

# PhD Diary Week Beginning 19th November

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## 1 Lab work

- Spent 2 days in lab with Jeroen
  - Harvested leaf tissue
  - Did PCR
  - Ran gels on DNA to check for homozygous

## 2 Ideas

### 2.1 **TODO** Jeroen suggested using 2D diffusion model to work out potential diffusion of chitin

- This would allow us to know where to make our "no-mans-zone" for sampling before RNA-seq
- Possible reference here (Chalykh et al., 2014) for the diffusion constants of chitin

#### 2.1.1 **TODO** Find out diffusion constant of chitin

## 3 General Maths/Programming

This week I have been working on some more maths in relation to programming. In particular better utilising the numpy library to support faster operations. An online resource: "From Python to Numpy" (Rougier, 2016) has been most enlightening.

### 3.1 **DONE** Generators

<https://www.geeksforgeeks.org/use-yield-keyword-instead-return-keyword-python/>

### 3.2 **DONE** Python Decorators

### 3.3 **Timing/Profiling** functions

---

```
1 def timeit(method):
2     import time
3     def timed(*args, **kw):
4         ts = time.time()
5         result = method(*args, **kw)
6         te = time.time()
7         return (result, (te-ts))
8
9     return timed
```

---

### 3.4 Vectorising with Numpy

#### 3.4.1 OOP Approach

---

```

1 import matplotlib.pyplot as plt
2 from random import randint
3 import seaborn as sns
4 sns.set()
5 %matplotlib inline
6
7
8 class RandomWalker:
9     def __init__(self):
10         self.position = 0
11
12     def walk(self, n):
13         self.position = 0
14         for i in range(n):
15             yield self.position
16             self.position += 2*randint(0, 1)-1
17
18
19 @timeit
20 def make_walkers(x,y):
21     return [[p for p in RandomWalker().walk(y)] for _ in range(x) ]
22
23 N = 10000
24 res, time = make_walkers(N,N)
25 for walker in res:
26     plt.plot(walker)
27 plt.suptitle('Time taken: {0:.2f}'.format(time))

```

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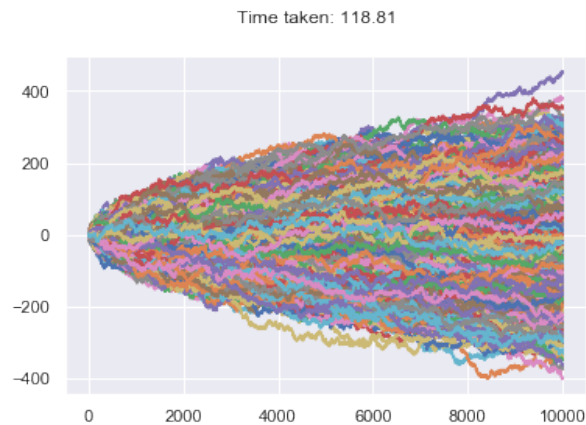


Figure 1: Random Walkers

### 3.4.2 Vectorised

---

```

1  from itertools import accumulate
2  from random import choices
3  import matplotlib.pyplot as plt
4  %matplotlib inline
5
6
7
8  def random_walk_faster(n=1000):
9      steps = choices([-1, +1], k=n)
10     return [0]+list(accumulate(steps))
11
12  @timeit
13  def make_walkers(x, y):
14      return [[p for p in random_walk_faster(x)] for _ in range(y)]
15
16
17  res, time = make_walkers(N,N)
18  for walker in res:
19      plt.plot(walker)
20  plt.suptitle('Time taken: {0:.2f}'.format(time))

```

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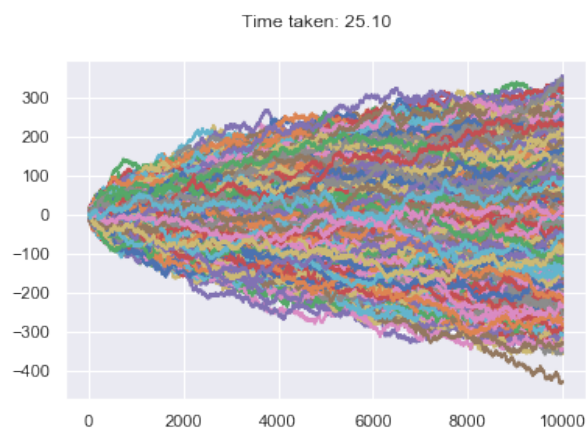


Figure 2: Random Walkers (vectorised)

### 3.4.3 Vectorised with numpy

---

```
1 import matplotlib.pyplot as plt
2 import numpy as np
3 %matplotlib inline
4
5
6 def random_walk_fastest(n=1000):
7     steps = np.random.choice([-1, 1], n)
8     return np.cumsum(steps)
9
10 @timeit
11 def make_walkers(x, y):
12     return [[p for p in random_walk_fastest(x)] for _ in range(y)]
13
14 res, time = make_walkers(N,N)
15
16 for walker in res:
17     plt.plot(walker)
18 plt.suptitle('Time taken: {0:.2f}'.format(time))
```

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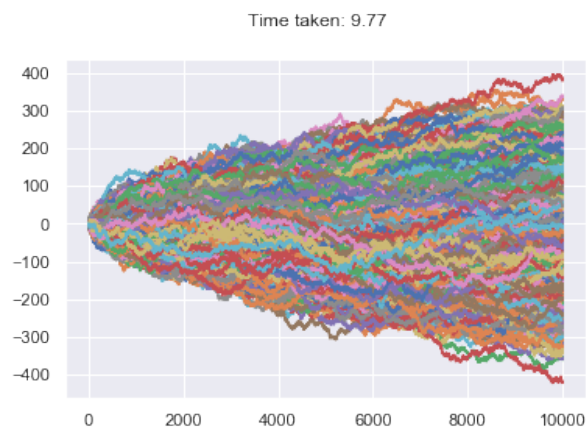


Figure 3: Random Walkers (numpy vectorised)

## References

- A. E. Chalykh, T. F. Petrova, R. R. Khasbiullin, and A. N. Ozerin. Water sorption on and water diffusion in chitin and chitosan. *Polymer Science Series A*, 56(5):614–622, September 2014. ISSN 1555-6107. doi: 10.1134/S0965545X14050034.
- Nicolas P. Rougier. *Rougier/from-Python-to-Numpy: Version 1.1*. Zenodo, December 2016. doi: 10.5281/zenodo.225783.