ylims)
         plt.xlabel("X", fontsize=30)
         plt.ylabel("Y", fontsize=30)
         ax.plot(x_true, y_true, 'b-', linewidth=3, label="true function")
         ax.plot(x_obs, y_obs, 'ko', markersize=4, label="observations")
         # ax.plot(x_obs, y_obs, 'ko', markersize=3)
         ax.plot(x_test, y_pred, '-', linewidth=3, color="#408765", __
      ⇔label="predictive mean")
         ax.fill_between(x_test, y_pred - 2 * y_std, y_pred + 2 * y_std, alpha=0.6,_
      ⇔color='#86cfac', zorder=5)
         plt.legend(loc=4, fontsize=15, frameon=False)
```

2 BNN with Gaussian prior and likelihood

```
[]: class Model(PyroModule):
         def __init__(self,h1, h2, prior_sigma):
             super().__init__()
             Define the model for the BNN
             param h1: number of neurons in the first hidden layer
             param h2: number of neurons in the second hidden layer
             param prior_sigma: standard deviation of the prior distribution of the ⊔
      \negweights and biases
             self.fc1 = PyroModule[nn.Linear](1, h1) # input to hidden layer
             self.fc2 = PyroModule[nn.Linear](h1, h2) # hidden layer 1 to hidden_
      ⇒layer 2
             self.fc3 = PyroModule[nn.Linear](h2, 1) # hidden layer 2 to output layer
             # set priors for all weights and biases
             self.fc1.weight = PyroSample(dist.Normal(0., prior_sigma).expand([h1,_u
      \hookrightarrow1]).to event(2))
             self.fc1.bias = PyroSample(dist.Normal(0., prior_sigma).expand([h1]).
      ⇔to_event(1))
```

```
self.fc2.weight = PyroSample(dist.Normal(0., prior_sigma).expand([h2,__
\rightarrowh1]).to_event(2))
       self.fc2.bias = PyroSample(dist.Normal(0., prior_sigma).expand([h2]).
→to event(1))
       self.fc3.weight = PyroSample(dist.Normal(0., prior_sigma).expand([1,__
\rightarrowh2]).to_event(2))
       self.fc3.bias = PyroSample(dist.Normal(0., prior_sigma).expand([1]).
⇔to_event(1))
       self.relu = nn.ReLU()
  def forward(self, x, y=None):
       x = x.reshape(-1, 1)
       x = self.relu(self.fc1(x))
       x = self.relu(self.fc2(x))
       mu = self.fc3(x).squeeze() # output of the neural network: mu = __
\neg neural\_net\_theta(x_i)
       sigma = pyro.sample("sigma", dist.Uniform(0., 1.)) # prior on \sigma_\sigma
\hookrightarrow (std dev of posterior of y; y/x, \hat{y}) of p(y_i \mid x_i, \hat{z})
\neg neural\_net\_theta(x_i), \sigma^2
       # sample y (observed) from the normal distribution with mean m u_{\sf L}
\hookrightarrow (neural_net_theta(x_i)) and std dev \sigma
       with pyro.plate("data", x.shape[0]):
           obs = pyro.sample("obs", dist.Normal(mu, sigma), obs=y) # y_i \sim N(y_i)
\neg neural\_net\_theta(x_i), \sigma^2
       return mu
```

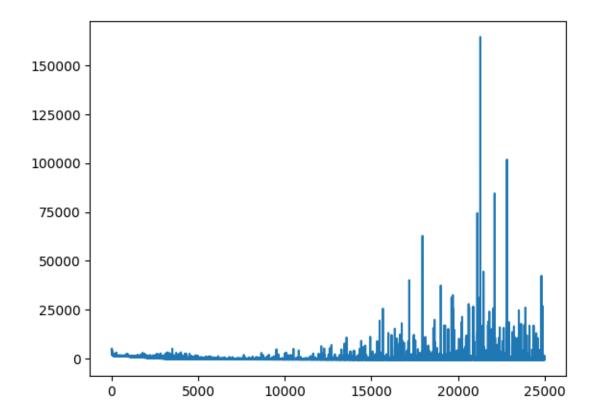
prior of the : dictated by the problem! network must be strictly convex in some cases typically use normal distribution prior: play around with the prior : influence of prior (gamma and lognormla and uniform)

```
[]: pyro.clear_param_store()
```

2.1 1.Training BNN (SVI)

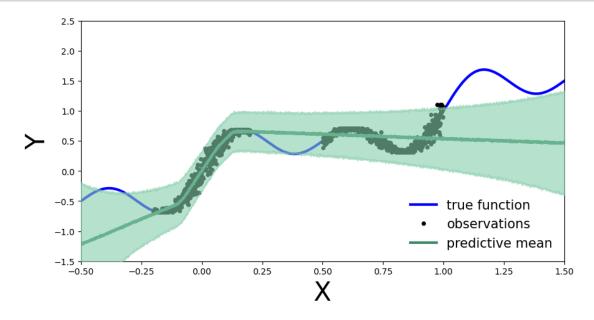
```
pyro.clear_param_store()
     num_epochs = 25000
     bar = trange(num_epochs)
     x_train = torch.from_numpy(x_obs).float()
     y_train = torch.from_numpy(y_obs).float()
     loss_svi = []
     for epoch in bar:
         elbo_loss = svi.step(x_train, y_train)
         bar.set_postfix(loss=f'{elbo_loss / x_obs.shape[0]:.3f}')
         loss_svi.append(elbo_loss)
         if epoch % 1000 == 0:
             logging.info("Elbo loss: {}".format(elbo_loss))
             print("Elbo loss: {}".format(elbo_loss))
      0%1
                   | 0/25000 [00:00<?, ?it/s]
    Elbo loss: 4990.442985653877
    Elbo loss: 1123.653981089592
    Elbo loss: 1240.5759227275848
    Elbo loss: 945.5774906277657
    Elbo loss: -137.5119373202324
    Elbo loss: -244.71727240085602
    Elbo loss: -261.5507645010948
    Elbo loss: -218.47509676218033
    Elbo loss: -309.2033520936966
    Elbo loss: -333.6280195713043
    Elbo loss: -370.50675970315933
    Elbo loss: -345.0105627775192
    Elbo loss: -262.72408026456833
    Elbo loss: -373.69888734817505
    Elbo loss: -424.50547766685486
    Elbo loss: -261.58591240644455
    Elbo loss: -425.9056247472763
    Elbo loss: -437.664081633091
    Elbo loss: -429.98125183582306
    Elbo loss: -429.68795144557953
    Elbo loss: -306.91762256622314
    Elbo loss: -373.73281812667847
    Elbo loss: -429.6245778799057
    Elbo loss: -455.4520588517189
    Elbo loss: -431.00230145454407
[]: plt.plot(loss_svi)
```

[]: [<matplotlib.lines.Line2D at 0x2aa626610>]



Compute the predictive distribution sampling from the trained variational density

```
[]: predictive_svi = Predictive(model, guide=guide, num_samples=500)
    preds_svi = predictive_svi(x_test)
    plot_predictions(preds_svi)
```



just reshaping and stuff for the samples of each param and creating a dict, svi_samples

```
[]: | svi_samples = {}
     for k, v in predictive_svi(x_test).items():
         if k != "obs":
             print(k, v.shape)
             svi_samples[k] = v.detach().cpu().numpy()
    fc1.weight torch.Size([500, 1, 20, 1])
    fc1.bias torch.Size([500, 1, 20])
    fc2.weight torch.Size([500, 1, 20, 20])
    fc2.bias torch.Size([500, 1, 20])
    fc3.weight torch.Size([500, 1, 1, 20])
    fc3.bias torch.Size([500, 1, 1])
    sigma torch.Size([500, 1])
    Table of means and standard deivations of weight and bias of each neuron in the 1st hidden layer
[]: layer1 = pd.DataFrame( {'fc1.weight-mean': list(torch.mean(torch.
      →Tensor(svi_samples['fc1.weight']), dim=0)[0].numpy().flatten()),
      'fc1.weight- std': list(torch.std(torch.Tensor(svi_samples['fc1.weight']),
      →dim=0)[0].numpy().flatten()),
       'fc1.bias-mean': list(torch.mean(torch.Tensor(svi_samples['fc1.bias']), ___

dim=0).numpy().flatten()),
        'fc2.bias-std': list(torch.std(torch.Tensor(svi_samples['fc1.bias']),u
      →dim=0)[0].numpy().flatten())}
     )
     layer1
[]:
         fc1.weight-mean fc1.weight- std
                                            fc1.bias-mean
                                                            fc2.bias-std
               -0.829854
                                                 -1.514703
                                                                0.560294
                                  0.799712
     0
     1
               -0.465254
                                  0.458180
                                                 -0.647727
                                                                0.260225
               -0.761623
                                  0.771858
     2
                                                 -1.396604
                                                                0.538531
     3
               -0.714349
                                  0.799186
                                                 -1.596251
                                                                0.622151
     4
               -1.057754
                                  0.095665
                                                  0.064661
                                                                0.006227
     5
               -0.874105
                                  0.745691
                                                 -1.378361
                                                                0.555964
     6
               -0.531903
                                                                0.273270
                                  0.502305
                                                 -0.743302
     7
               -0.298160
                                  0.429458
                                                 -0.703995
                                                                0.297263
     8
               -1.391562
                                  0.054142
                                                  0.176782
                                                                0.003665
     9
               -0.707482
                                  0.804399
                                                 -1.383672
                                                                0.518293
     10
               -0.756813
                                  0.786746
                                                 -1.583973
                                                                0.622611
     11
               -0.751217
                                  0.868038
                                                 -1.472057
                                                                0.612134
     12
               -0.883590
                                                 -1.589278
                                                                0.626012
                                  0.854149
     13
```

-0.158171

-1.195304

-0.562530

0.094532

0.460413

0.223625

0.279741

0.759602

0.391038

-0.168340

-0.837380

-0.408031

14

15

16	-0.910485	0.788020	-1.371178	0.512947
17	-0.908423	0.797930	-1.503552	0.582159
18	-0.915543	0.819935	-1.341792	0.540042
19	0.611785	0.028048	0.062734	0.004860

Means of bias of each neuron in the 2nd hidden layer

```
[]: torch.mean(torch.Tensor(svi_samples['fc2.bias']), dim=0)
```

```
[]: tensor([[-1.6665, 1.0370, -1.7540, -1.5506, -1.7267, -1.6852, -1.6210, -1.7358, -0.7219, 0.2379, -1.6762, -1.5124, -1.7669, -1.6316, -1.2646, -1.3294, -1.6753, -1.8902, 0.4069, -1.6720]])
```

Std deviations of weights of each neuron in the 2nd hidden layer

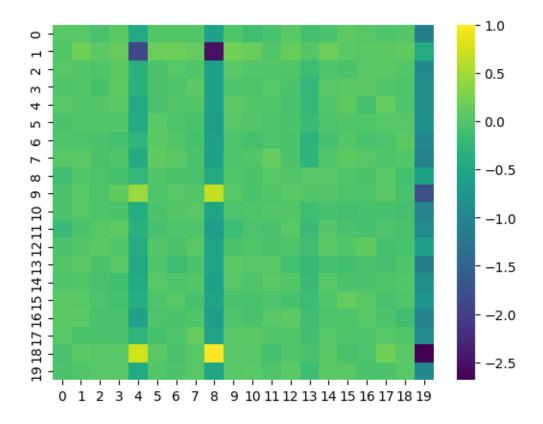
```
[]: torch.std(torch.Tensor(svi_samples['fc2.bias']), dim=0)
```

```
[]: tensor([[0.6554, 0.0088, 0.7184, 0.6475, 0.6745, 0.6508, 0.6938, 0.7059, 0.3457, 0.0251, 0.7133, 0.6559, 0.6591, 0.6654, 0.5006, 0.5298, 0.6608, 0.6994, 0.0123, 0.6709]])
```

Means of weights of each neuron in the 2nd hidden layer

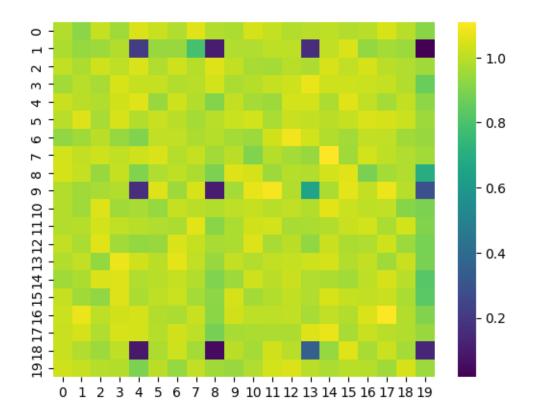
```
[]: m = torch.mean(torch.Tensor(svi_samples['fc2.weight']), dim=0)
m_np = m.numpy().flatten().reshape(20,20)
sns.heatmap(m_np, cmap='viridis')

plt.show()
```



Std deviations of weights of each neuron in the 2nd hidden layer

```
[]: std_devs = torch.std(torch.Tensor(svi_samples['fc2.weight']), dim=0)
std_devs_np = std_devs.numpy().flatten().reshape(20,20)
sns.heatmap(std_devs_np, cmap='viridis')
plt.show()
```



2.2 2. Training BNN (HMC)

```
[]: # Set Pyro random seed
     pyro.set_rng_seed(42)
     # Define Hamiltonian Monte Carlo (HMC) kernel
     hmc_kernel = HMC(model,
                     # num_steps=1000,
                      step_size=0.0001,
     trajectory_length=5,
     # adapt_step_size=True,
     # adapt_mass_matrix=True,
       full_mass=False,
        jit_compile=True)
     # Define MCMC sampler, get 'num_samples' posterior samples with burin in_
     →'warmup_steps'
     mcmc = MCMC(hmc_kernel, num_samples=50, warmup_steps=50)
     # Run MCMC sampler
     mcmc.run(x_train, y_train)
```

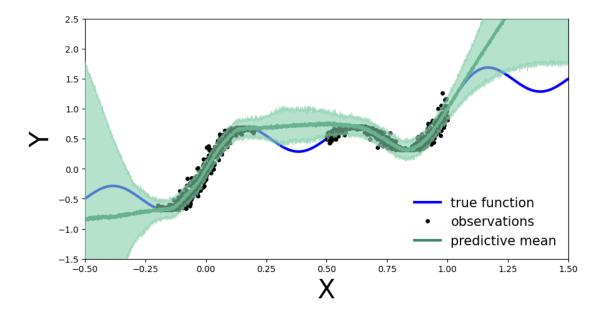
```
hmc_samples = {k: v.detach().cpu().numpy() for k, v in mcmc.get_samples().

items()}
```

Warmup: 0% | 0/100 [00:00,

?it/s]/Users/purnavindhyakota/miniconda3/envs/bnn_trials/lib/python3.11/site-packages/pyro/poutine/subsample_messenger.py:63: TracerWarning: torch.tensor results are registered as constants in the trace. You can safely ignore this warning if you use this function to create tensors out of constant variables that would be the same every time you call this function. In any other case, this might cause the trace to be incorrect.

[]: predictive_hmc = Predictive(model=model, posterior_samples=mcmc.get_samples())
 preds_hmc = predictive_hmc(x_test.flatten())
 plot_predictions(preds_hmc)



2.3 3. Training BNN (NUTS)

```
[]: pyro.set_rng_seed(42)

# Define Hamiltonian Monte Carlo (NUTS) kernel
nuts_kernel = NUTS(model, jit_compile=False)

# Define MCMC sampler, get 'num_samples' posterior samples with burin in_
    'warmup_steps'
```

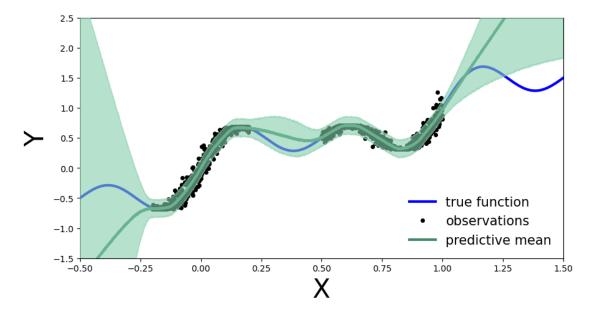
```
mcmc_nuts = MCMC(nuts_kernel, num_samples=500, warmup_steps=50)

# Run MCMC sampler
mcmc_nuts.run(x_train, y_train)

nuts_samples = {k: v.detach().cpu().numpy() for k, v in mcmc_nuts.get_samples().
items()}
```

```
Sample: 100% | 550/550 [08:44, 1.05it/s, step size=9.16e-04, acc. prob=0.913]
```

Calculating and plotting the predictive distribution



Iterate over each site in the hmc_samples dictionary and print the site name and its corresponding values.

```
[]: # for site, values in summary(nuts_samples).items():
# print("Site: {}".format(site))
# print(values, "\n")
```

2.3.1 Comparing Posterior Distributions

2.3.2 plot of posteriors of samples of SVI and NUTS of weights of layer 1

```
[]: sites = list(svi_samples.keys())
     num_neurons = svi_samples[sites[0]].reshape(500,-1).shape[1]
     fig, axs = plt.subplots(nrows=num_neurons, figsize=(12, 10*3))
     fig.suptitle(f"Marginal Posterior density - {sites[0]}", fontsize=16)
         # If there's only one neuron, make axs a list
     if num neurons == 1:
         axs = [axs]
     for i in range(num neurons):
         neuron_weights_svi = svi_samples[sites[0]].reshape(500,-1)[:, i]
         neuron_weights_nuts = nuts_samples[sites[0]].reshape(500,-1)[:, i]
         sns.distplot(neuron_weights_svi, ax=axs[i], label="SVI (DiagNormal)")
         sns.distplot(neuron_weights_nuts, ax=axs[i], label="NUTS")
         axs[i].set_title(f"Neuron {i+1}")
     handles, labels = axs[0].get_legend_handles_labels()
     fig.legend(handles, labels, loc='upper right')
    /var/folders/f6/9mr1g0xj6mqf2j18_cvr2lwc0000gn/T/ipykernel_11651/4163580306.py:1
    4: UserWarning:
    `distplot` is a deprecated function and will be removed in seaborn v0.14.0.
    Please adapt your code to use either `displot` (a figure-level function with
    similar flexibility) or `histplot` (an axes-level function for histograms).
    For a guide to updating your code to use the new functions, please see
    https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
      sns.distplot(neuron weights svi, ax=axs[i], label="SVI (DiagNormal)")
    /var/folders/f6/9mr1g0xj6mqf2j18_cvr2lwc0000gn/T/ipykernel_11651/4163580306.py:1
    5: UserWarning:
    `distplot` is a deprecated function and will be removed in seaborn v0.14.0.
    Please adapt your code to use either `displot` (a figure-level function with
    similar flexibility) or `histplot` (an axes-level function for histograms).
    For a guide to updating your code to use the new functions, please see
    https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
      sns.distplot(neuron_weights_nuts, ax=axs[i], label="NUTS")
```

/var/folders/f6/9mr1g0xj6mqf2j18_cvr2lwc0000gn/T/ipykernel_11651/4163580306.py:1

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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sns.distplot(neuron_weights_svi, ax=axs[i], label="SVI (DiagNormal)")
/var/folders/f6/9mr1g0xj6mqf2j18_cvr2lwc0000gn/T/ipykernel_11651/4163580306.py:1
5: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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sns.distplot(neuron_weights_nuts, ax=axs[i], label="NUTS")
/var/folders/f6/9mr1g0xj6mqf2j18_cvr2lwc0000gn/T/ipykernel_11651/4163580306.py:1
4: UserWarning:

'distplot' is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(neuron_weights_svi, ax=axs[i], label="SVI (DiagNormal)")
/var/folders/f6/9mr1g0xj6mqf2j18_cvr2lwc0000gn/T/ipykernel_11651/4163580306.py:1
5: UserWarning:

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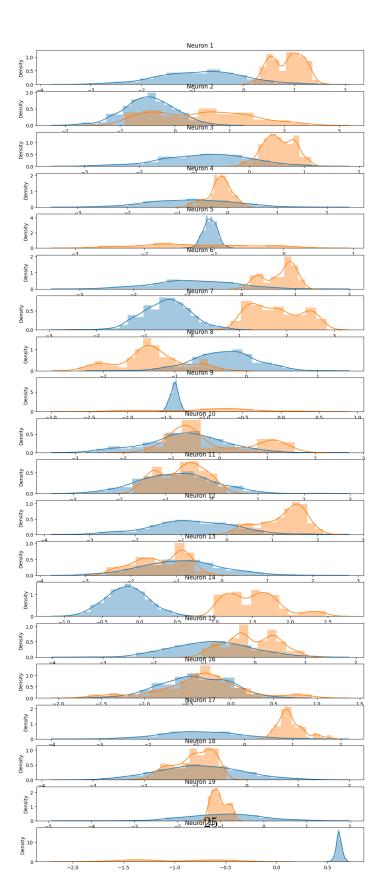
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sns.distplot(neuron_weights_nuts, ax=axs[i], label="NUTS")

[]: <matplotlib.legend.Legend at 0x2aa7dcb10>



3 4. Deterministic NN (Point Estimates)

```
[]: class MLP(nn.Module):
        def __init__(self, input_dim=1, output_dim=1, hidden_dim=10,__
      →n_hidden_layers=2, use_dropout=False):
            super().__init__()
            self.use_dropout = use_dropout
            if use_dropout:
                 self.dropout = nn.Dropout(p=0.5)
            self.activation = nn.Tanh()
             # dynamically define architecture
            self.layer_sizes = [input_dim] + n_hidden_layers * [hidden_dim] +__
      layer_list = [nn.Linear(self.layer_sizes[idx - 1], self.
      ⇒layer_sizes[idx]) for idx in
                           range(1, len(self.layer_sizes))]
            self.layers = nn.ModuleList(layer_list)
        def forward(self, input):
            hidden = self.activation(self.layers[0](input))
            for layer in self.layers[1:-1]:
                 hidden_temp = self.activation(layer(hidden))
                 if self.use_dropout:
                    hidden_temp = self.dropout(hidden_temp)
            output_mean = self.layers[-1](hidden).squeeze()
            return output_mean
```

3.0.1 Training one NN

```
def get_simple_data_train():
    x = np.linspace(-.2, 0.2, 500)
    x = np.hstack([x, np.linspace(.6, 1, 500)])
    eps = 0.02 * np.random.randn(x.shape[0])
    y = x + 0.3 * np.sin(2 * np.pi * (x + eps)) + 0.3 * np.sin(4 * np.pi * (x + eps)) + eps
    x_train = torch.from_numpy(x).float()[:, None]
    y_train = torch.from_numpy(y).float()
    return x_train, y_train

def train(net, train_data):
```

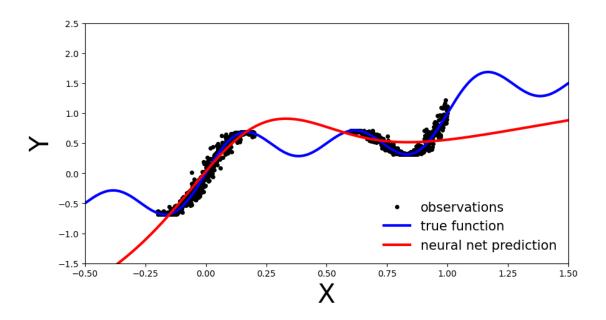
```
x_train, y_train = train_data
optimizer = torch.optim.Adam(params=net.parameters(), lr=1e-3)
criterion = nn.MSELoss()

progress_bar = trange(3000)
for _ in progress_bar:
    optimizer.zero_grad()
    loss = criterion(y_train, net(x_train))
    progress_bar.set_postfix(loss=f'{loss / x_train.shape[0]:.3f}')
    loss.backward()
    optimizer.step()
return net
```

Retaining the same architecture as that of the BNN, we use a FCN with 2 hidden layers, each with 20 neurons

```
[]: train_data = get_simple_data_train()
     x_test = torch.linspace(-.5, 1.5, 3000)[:, None] # test over the whole range
     net_MLP = MLP(hidden_dim=20, n_hidden_layers=2)
     net_MLP_train = train(net_MLP, train_data)
     y_preds_single_NN = net_MLP_train(x_test).clone().detach().numpy()
     fig, ax = plt.subplots(figsize=(10, 5))
     plt.xlim([-.5, 1.5])
     plt.ylim([-1.5, 2.5])
     plt.xlabel("X", fontsize=30)
     plt.ylabel("Y", fontsize=30)
     x_train, y_train = get_simple_data_train()
     x_{true} = np.linspace(-.5, 1.5, 1000)
     y_{true} = x_{true} + 0.3 * np.sin(2 * np.pi * x_{true}) + 0.3 * np.sin(4 * np.pi *_{l}
      ⊶x_true)
     ax.plot(x_train, y_train, 'ko', markersize=4, label="observations")
     ax.plot(x true, y true, 'b-', linewidth=3, label="true function")
     ax.plot(x_test, y_preds_single_NN, 'r-', linewidth=3, label='neural net_
      ⇔prediction')
     plt.legend(loc=4, fontsize=15, frameon=False)
     plt.show()
```

0%| | 0/3000 [00:00<?, ?it/s]

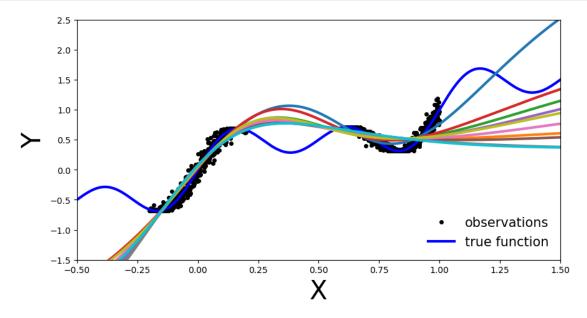


3.1 5. Deep Ensemble

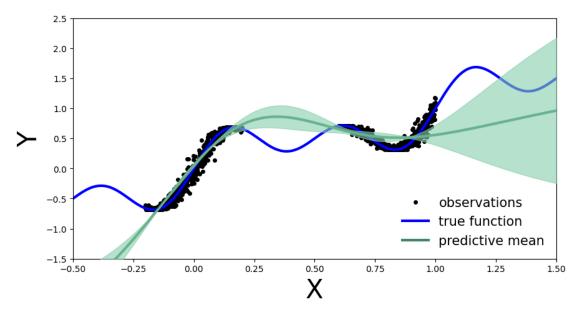
```
[]: ensemble_size = 10
     ensemble = [MLP(hidden_dim=20, n_hidden_layers=2) for _ in range(ensemble_size)]
     for net in ensemble:
         train(net, train_data)
      0%|
                    | 0/3000 [00:00<?, ?it/s]
      0%|
                    | 0/3000 [00:00<?, ?it/s]
      0%1
                    | 0/3000 [00:00<?, ?it/s]
      0%1
                   | 0/3000 [00:00<?, ?it/s]
                   | 0/3000 [00:00<?, ?it/s]
      0%1
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      0%|
                    | 0/3000 [00:00<?, ?it/s]
      0%1
                    | 0/3000 [00:00<?, ?it/s]
[]: y_preds_ensemble = [np.array(net(x_test[:, None]).clone().detach().numpy()) for__
      onet in ensemble]
```

Plot each ensemble member's predictive function.

```
[ ]: def plot_generic(add_to_plot=None):
         fig, ax = plt.subplots(figsize=(10, 5))
         plt.xlim([-.5, 1.5])
         plt.ylim([-1.5, 2.5])
         plt.xlabel("X", fontsize=30)
         plt.ylabel("Y", fontsize=30)
         x_train, y_train = get_simple_data_train()
         x_{true} = np.linspace(-.5, 1.5, 1000)
         y_{true} = x_{true} + 0.3 * np.sin(2 * np.pi * x_{true}) + 0.3 * np.sin(4 * np.pi_
      →* x_true)
         ax.plot(x_train, y_train, 'ko', markersize=4, label="observations")
         ax.plot(x_true, y_true, 'b-', linewidth=3, label="true function")
         if add_to_plot is not None:
             add_to_plot(ax)
         plt.legend(loc=4, fontsize=15, frameon=False)
         plt.show()
     def plot_multiple_predictions(x_test, y_preds):
         def add_multiple_predictions(ax):
             for idx in range(len(y_preds)):
                 ax.plot(x_test, y_preds[idx], '-', linewidth=3)
         plot_generic(add_multiple_predictions)
     plot_multiple_predictions(x_test, y_preds_ensemble)
```



Visualizing the uncertainty bands



strengths and weaknesses (?) of each approach compare with gpr (?) prediction uncertainty (interpolation and extrapolation) predicting outside the range of data magnitude of noise and its effect on predictive capability distribution of the weights (?)/ posterior joint distribution of posterior of weights: hmc vs vi approache maybe (?) some weights might be deterministic (?), the effect of removing/including these put in the basics of the basics of the formulations

3.2 6. Gaussian Process Regression

GPR with RBF kernel (variance = 0.02, length scale = 0.05) - starting parameters

```
[]: kernel = GPy.kern.RBF(input_dim=1, variance=0.02, lengthscale=0.05)
    m = GPy.models.GPRegression(x_obs.reshape(-1,1),y_obs.reshape(-1,1),kernel)
    print(m)
    m.plot()
    plt.title('GPR on train data')
    plt.xlabel('X')
    plt.ylabel('Y')
    plt.legend(['obrsevations', 'GPR mean', 'GPR std'])
```

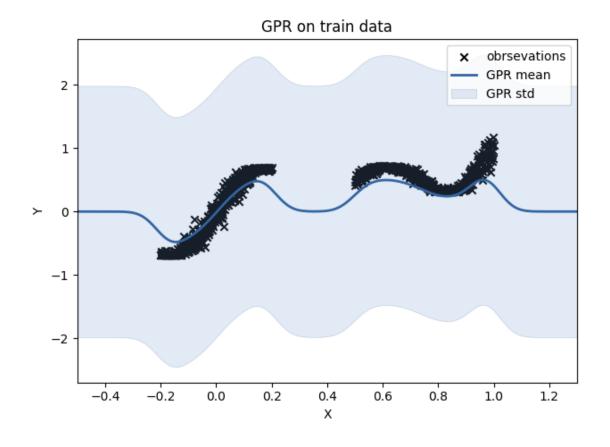
Name : GP regression

Objective: 976.2354051648001 Number of Parameters: 3

Number of Optimization Parameters : 3

Updates : True Parameters:

[]: <matplotlib.legend.Legend at 0x2dd00be50>

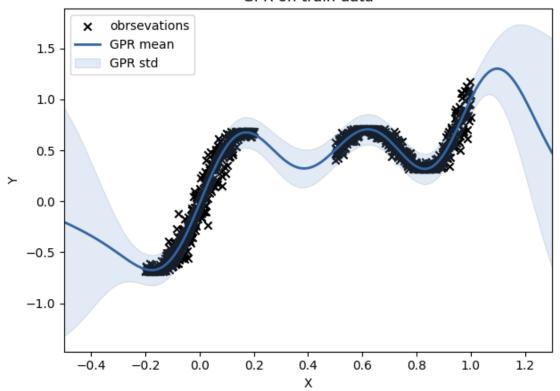


Optimizing GPR hyperparameters (variance and lengthscale)

```
[]: m.unconstrain('rbf.variance')
    m.optimize()
    print(m)
    m.plot()
    plt.title('GPR on train data')
    plt.xlabel('X')
    plt.ylabel('Y')
    plt.legend(['obrsevations', 'GPR mean', 'GPR std'])
    Name : GP regression
    Objective: -1125.913860524442
    Number of Parameters : 3
    Number of Optimization Parameters : 3
    Updates : True
    Parameters:
      GP_regression.
                              1
                                                value | constraints |
    priors
      rbf.variance
                                 0.48091788059212526
                                                              +ve
      rbf.lengthscale
                                   0.1877904485630873
                                                              +ve
      Gaussian_noise.variance | 0.005733342946187983 |
                                                              +ve
```

[]: <matplotlib.legend.Legend at 0x2dd1d6690>

GPR on train data



Plot of predictions on test data $(x_{test} = [-0.5, 1.5])$

