# set max iteration to 3500 opt params['max iteration'] = real max iteration # initialize train and validation data x\_train, y\_train, x\_test = generate\_data(training\_seed) x\_valid, y\_valid, \_ = generate\_data(valid\_seed) In [3]: #JACK # Edit grad\_func\_specs dictionary for fixed step sizes grad\_func\_specs\_80 = {'random' : 0.8} grad\_func\_specs\_60 = {'random' : 0.6} grad\_func\_specs\_40 = {'random' : 0.4} grad\_func\_specs\_20 = {'random' : 0.2} grad\_func\_specs\_10 = {'random' : 0.1} grad\_func\_specs\_1 = {'random' : 0.01} In [4]: opt\_params Out[4]: {'step\_size': 0.001, 'max\_iteration': 3500, 'random\_restarts': 1, 'optimizer': 'adam'} **Train LUNA** luna\_80percent uses 80 % of indices for sampling luna\_60percent uses 60 % of indices for sampling luna 40percent uses 40 % of indices for sampling luna\_20percent uses 20 % of indices for sampling luna\_10percent uses 10 % of indices for sampling luna\_1percent uses 1 % of indices for sampling LUNA 80% of indices sampled t0 = time.time()luna\_80percent = LUNA(prior\_variance, y\_noise\_variance, regularization\_param\_luna, similarity\_param, luna\_ar random\_seed, grad\_func\_specs= grad\_func\_specs\_80) luna 80percent.train(x train, y train, opt params) times.append(np.round(time.time() - t0, 3)) print(f"time: {np.round(time.time() - t0, 3)} seconds") Iteration 3400 lower bound -5588.709062666859; gradient mag: 461.118028770167726 Done Training time: 2921.86 seconds LUNA 60% of indices sampled t0 = time.time()luna\_60percent = LUNA(prior\_variance, y\_noise\_variance, regularization\_param\_luna, similarity param, luna ar random\_seed, grad\_func\_specs= grad\_func\_specs\_60) luna\_60percent.train(x\_train, y\_train, opt\_params) times.append(np.round(time.time() - t0, 3)) print(f"time: {np.round(time.time() - t0, 3)} seconds") Iteration 3400 lower bound -6443.384959618316; gradient mag: 123.269959547352236 Done Training time: 2862.534 seconds LUNA 40% of indices sampled t0 = time.time()luna\_40percent = LUNA(prior\_variance, y\_noise\_variance, regularization\_param\_luna, similarity\_param, luna\_ar random\_seed, grad\_func\_specs= grad\_func\_specs\_40) luna\_40percent.train(x\_train, y\_train, opt\_params) times.append(np.round(time.time() - t0, 3)) print(f"time: {np.round(time.time() - t0, 3)} seconds") Iteration 3400 lower bound -4220.278562518963; gradient mag: 270.960026034326823 Done Training time: 2881.376 seconds **LUNA 20% of indices sampled** t0 = time.time()luna\_20percent = LUNA(prior\_variance, y\_noise\_variance, regularization\_param\_luna, similarity\_param, luna ar random\_seed, grad\_func\_specs= grad\_func\_specs\_20) luna\_20percent.train(x\_train, y\_train, opt\_params) times.append(np.round(time.time() - t0, 3)) print(f"time: {np.round(time.time() - t0, 3)} seconds") Iteration 3400 lower bound -5610.3286802644325; gradient mag: 303.25387124922406 Done Training time: 2882.896 seconds #### LUNA 10% of indices sampled t0 = time.time()luna\_10percent = LUNA(prior\_variance, y\_noise\_variance, regularization\_param\_luna, similarity\_param, luna ar random\_seed, grad\_func\_specs= grad\_func\_specs\_10) luna 10percent.train(x\_train, y\_train, opt\_params) times.append(np.round(time.time() - t0, 3)) print(f"time: {np.round(time.time() - t0, 3)} seconds") Iteration 3400 lower bound -5300.80615910636; gradient mag: 507.2983169780017047 Done Training time: 2951.541 seconds #### LUNA 1% of indices sampled t0 = time.time()luna\_1percent = LUNA(prior\_variance, y\_noise\_variance, regularization\_param\_luna, similarity\_param, luna\_arc random seed, grad func specs grad func specs 1) luna\_1percent.train(x\_train, y\_train, opt\_params) times.append(np.round(time.time() - t0, 3)) print(f"time: {np.round(time.time() - t0, 3)} seconds") Iteration 3400 lower bound -5067.575411729897; gradient mag: 103.299955190134345 Done Training time: 3009.353 seconds LL rand index = [] for model in [luna\_1percent, luna\_10percent, luna\_20percent, luna\_40percent, luna\_60percent, luna\_80percent] LL rand index.append(model.get log l(x train, y train, x valid, y valid)) In [14]: print('log-likelihood for Sampling Analysis:\n') print(f'log-likelihood for 1% random indices: {LL\_rand\_index[0]}')
print(f'log-likelihood for 10% random indices: {LL\_rand\_index[1]}')
print(f'log-likelihood for 20% random indices: {LL\_rand\_index[2]}')
print(f'log-likelihood for 40% random indices: {LL\_rand\_index[3]}')
print(f'log-likelihood for 60% random indices: {LL\_rand\_index[4]}')
print(f'log-likelihood for 80% random indices: {LL\_rand\_index[4]}') log-likelihood for Sampling Analysis: log-likelihood for 1% random indices: -143.0396802074957 -135.77955710085863 log-likelihood for 10% random indices: log-likelihood for 20% random indices: -132.5767747361672 log-likelihood for 40% random indices: -131.4964262952128 log-likelihood for 60% random indices: -134.1246586098263 log-likelihood for 80% random indices: -133.96293658992488 Compare Results Not sure what to conclude. In [15]: # xx == percentage of indices sampled xx = [1, 10, 20, 40, 60, 80]fig, ax = plt.subplots(1,1,figsize = (10,8))#ax = axes.flatten()ax.plot(xx, LL rand index, 'black', lw='3', label='log-likelihood given indices sampled') ax.set xlabel('Percentage of Indices Sampled', fontsize = 14) ax.set ylabel('log-likelihood', fontsize = 14) ax.set title('log-likelihood vs. Percentage of Indices Sampled', fontsize = 18) #ax.legend(fontsize=12); fig.savefig(f"figs/LUNA log likelihood sampling analysis {str(opt params['max iteration'])} iterations.png") log-likelihood vs. Percentage of Indices Sampled -132-134-136log-likelihood -138 -140-142Percentage of Indices Sampled fig,ax = plt.subplots(2,3,figsize=(20,15))j = 0k = 0for model in [luna\_1percent, luna\_10percent, luna\_20percent, luna\_40percent, luna\_60percent, luna\_80percent] predictives, predictive samples = model.predict(x test) log\_l = model.get\_log\_l(x\_train, y\_train, x\_valid, y\_valid) name = f"\nLuna: sample % = {percs[i]\*100}%" + f"\n Log Likelihood: {round(log 1,2)}" + f"\n Time: {roun ax[j,k]=bh.viz\_pp\_samples(x\_train, y\_train,x\_test.flatten(),predictive\_samples,name, multi = ax[j,k]) # ensure graphs are plotted in a logical order i +=1 k **+=** 1 **if** i == 3: k = 0j = 1plt.legend() fig.savefig(f"figs/LUNA\_sampling\_analysis\_{str(opt\_params['max\_iteration'])}\_iterations.png") Luna: sample % = 1.0% Luna: sample % = 10.0% Luna: sample % = 20.0% Log Likelihood: -132.58 Log Likelihood: -143.04 Time: 0.81 Hours Log Likelihood: -135.78 Time: 0.8 Hours Time: 0.8 Hours 150 150 150 100 100 100 50 50 50 0 0 -50 -50 -50-100 -100 -100 -150-150-150Luna: sample % = 40.0% Luna: sample % = 60.0% Luna: sample % = 80.0% Log Likelihood: -131.5 Log Likelihood: -134.12 Log Likelihood: -133.96 Time: 0.8 Hours Time: 0.82 Hours Time: 0.84 Hours mean prediction true function 95% Conf. Interva training data 100 100 100 50 50 50 0 -50-50-50-100-100-100-150-150-150opt\_params {'step\_size': 0.001, 'max iteration': 3500, 'random restarts': 1, 'optimizer': 'adam'} Out[18]: [2921.86, 2862.534, 2881.376, 2882.896, 2951.541, 3009.352]

In [1]: # import standard libraries

import pandas as pd

# import our libraries
import bayes helpers as bh

from nlm import NLM
from luna import LUNA
from config import \*

import numpy

import sys
import time

from autograd import numpy as np

from autograd import scipy as sp
import autograd.numpy.random as npr

import matplotlib.pyplot as plt

from autograd.misc.optimizers import adam, sgd

from utils import generate data, run toy nn

Generates 100 datapoints for train and 100 points for test according to the function

 $y = \frac{1}{2}x^3 + \epsilon$ 

 $\epsilon \sim N(0,3^2)$ 

from feed forward import Feedforward

Define LUNA Hyperparameters

# list of index percentages to test:
percs = [0.01, 0.1, 0.2, 0.4, 0.6, 0.8]

**Generate Cubic Dataset** 

In [2]: # list to track runtimes
times = []

from autograd import grad