

# Grando\_ABC\_all\_param

March 16, 2021

## 1 Grando ABC - Estimating all paramters

Calculations here were made using the same variables as used by grando.

• 1 U ([5, 15]) , 2 U ([16, 25]) and 3 U ([26, 35]) ; • 1 U ([0.03, 0.07]) , 2 U ([0.008, 0.02]) and 3 U ([0.001, 0.007]) ; • 1 U ([0.8, 1.4]) , 2 U ([0.3, 0.7]) and 3 U ([0.07, 0.2]) .

The data includes 200 iterations, again as per Grando.

Computation took 42.2841 seconds.

The graphs below correspond to the ones found in Grando's paper pages 75 and 76. As one may notice the results are extremely close to those found in the original paper.

```
[1]: import numpy as np
import seaborn as sns
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
```

```
[2]: headers = ['lambda1', 'lambda2', 'lambda3', 'xi_1', 'xi_2', 'xi_3', 'tau_1',
↳ 'tau_2', 'tau_3', 'distSit', 'distSitAss']
```

```
[3]: oc_reader.cpp sampler.cpp -o ggdata = pd.read_csv('avg.csv', names = headers)
```

```
[4]: data.head()
```

```
[4]:
```

	lambda1	lambda2	lambda3	xi_1	xi_2	xi_3	tau_1 \
0	7.72777	21.1273	33.5820	0.040752	0.019965	0.002872	1.056420
1	11.74540	17.3097	28.3375	0.041449	0.009954	0.001259	0.893861
2	12.46180	16.7491	28.1072	0.069705	0.017071	0.001841	1.293110
3	14.83420	16.1679	33.6090	0.037133	0.016443	0.005841	1.159330
4	11.20010	23.1095	27.7154	0.060054	0.010015	0.002161	1.370370

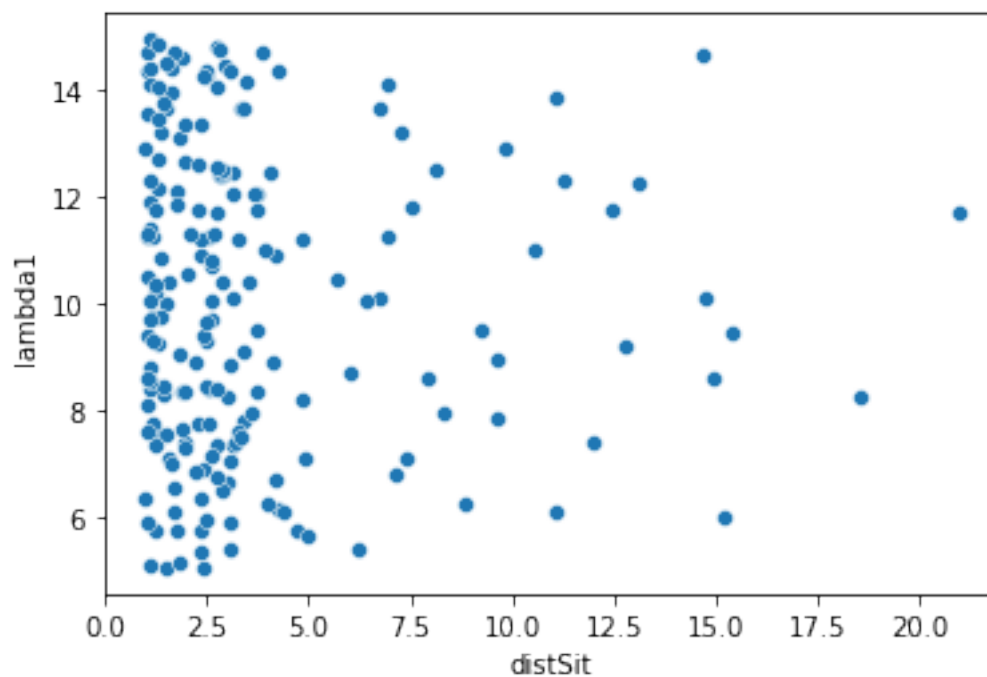
  

	tau_2	tau_3	distSit	distSitAss
0	0.441652	0.145414	2.29789	40.3844
1	0.656494	0.089387	12.41720	240.1140
2	0.530221	0.134914	4.08530	74.9581
3	0.653742	0.174706	2.74648	60.6030
4	0.347369	0.141211	3.24587	57.1348

```
[5]: # Lambda1 vs distance
```

```
sns.scatterplot(x = data['distSit'], y = data['lambda1'])
```

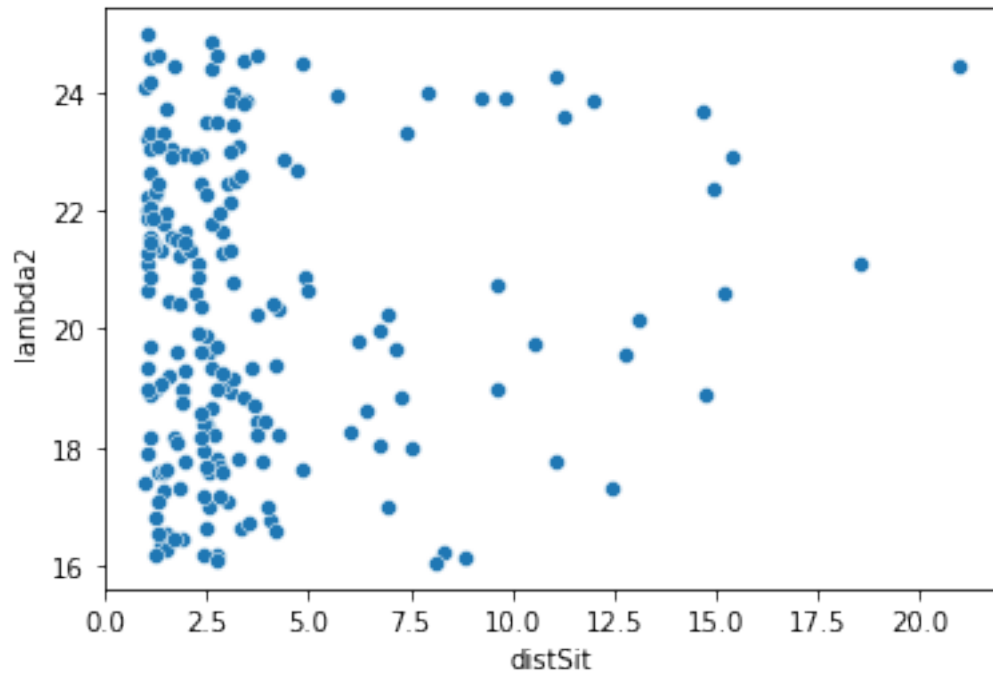
```
[5]: <AxesSubplot:xlabel='distSit', ylabel='lambda1'>
```



```
[6]: # Lambda2 vs distance
```

```
sns.scatterplot(x = data['distSit'], y = data['lambda2'])
```

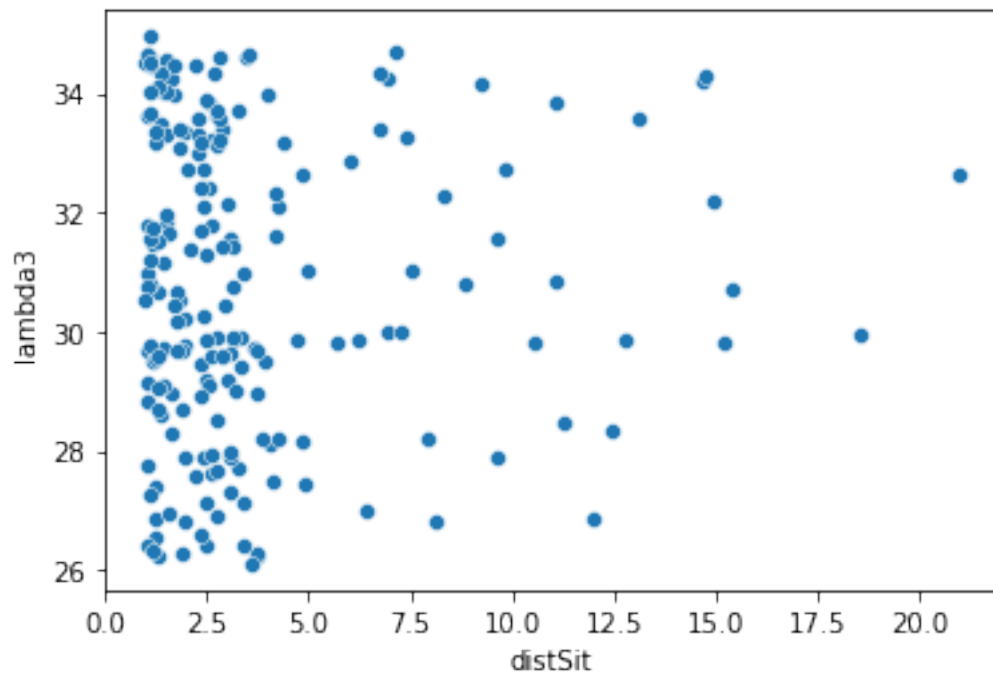
```
[6]: <AxesSubplot:xlabel='distSit', ylabel='lambda2'>
```



```
[7]: # Lambda3 vs distance
```

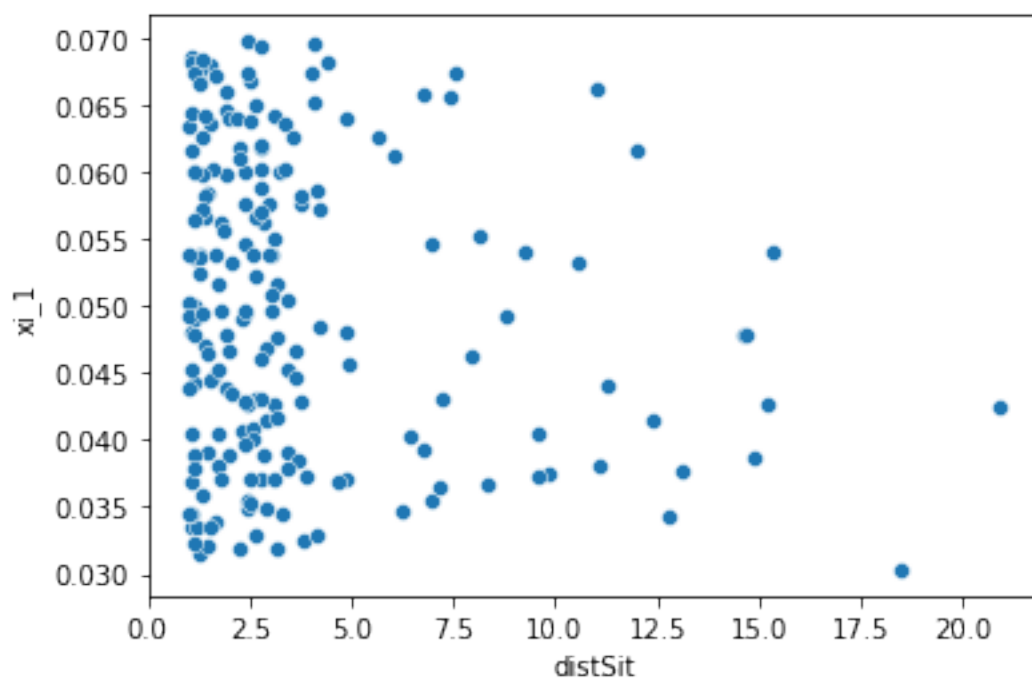
```
sns.scatterplot(x = data['distSit'], y = data['lambda3'])
```

```
[7]: <AxesSubplot:xlabel='distSit', ylabel='lambda3'>
```



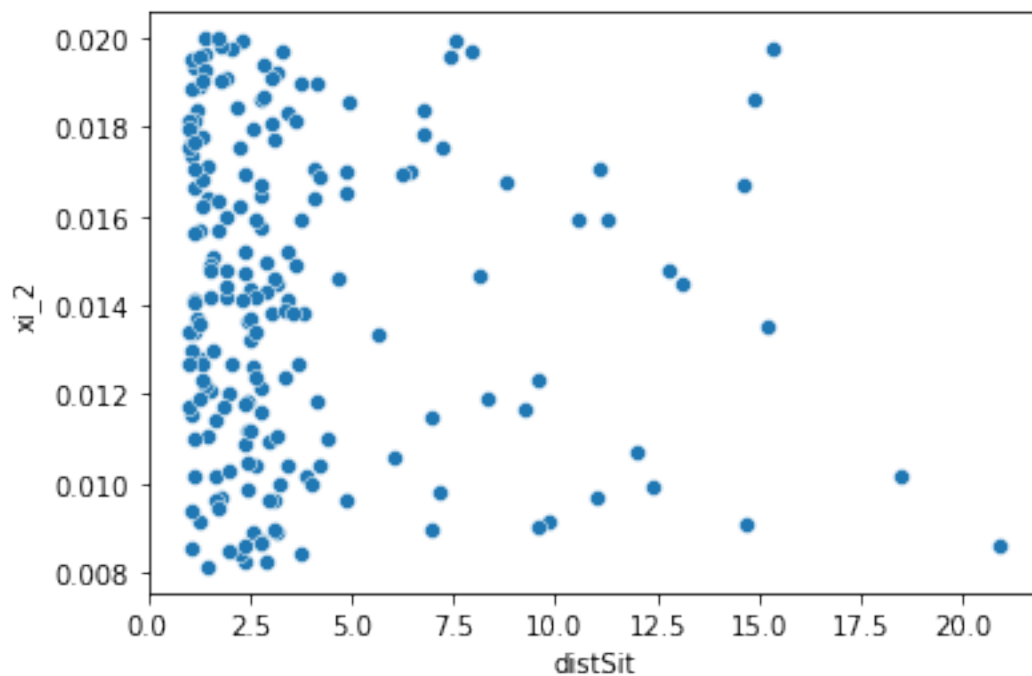
```
[8]: # xi_1 vs distance  
sns.scatterplot(x = data['distSit'], y = data['xi_1'])
```

```
[8]: <AxesSubplot:xlabel='distSit', ylabel='xi_1'>
```



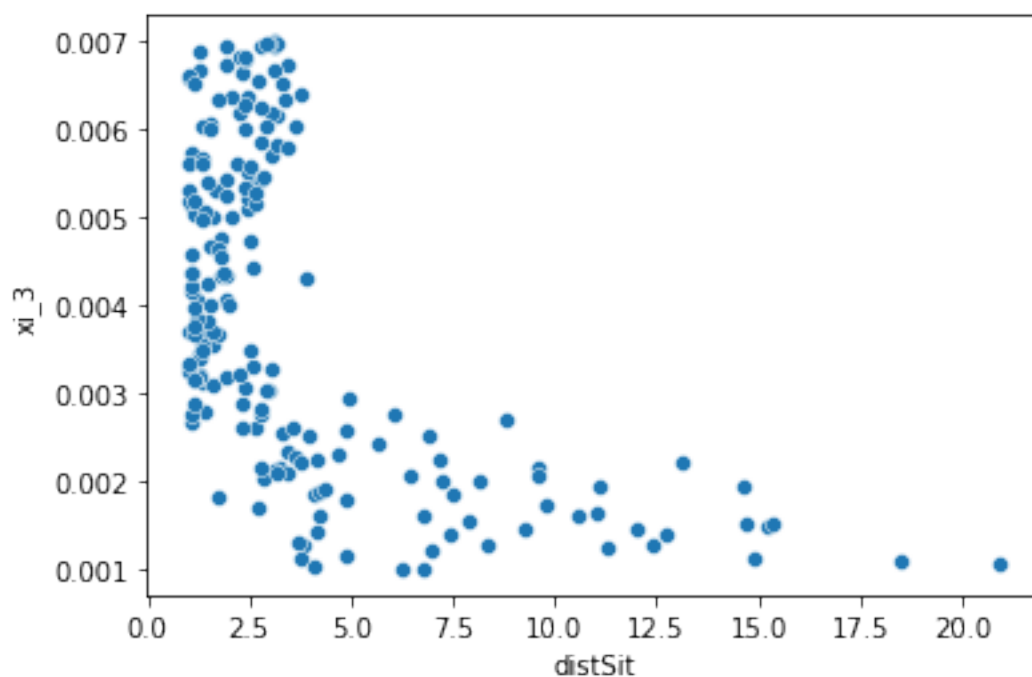
```
[9]: # xi_2 vs distance  
sns.scatterplot(x = data['distSit'], y = data['xi_2'])
```

```
[9]: <AxesSubplot:xlabel='distSit', ylabel='xi_2'>
```



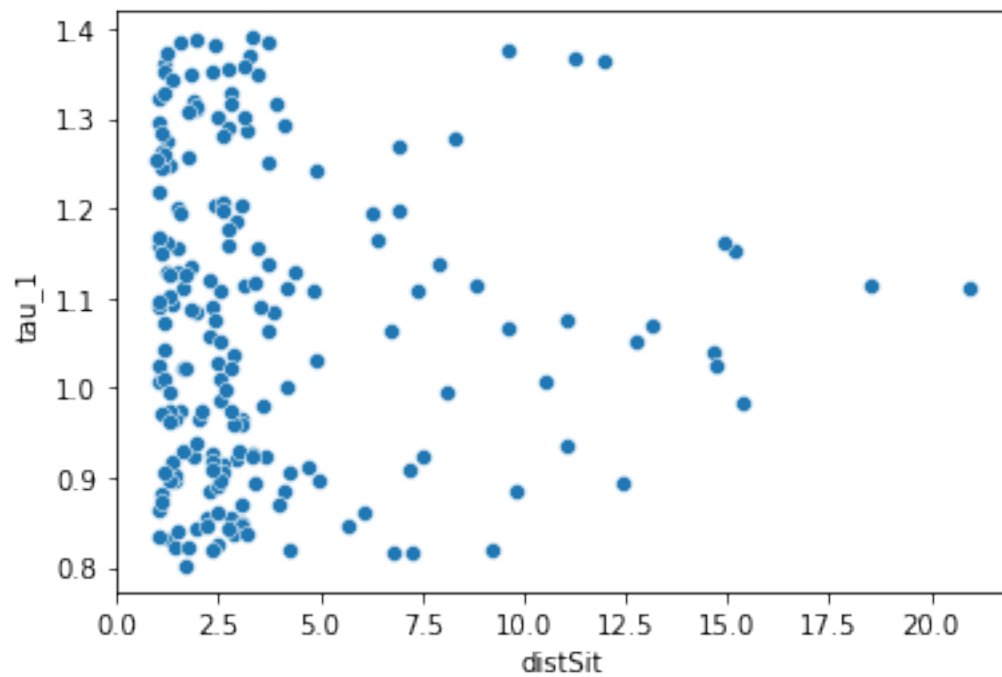
```
[10]: # xi_3 vs distance
sns.scatterplot(x = data['distSit'], y = data['xi_3'])
```

```
[10]: <AxesSubplot:xlabel='distSit', ylabel='xi_3'>
```



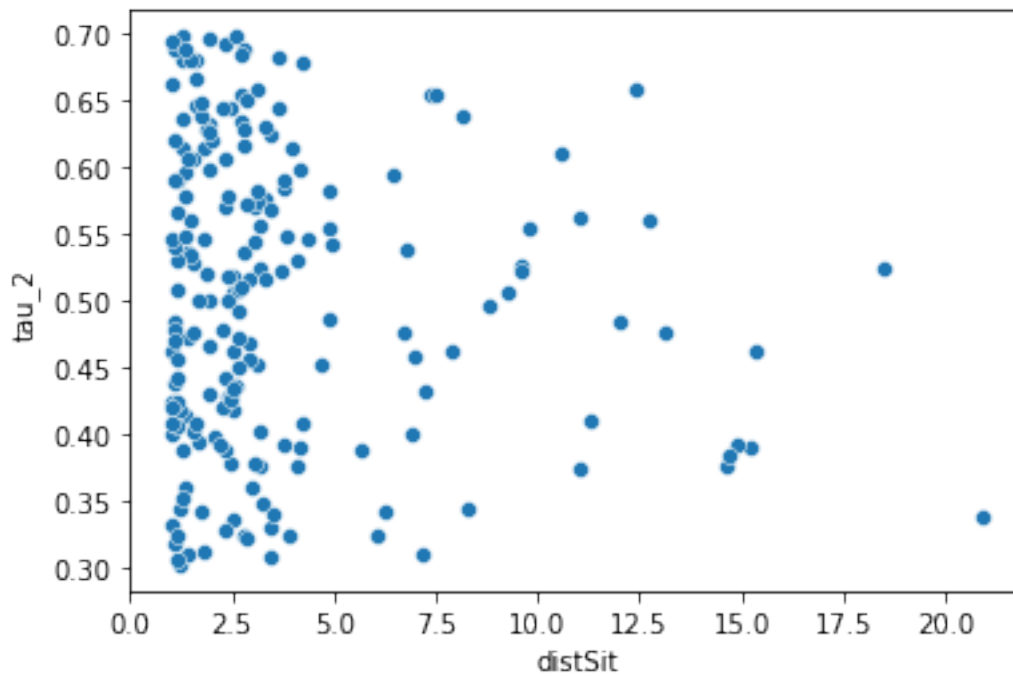
```
[11]: # tau_1 vs distance
sns.scatterplot(x = data['distSit'], y = data['tau_1'])
```

```
[11]: <AxesSubplot:xlabel='distSit', ylabel='tau_1'>
```



```
[12]: # tau_2 vs distance
sns.scatterplot(x = data['distSit'], y = data['tau_2'])
```

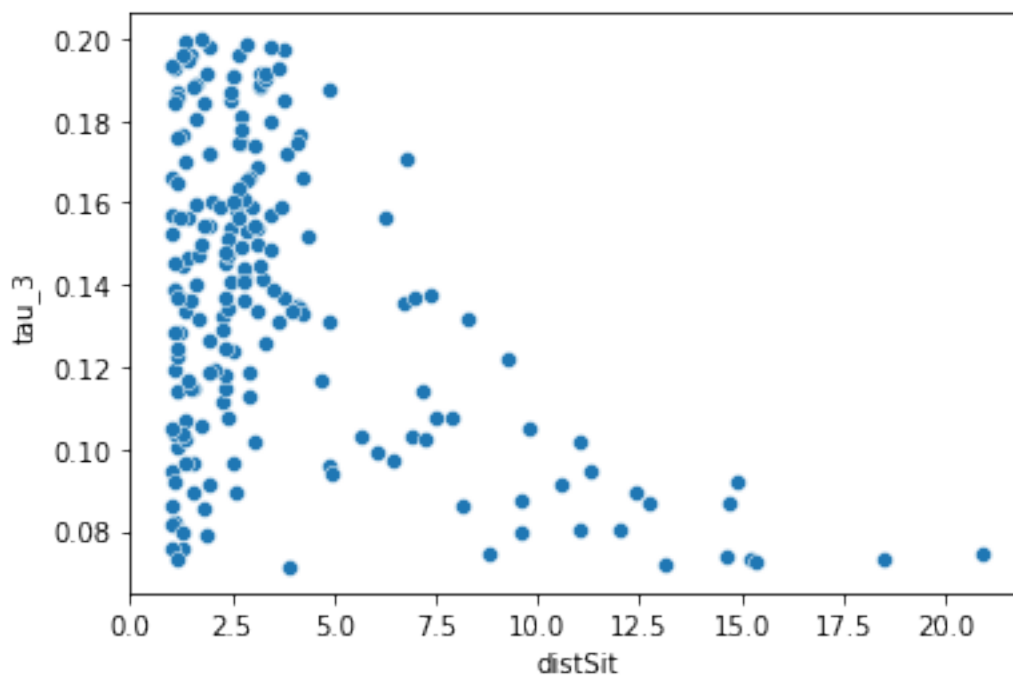
```
[12]: <AxesSubplot:xlabel='distSit', ylabel='tau_2'>
```



```
[13]: # tau_3 vs distance
```

```
sns.scatterplot(x = data['distSit'], y = data['tau_3'])
```

```
[13]: <AxesSubplot:xlabel='distSit', ylabel='tau_3'>
```



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