

```
In [ ]: import numpy as np
import torch as torch
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
import seaborn as sns
import pandas as pd
import matplotlib.pyplot as plt
from datetime import datetime
import random
from scipy.stats import norm

import os
import pathlib

from Model import *
from utils import *

torch.autograd.set_detect_anomaly(True)
start_time=datetime.now().strftime('%B %d - %H:%M:%S')
```

```
In [ ]: #Global parameters
GlobalParams1=Params(param_type='k1',target_type='indicator',trick='clamp',loss_type='MSELoss',delta=0.01,w=0.5,lr=0.0005)
dB1 = SampleBMIncr(GlobalParams=GlobalParams1)
init_x1 = Sample_Init(GlobalParams=GlobalParams1)
init_c1= torch.zeros_like(init_x1)

GlobalParams2=Params(param_type='k2',target_type='indicator',trick='clamp',loss_type='MSELoss',delta=0.01,w=0.5,lr=0.0005)
dB2 = SampleBMIncr(GlobalParams=GlobalParams2)    ## TODO: same dB?????
init_x2 = Sample_Init(GlobalParams=GlobalParams2)
init_c2= torch.zeros_like(init_x2)

NT1=GlobalParams1.NT1
NT2=GlobalParams1.NT2
dt=GlobalParams1.dt
device=GlobalParams1.device
learning_rate = GlobalParams1.lr

#Forward Loss
forward_losses = []

#How many batches
MaxBatch= 500

#How many optimization steps per batch
OptimSteps= 25

#Train on a single batch?
single_batch = True
```

```

#Set up main models for y0, yt1, and z (z will be list of models)
y0_model_main1 = Network(scaler_type='sigmoid')
yt1_model_main1 = Network(scaler_type='sigmoid')
zy_models_main1 = [Network() for i in range(NT2)]
main_models1=Main_Models(GlobalParams=GlobalParams1)
main_models1.create(y0_model=y0_model_main1,
                    yt1_model=yt1_model_main1,
                    zy_models=zy_models_main1,
                    forward_loss=forward_losses,
                    dB=dB1,
                    init_x=init_x1,
                    init_c=init_c1)

y0_model_main2 = Network(scaler_type='sigmoid')
yt1_model_main2 = Network(scaler_type='sigmoid')
zy_models_main2 = [Network() for i in range(NT2)]
main_models2=Main_Models(GlobalParams=GlobalParams2)
main_models2.create(y0_model=y0_model_main2,
                    yt1_model=yt1_model_main2,
                    zy_models=zy_models_main2,
                    forward_loss=forward_losses,
                    dB=dB2,
                    init_x=init_x2,
                    init_c=init_c2)

pop1_dict={'dB':dB1,
           'init_x':init_x1 ,
           'init_c':init_c1 ,
           'GlobalParams':GlobalParams1,
           'main_models':main_models1}

pop2_dict={'dB':dB2,
           'init_x':init_x2 ,
           'init_c':init_c2 ,
           'GlobalParams':GlobalParams2,
           'main_models':main_models2}

```

```

In [ ]: #Define optimization parameters
params=[]
params = list(main_models1.y0_model.parameters())+\
         list(main_models1.yt1_model.parameters())+\
         list(main_models2.y0_model.parameters())+\
         list(main_models2.yt1_model.parameters())

for i in range(NT2):
    params += list(main_models1.zy_models[i].parameters())
    params += list(main_models2.zy_models[i].parameters())

#Set up optimizer and scheduler
optimizer = optim.Adamax(params, lr=learning_rate)

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scheduler = torch.optim.lr_scheduler.StepLR(optimizer, step_size=100, gamma=0.95)

for k in range(0,MaxBatch):

    print("Batch Number: ", k+1)
    sloss=0
    #optimize main network wrt the foward loss
    for l in range(0,OptimSteps):
        optimizer.zero_grad()
        loss = get_foward_loss(pop1_dict=pop1_dict, pop2_dict=pop2_dict)
        loss.backward()
        # torch.nn.utils.clip_grad_norm_(parameters=params, max_norm=0.7)
        optimizer.step()
        scheduler.step()
        nloss = loss.detach().numpy()
        sloss += nloss
        # print('OptimStep: '+ str(l+1))
        # print('forward_loss: ' + str(nloss))
    avgloss = sloss/OptimSteps
    print("Average Error Est: ", avgloss)
    forward_losses.append(avgloss)

#Generate a new batch if using multiple batches
if(not single_batch):

    dB1 = SampleBMIncr(GlobalParams=GlobalParams1)
    init_x1 = Sample_Init(GlobalParams=GlobalParams1)
    init_c1= torch.zeros_like(init_x1)
    pop1_dict={'dB':dB1,
               'init_x':init_x1 ,
               'init_c':init_c1 ,
               'GlobalParams':GlobalParams1,
               'main_models':main_models1}

    dB2 = SampleBMIncr(GlobalParams=GlobalParams2)    ## TODO: same dB?????
    init_x2 = Sample_Init(GlobalParams=GlobalParams2)
    init_c2= torch.zeros_like(init_x2)
    pop2_dict={'dB':dB2,
               'init_x':init_x2 ,
               'init_c':init_c2 ,
               'GlobalParams':GlobalParams2,
               'main_models':main_models2}

```

```
KeyError
```

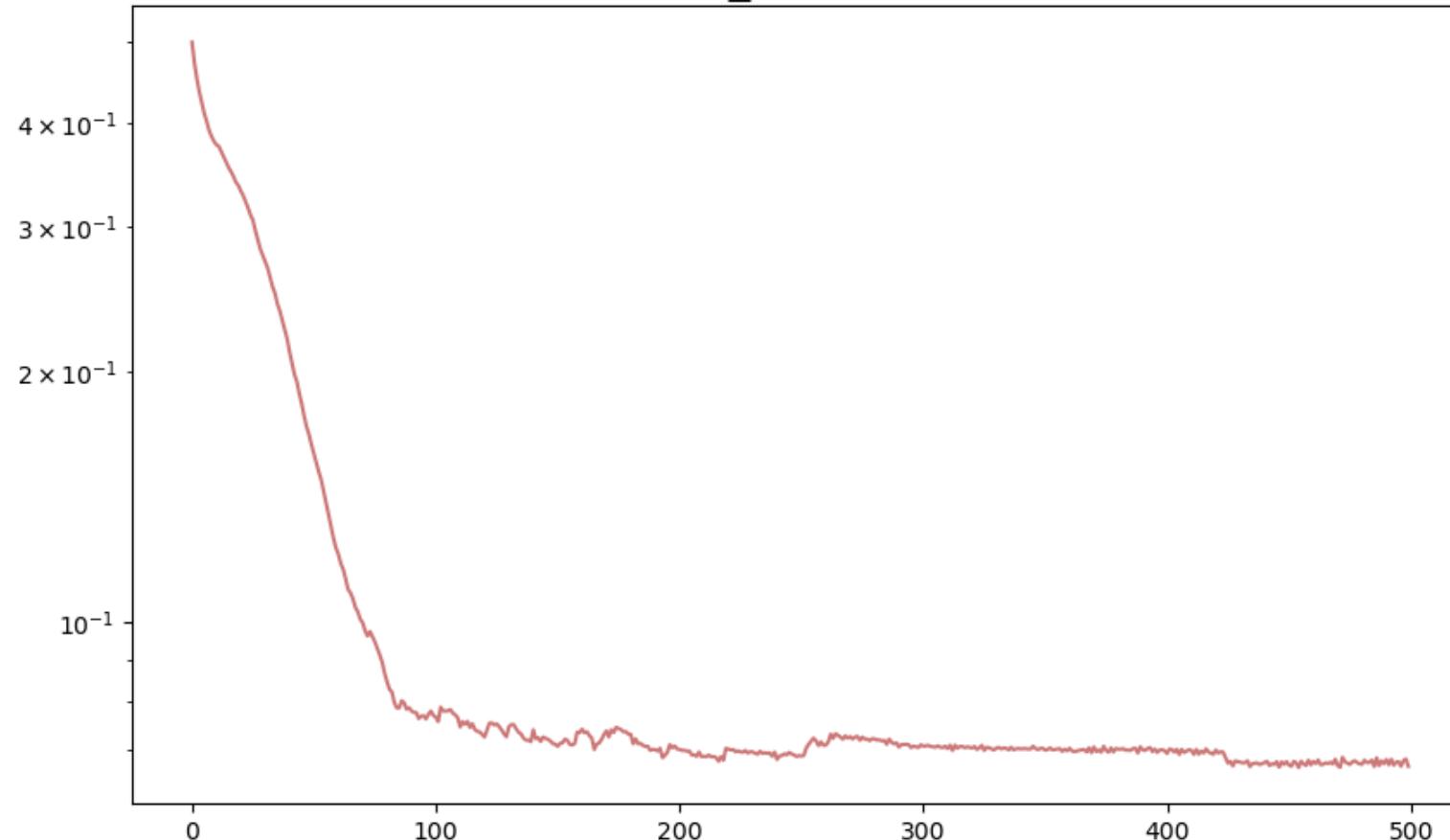
```
Traceback (most recent call last)
```

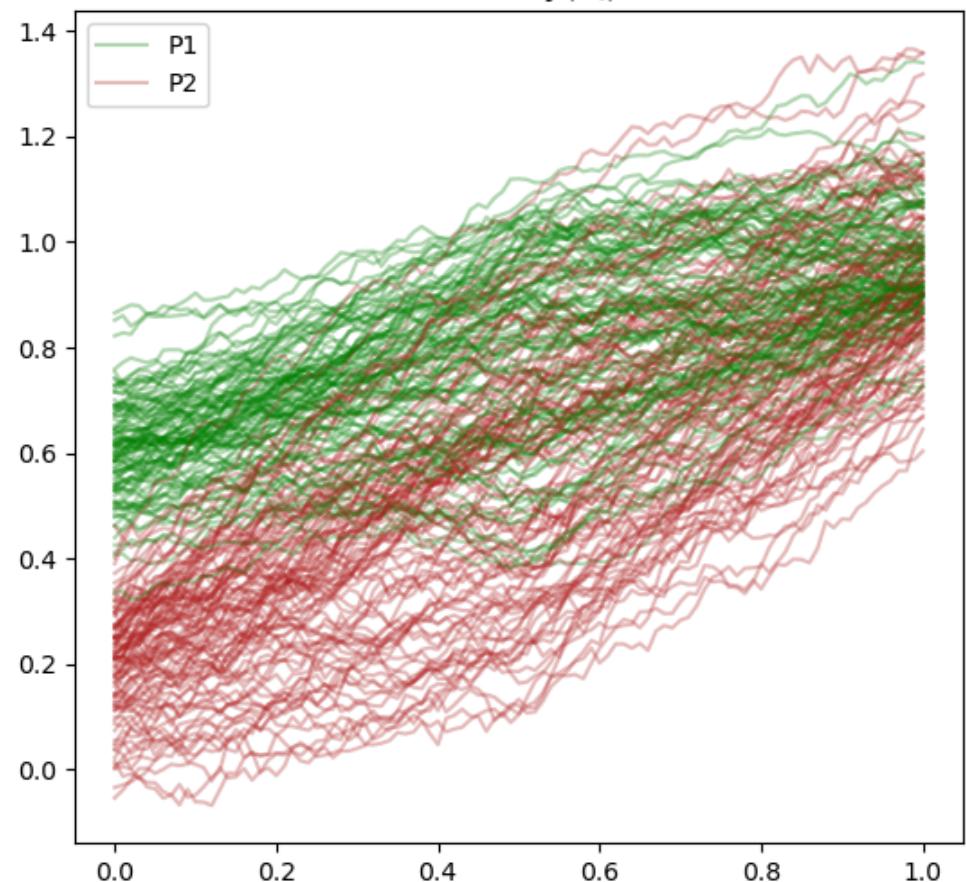
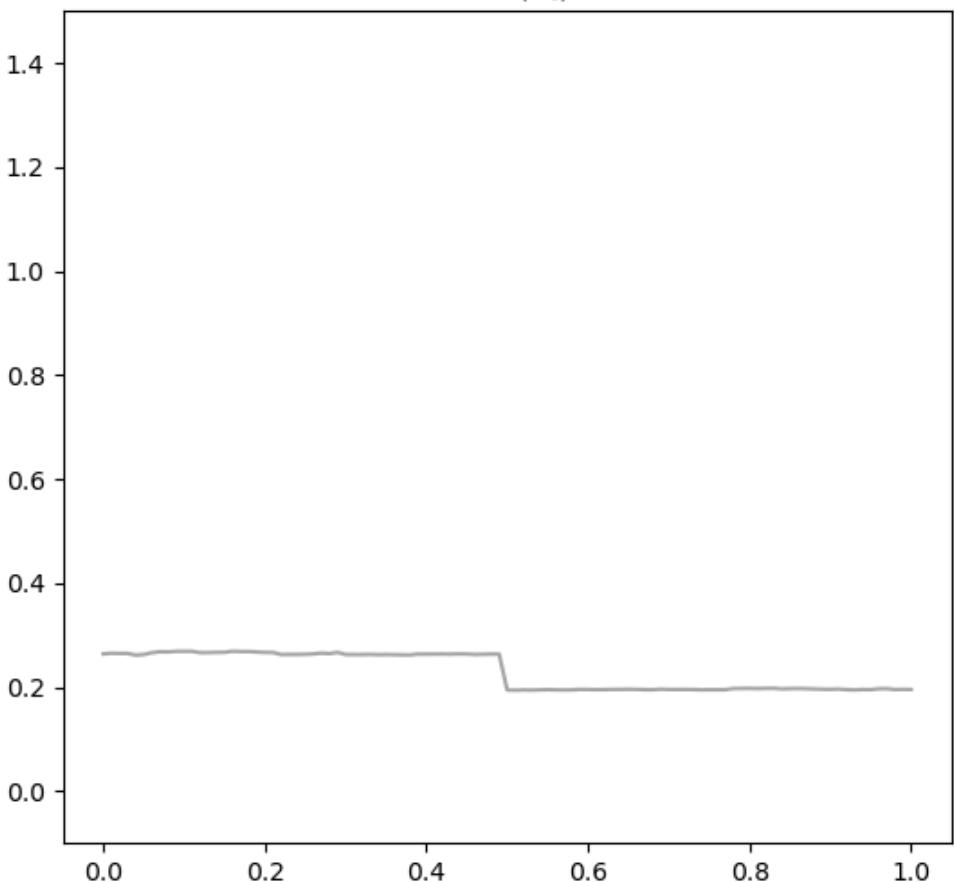
```
Cell In[4], line 5
    3 plot.Inventory_And_Price()
    4 plot.Decomposition_Inventory()
--> 5 plot.Key_Processes()
    6 plot.Terminal_Convergence()
```

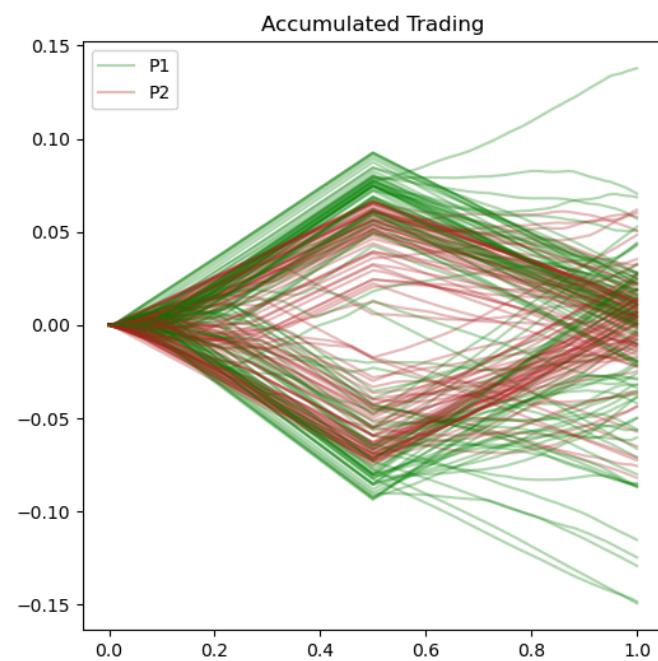
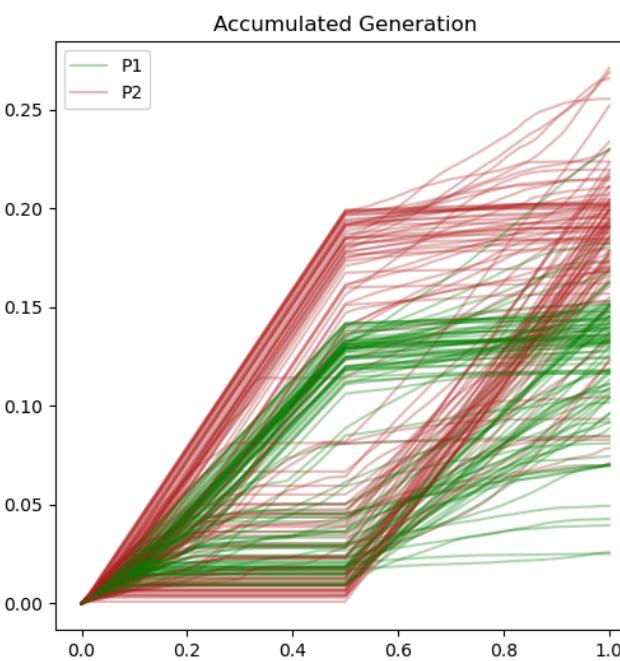
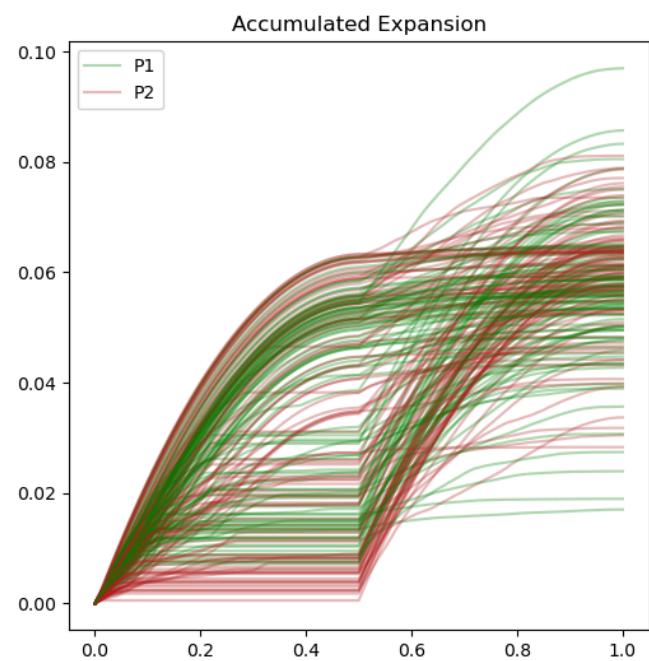
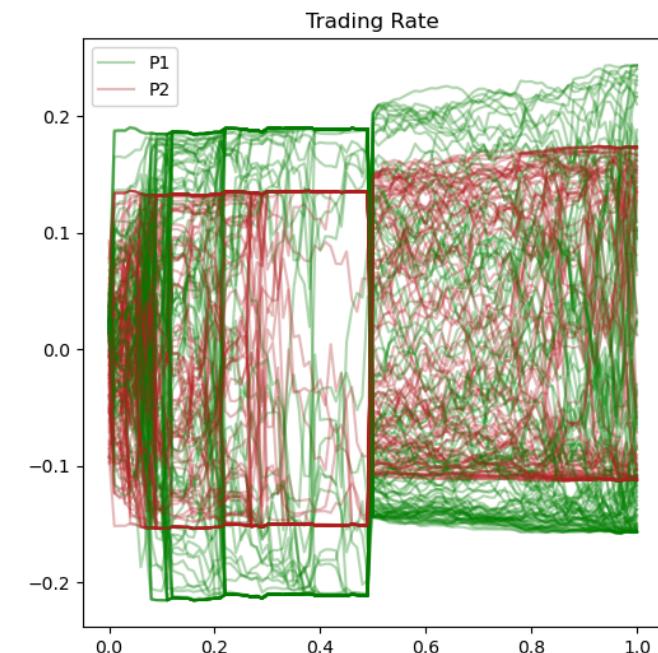
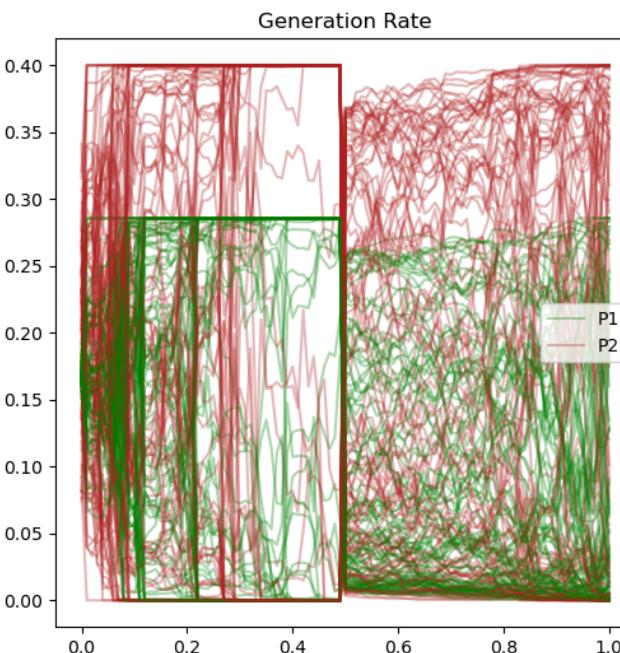
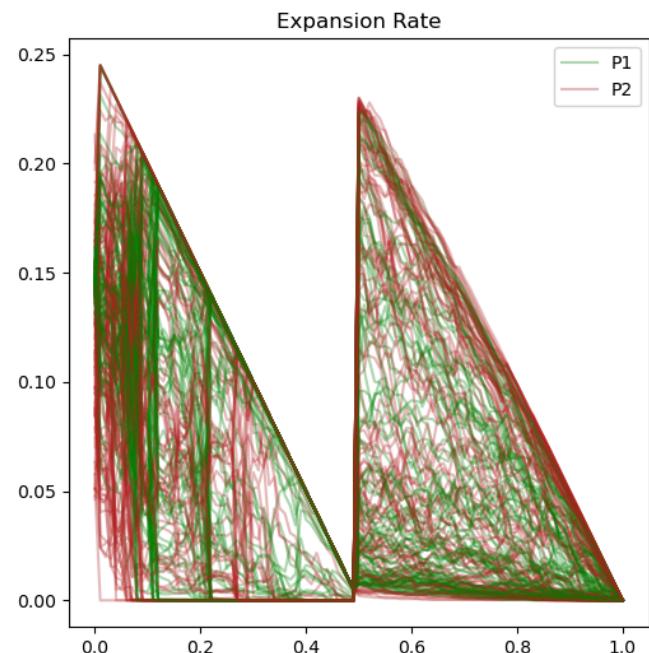
```
File ~/Orange Ao/Research And Projects/2024.5 Columbia_Steven/2Period/RunTwice_1Prd/utils.py:531, in plot_results.Key_Processes(self, V, U, Y)
  528 plt.title("$V_t$")
  529 for i in range(self.number_of_paths):
  530     # print(self.t.shape, self.pop1_path_dict['v'][i].shape)
--> 531     ax1=plt.plot(self.t, self.pop1_path_dict['v'][i], color="green", alpha=0.3)
  532     ax2=plt.plot(self.t, self.pop2_path_dict['v'][i], color="firebrick", alpha=0.3)
  533 plt.legend({'P1':ax1,'P2':ax2})
```

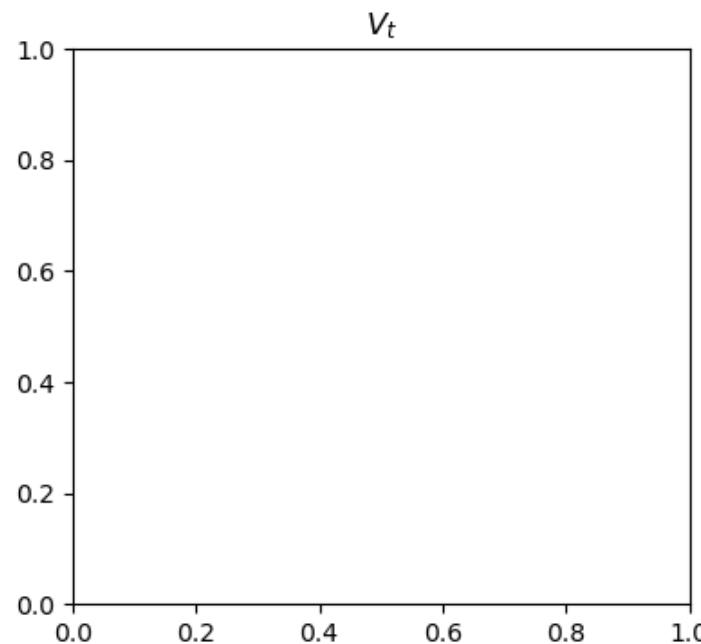
```
KeyError: 'v'
```

Forward\_Loss vs Batch



$Inventory(X_t)$  $Price(S_t)$ 





## Save The Models

```
In [ ]: print(f"{len(main_models1.loss)} steps\nStarted @ {start_time}\nSaved @ {datetime.now().strftime('%B %d - %H:%M:%S')}" ) ## to examine whether
dir_path=pathlib.Path(os.getcwd(),
                      'Results',
                      'Best Models Saved',
                      'sigmoid_ind_0.0005lr_500steps_MSE_0.5w')
dir_path.mkdir()
path1=pathlib.Path(dir_path,'pop1.pt')
path2=pathlib.Path(dir_path,'pop2.pt')
main_models1.save_entire_models(path=path1)
main_models2.save_entire_models(path=path2)
```

500 steps  
 Started @ August 21 – 23:17:08  
 Saved @ August 22 – 08:21:34

## Load The Model

```
In [ ]: # GlobalParams1=Params(param_type='k1',target_type='indicator',trick='clamp',loss_type='MSELoss',delta=0.03,K=0.9,lr=0.01)
# GlobalParams2=Params(param_type='k2',target_type='indicator',trick='clamp',loss_type='MSELoss',delta=0.03,K=0.9,lr=0.01)
models1=Main_Models(GlobalParams=GlobalParams1)
models2=Main_Models(GlobalParams=GlobalParams2)
path_dir=pathlib.Path(os.getcwd(),
                      "Results",
                      "Best Models Saved",
```

```

'sigmoid_ind_0.0005lr_500steps_MSE_0.5w')
path1=pathlib.Path(path_dir,'pop1.pt')
path2=pathlib.Path(path_dir,'pop2.pt')
model_dict1=models1.load_entire_models(path=path1,overwrite=True)
model_dict2=models2.load_entire_models(path=path2,overwrite=True)

dB1=model_dict1['dB']
init_x1=model_dict1['init_x']
init_c1=model_dict1['init_c']
pop1_dict= {'dB':dB1,
            'init_x':init_x1 ,
            'init_c':init_c1 ,
            'GlobalParams':GlobalParams1,
            'main_models':models1}

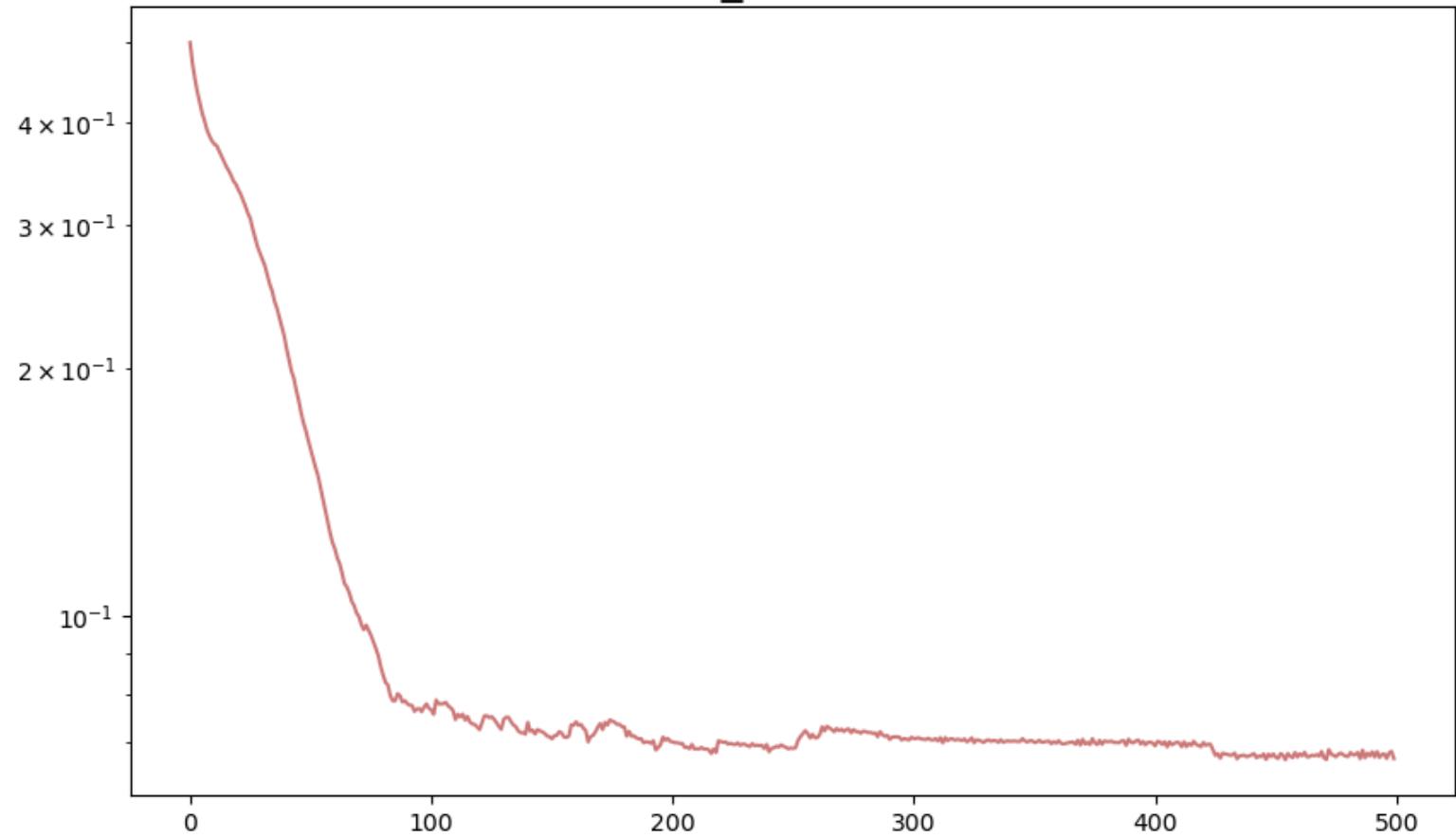
dB2=model_dict2['dB']
init_x2=model_dict2['init_x']
init_c2=model_dict2['init_c']

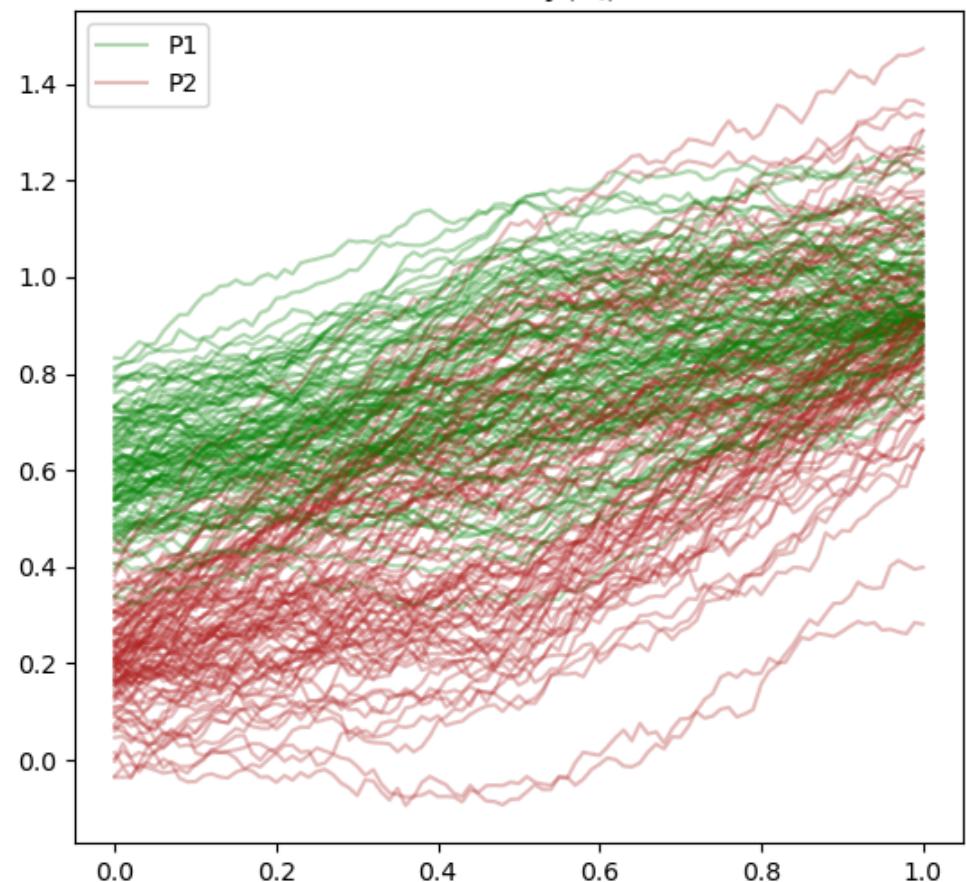
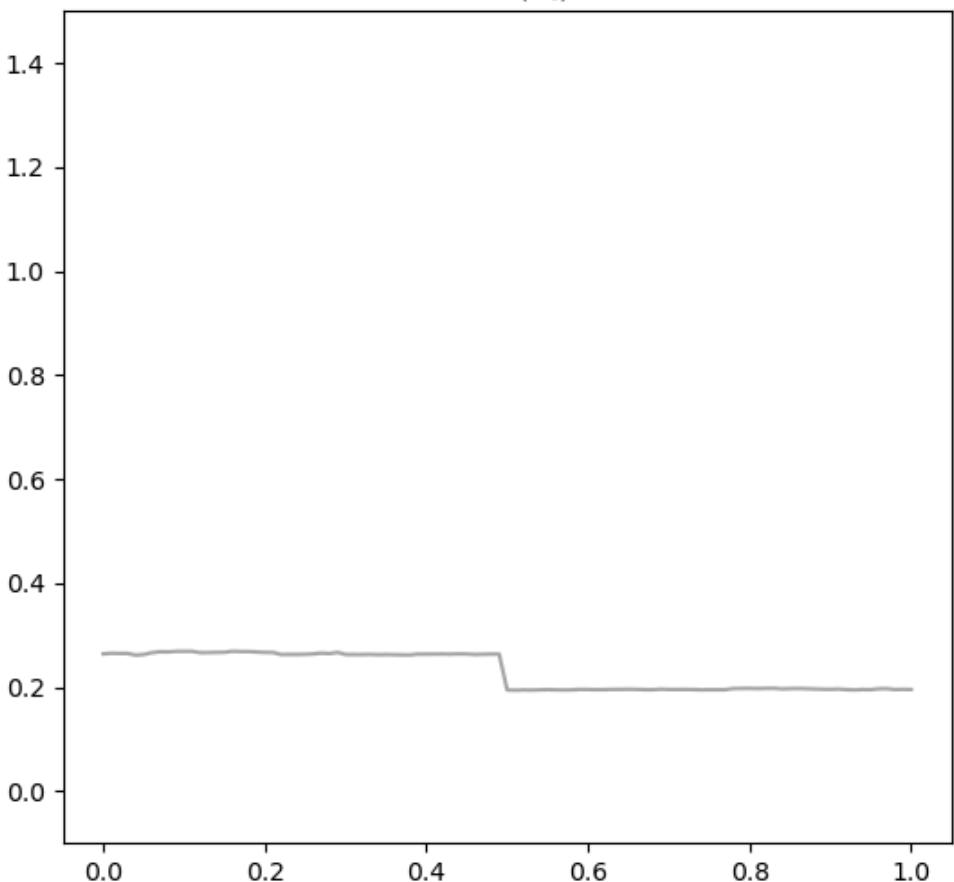
pop2_dict= {'dB':dB2,
            'init_x':init_x2 ,
            'init_c':init_c2 ,
            'GlobalParams':GlobalParams2,
            'main_models':models2}
dt=GlobalParams1.dt
NT1=GlobalParams1.NT1
NT2=GlobalParams1.NT2
NumTrain=GlobalParams1.NumTrain
K=GlobalParams1.K
loss=models1.loss

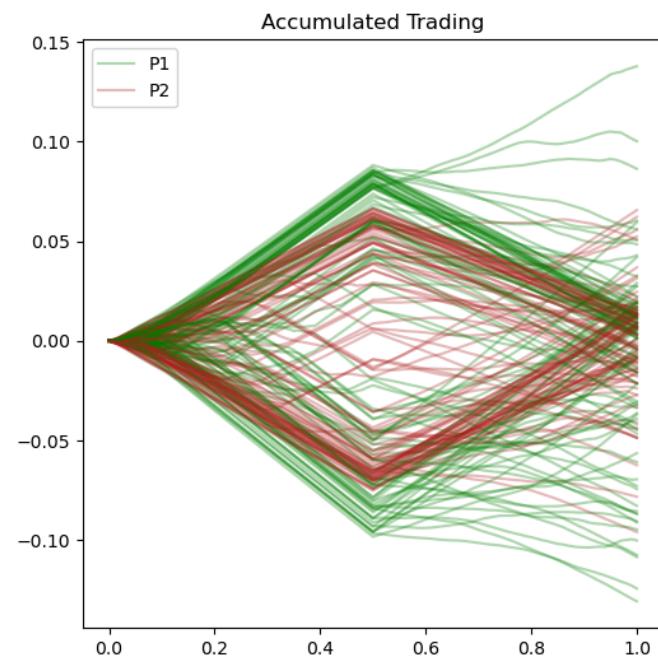
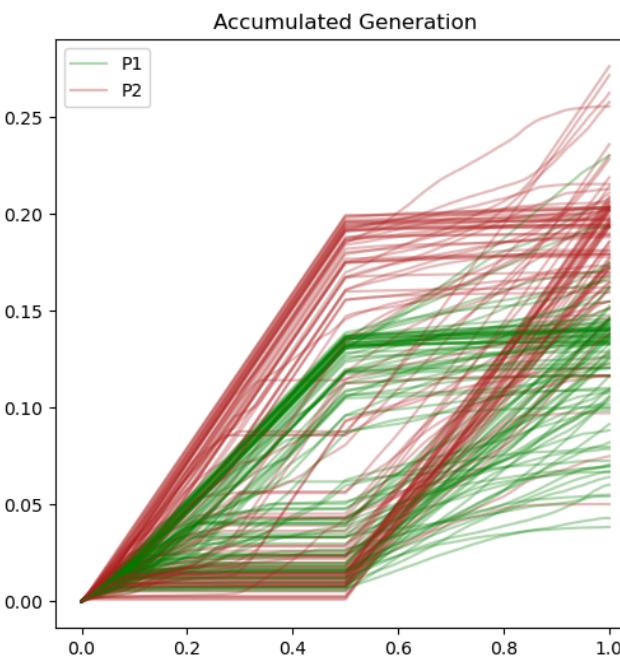
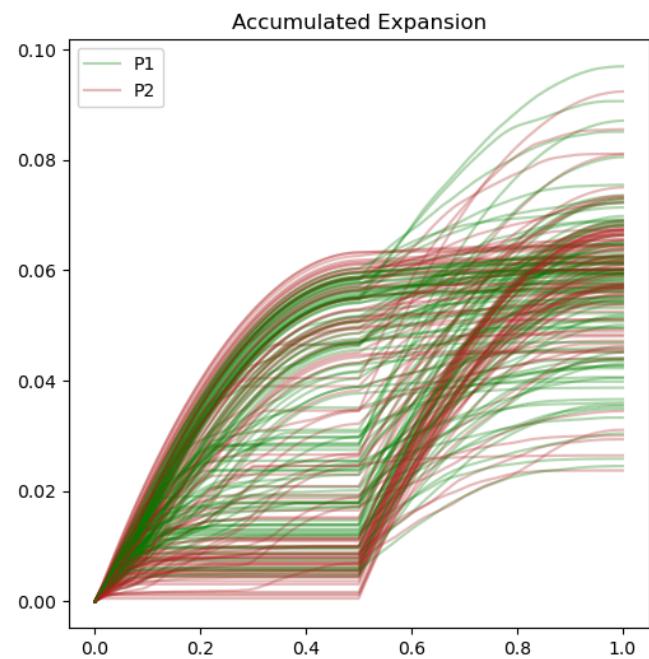
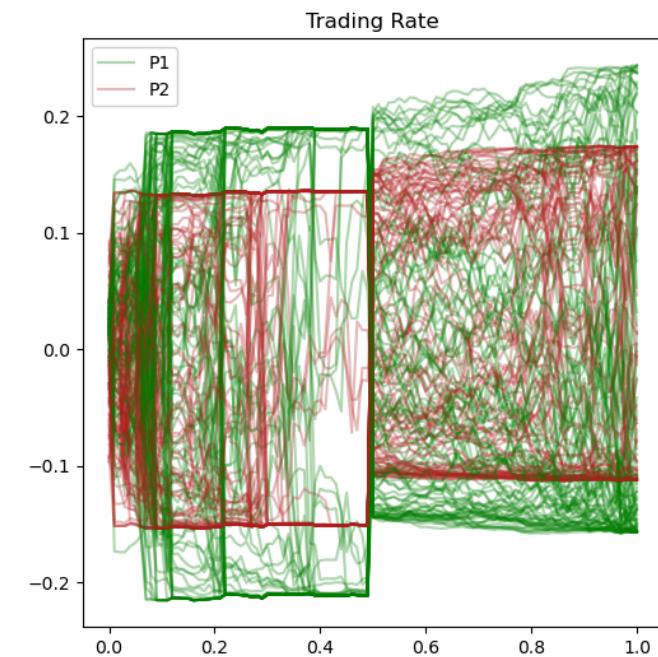
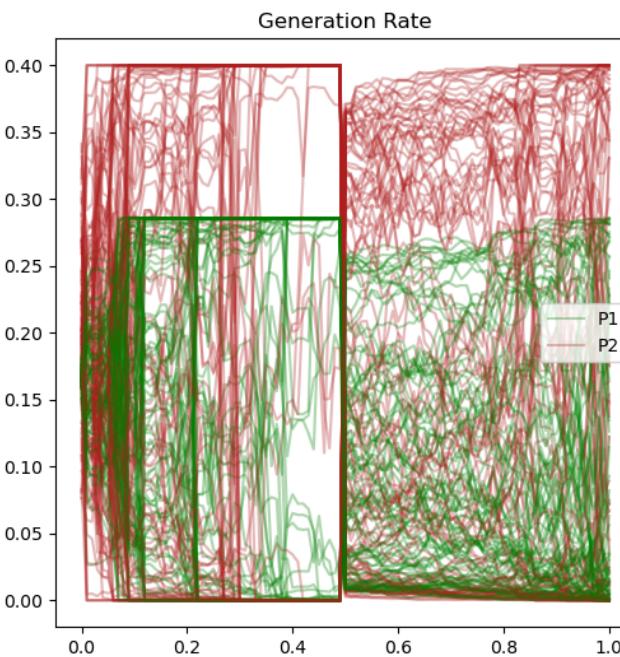
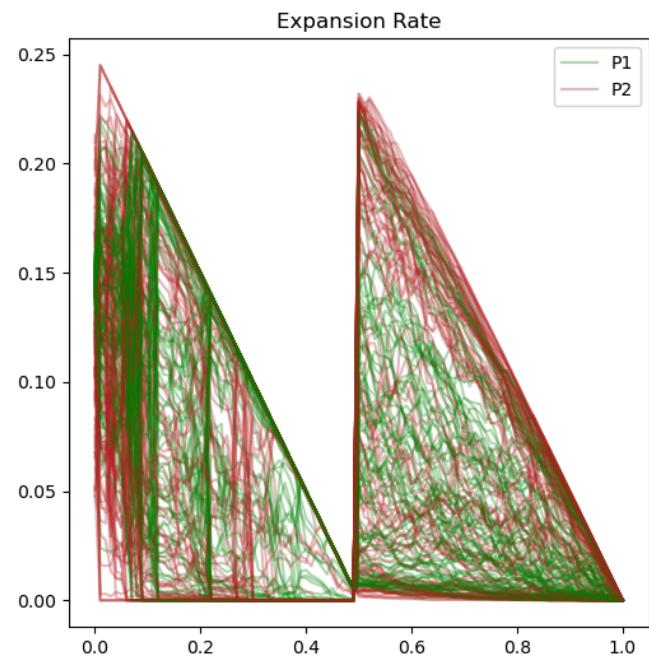
plot=plot_results(pop1_dict=pop1_dict, pop2_dict=pop2_dict, loss=loss)
plot.FwdLoss(log=True)
plot.Inventory_And_Price()
plot.Decomposition_Inventory()
plot.Key_Processes()
plot.Terminal_Convergence()

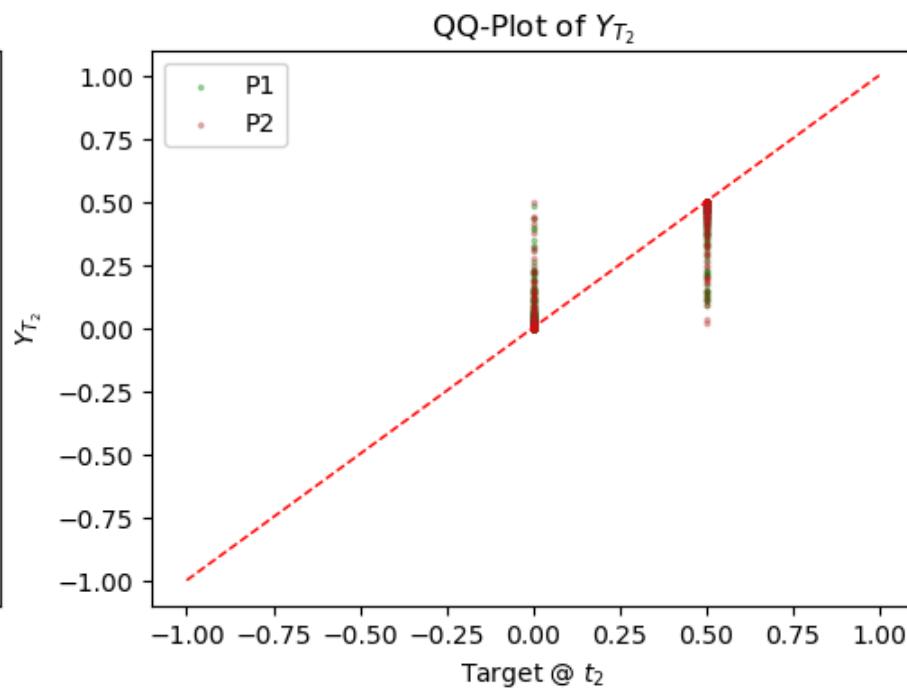
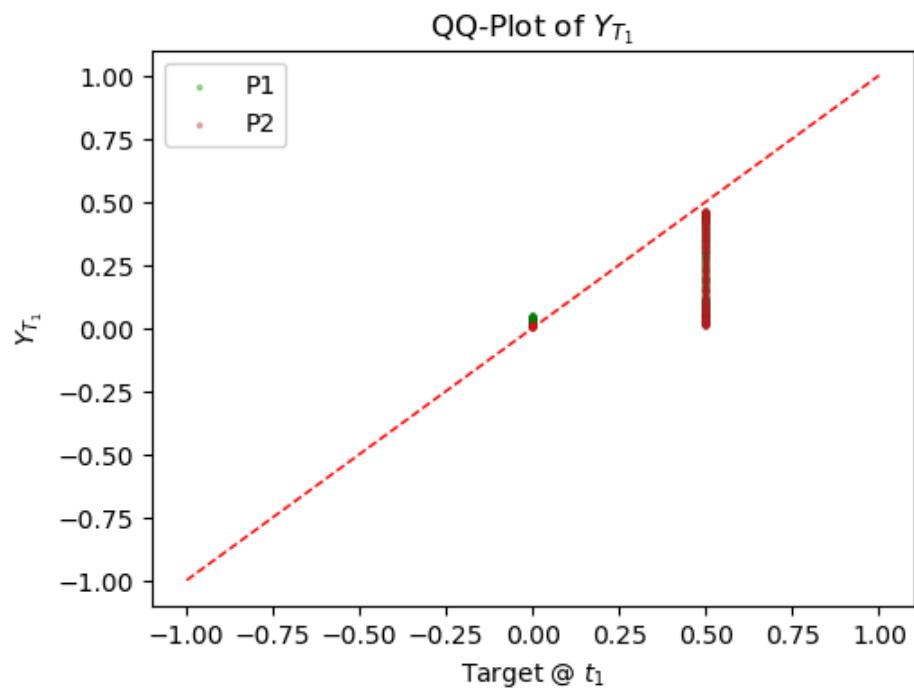
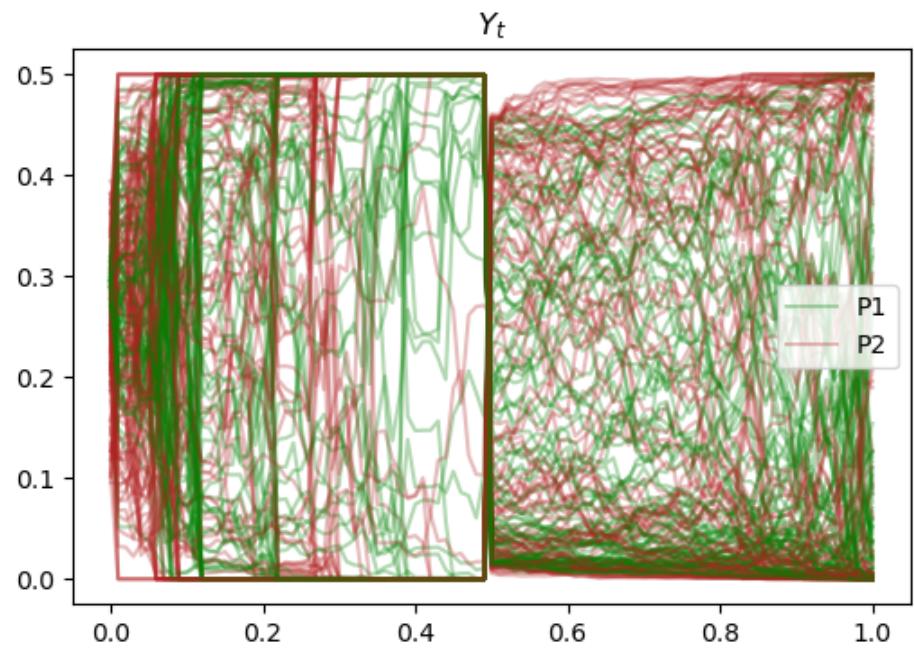
```

### Forward\_Loss vs Batch

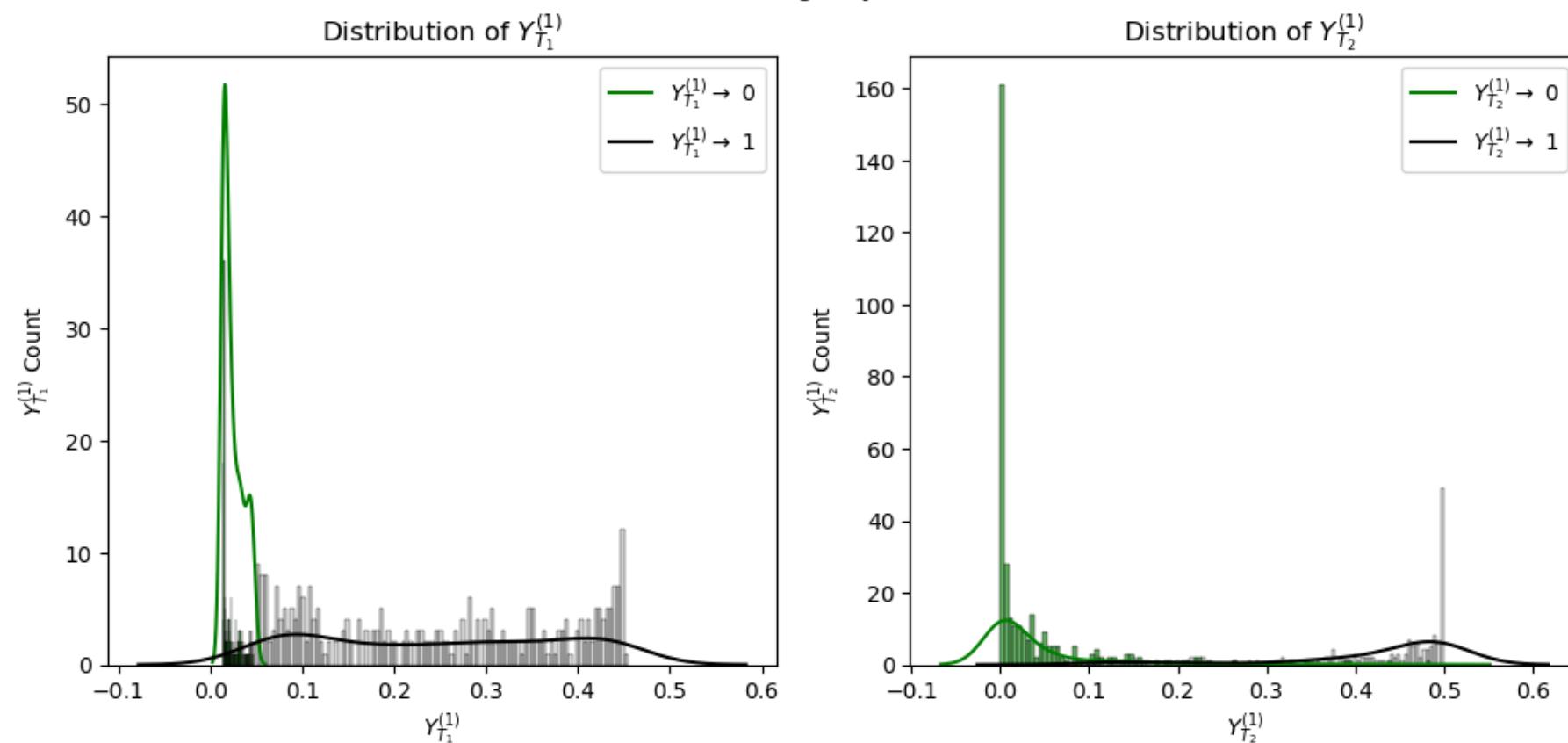


$Inventory(X_t)$  $Price(S_t)$ 

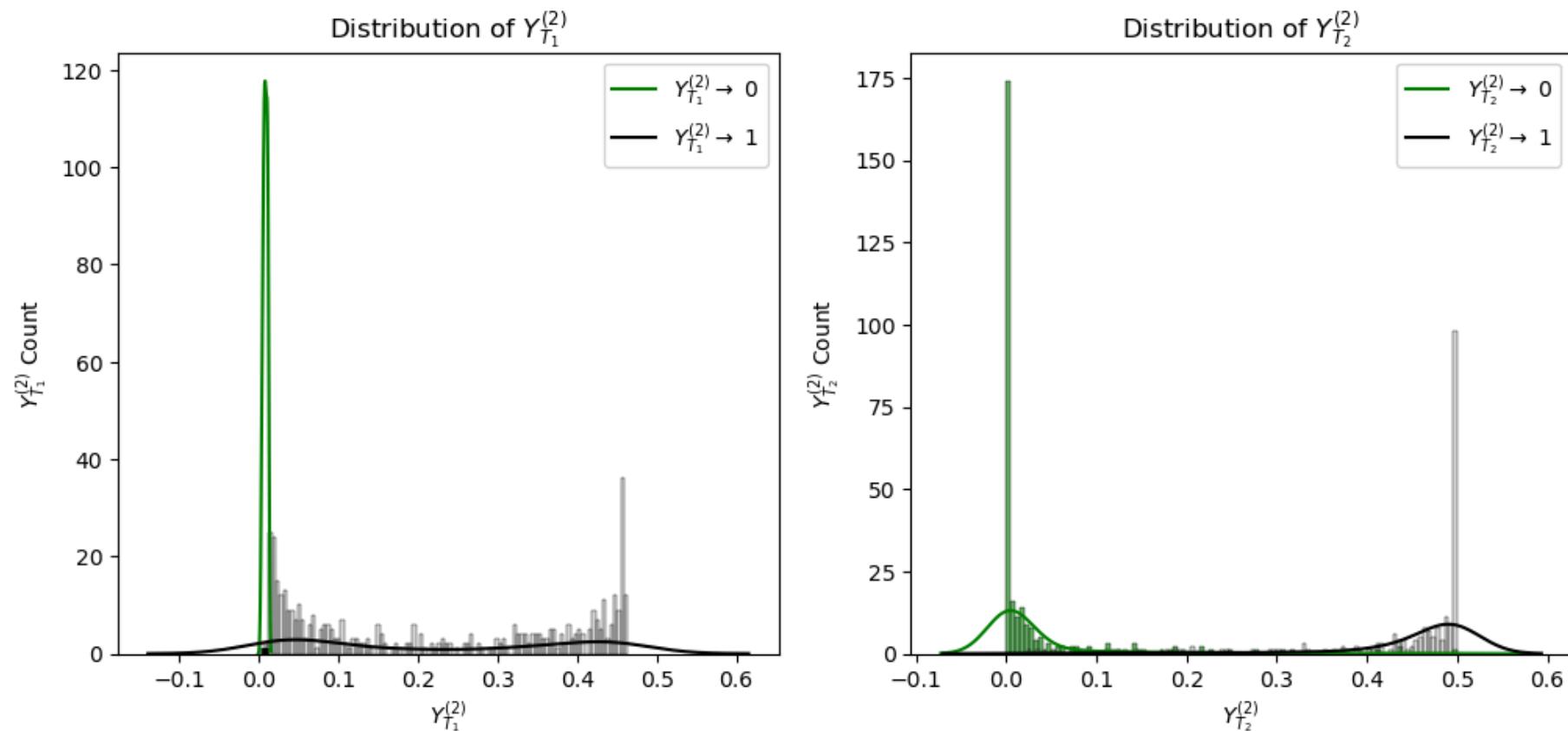




## Convergency - P1



## Convergency - P2



In [ ]: