

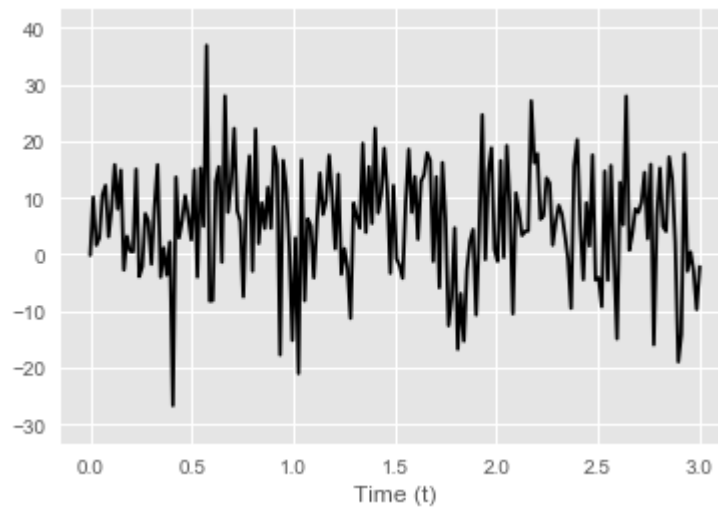
Handout 23: Power Spectral Density

Note: Gaussian process functionality is still in experimental phase and not available in the official Symulate release.

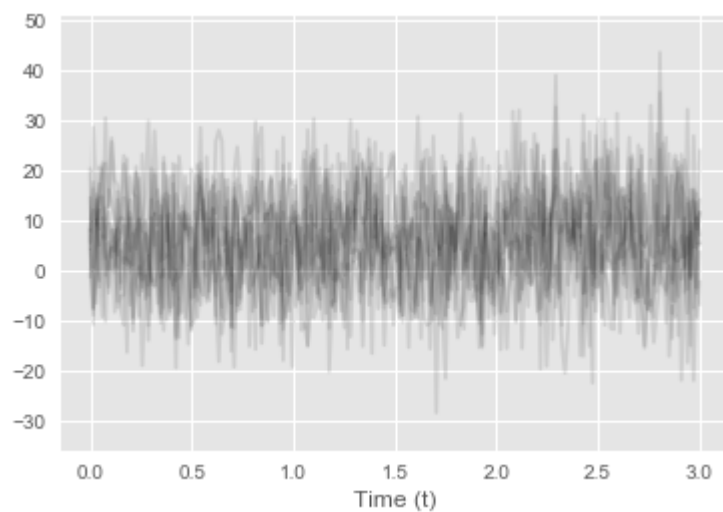
```
In [1]: from symulate import *  
        %matplotlib inline
```

Exer 23.2

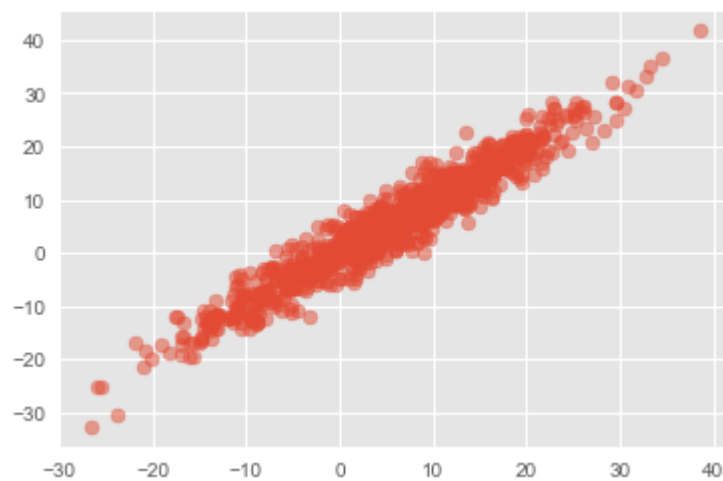
```
In [2]: P = GaussianProcess(mean_fn=lambda t: sqrt(30), cov_fn=lambda s, t: 100 *  
        exp(-40000 * (s - t) ** 2))  
        X = RandomProcess(P, TimeIndex(fs=inf))  
        X.sim(1).plot(alpha=1, tmax=3)
```



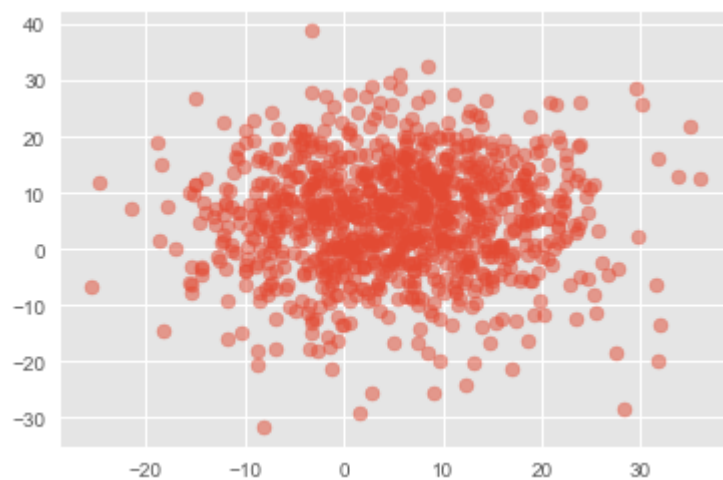
```
In [3]: X.sim(10).plot(tmax=3)
```



```
In [4]: (X[0] & X[0.001]).sim(1000).plot()
```

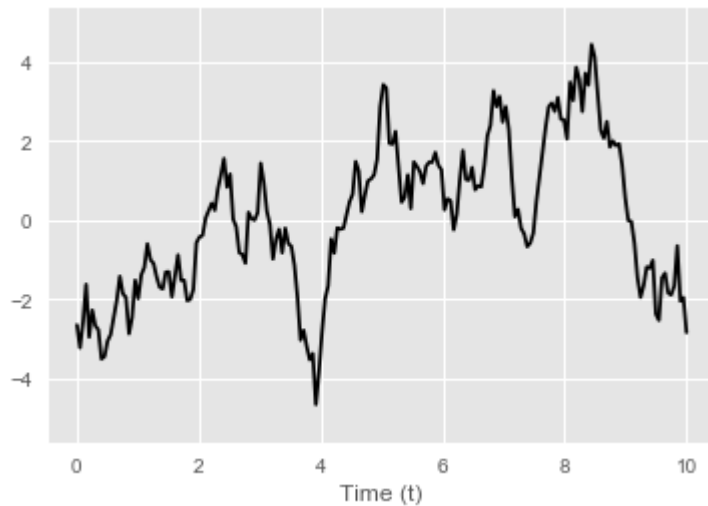


```
In [5]: (X[0] & X[0.01]).sim(1000).plot()
```



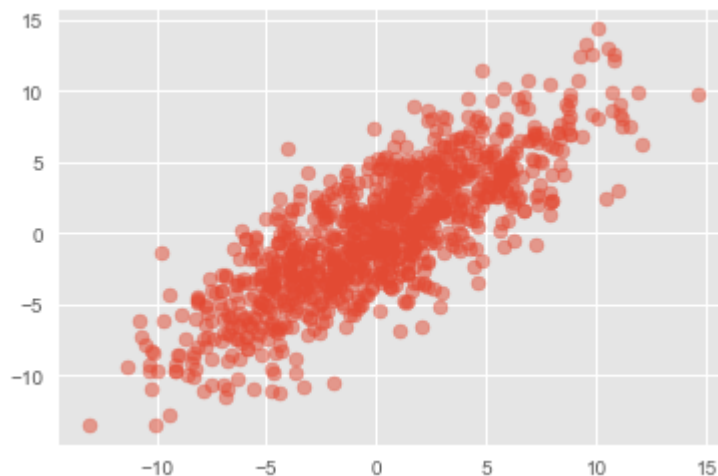
Exer 23.3

```
In [6]: def R1(s, t):  
        return 20 * max(0, 1 - abs(s-t) / 5)  
  
P1 = GaussianProcess(mean_fn=lambda t: 0, cov_fn=R1)  
X1 = RandomProcess(P1, TimeIndex(fs=inf))  
X1.sim(1).plot(alpha=1)
```



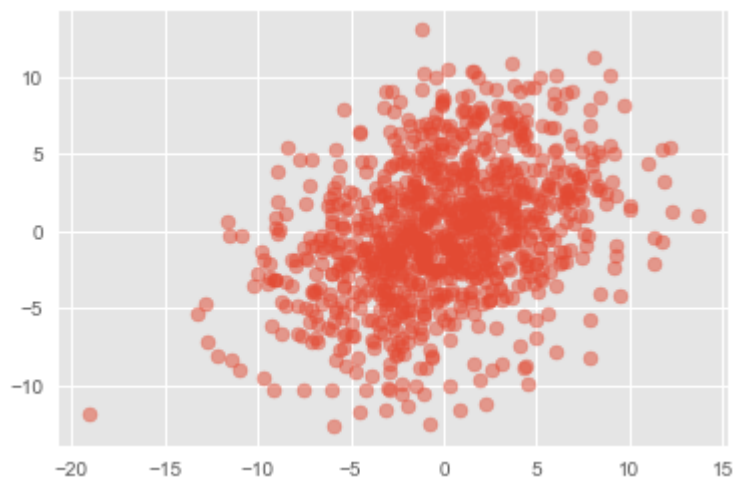
```
In [7]: xy = (X1[1] & X1[2]).sim(1000)  
xy.plot()  
xy.cov()
```

Out[7]: 16.765642709184025



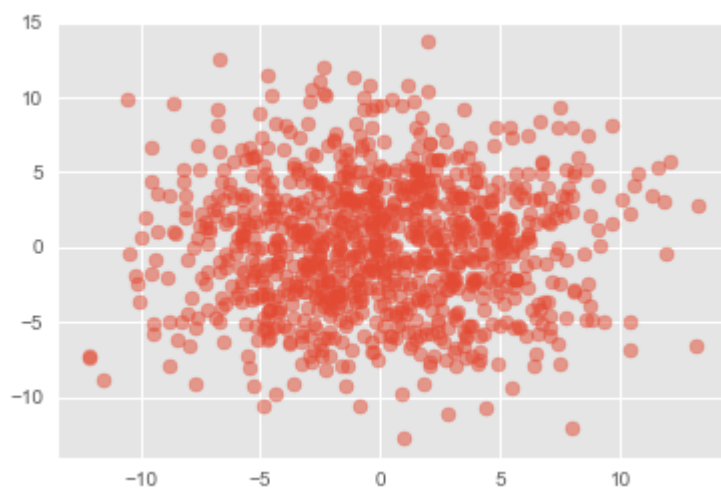
```
In [8]: xy = (X1[1] & X1[4]).sim(1000)
xy.plot()
xy.cov()
```

Out[8]: 6.9284262499788278

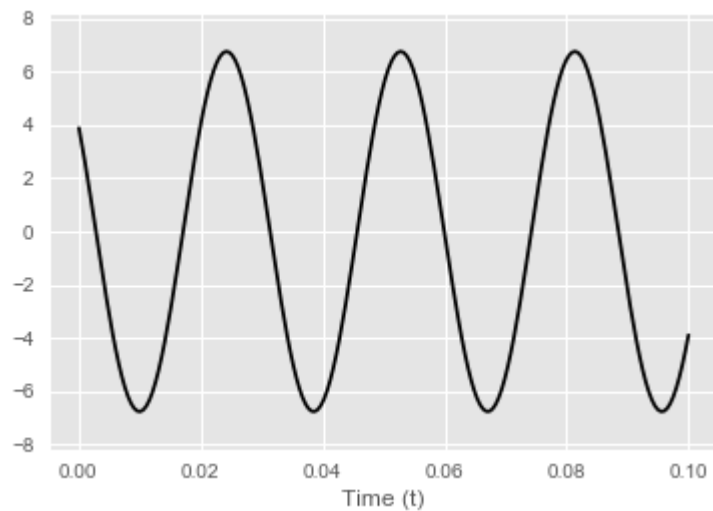


```
In [9]: xy = (X1[1] & X1[6]).sim(1000)
xy.plot()
xy.cov()
```

Out[9]: 0.43592533732092192

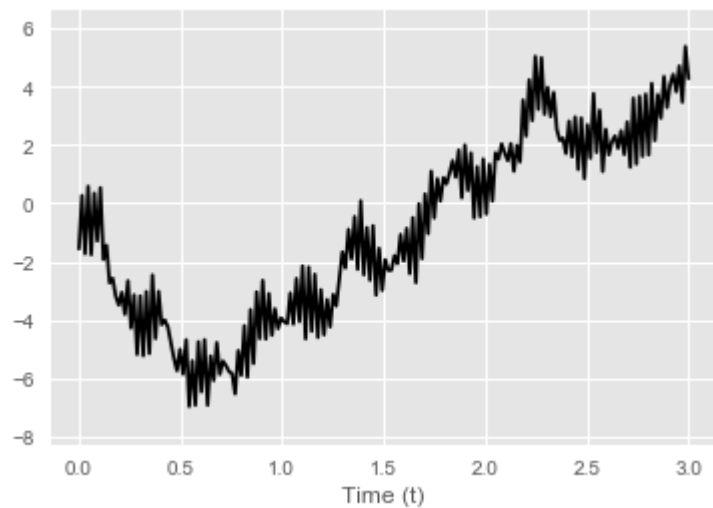


```
In [10]: def R2(s, t):  
         return 6 * cos(2 * pi * 35 * (s - t))  
  
P2 = GaussianProcess(mean_fn=lambda t: 0, cov_fn=R2)  
X2 = RandomProcess(P2, TimeIndex(fs=inf))  
X2.sim(1).plot(alpha=1, tmax=0.1)
```



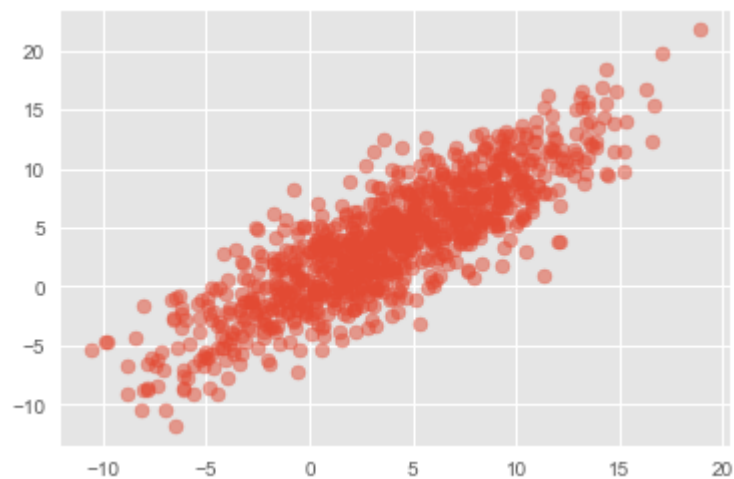
```
In [11]: X1, X2 = AssumeIndependent(X1, X2)  
Y = X1 + X2 + 4
```

```
In [12]: Y.sim(1).plot(alpha=1, tmax=3)
```



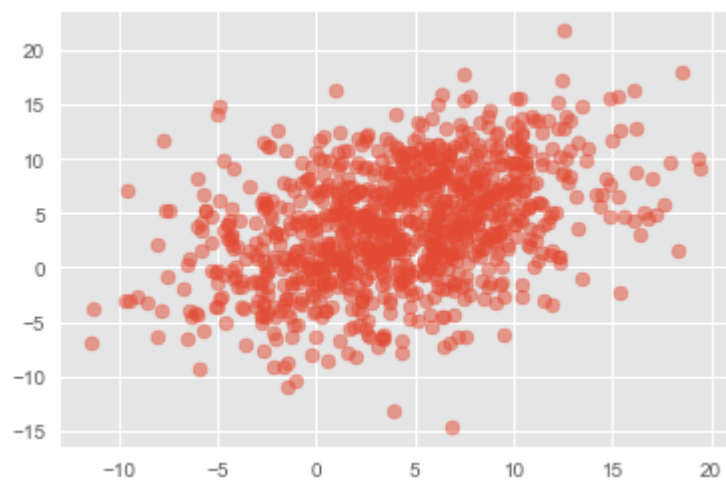
```
In [13]: xy = (Y[1] & Y[2]).sim(1000)
xy.plot()
xy.cov()
```

Out[13]: 22.226042672361569



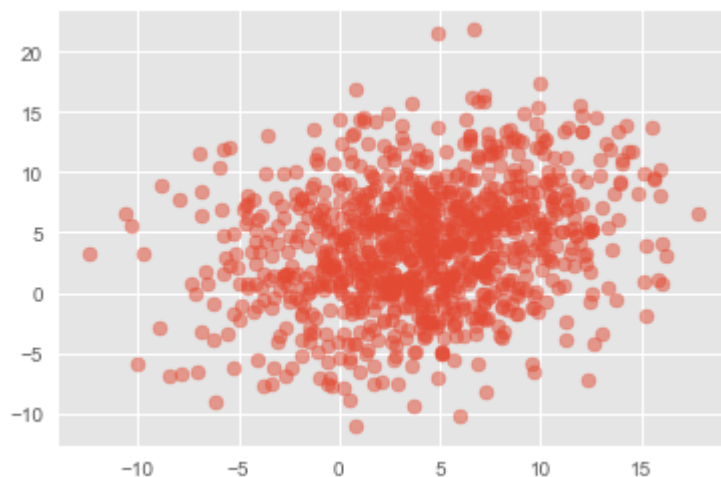
```
In [14]: xy = (Y[1] & Y[2 + 1/70]).sim(1000)
xy.plot()
xy.cov()
```

Out[14]: 10.903348303071702



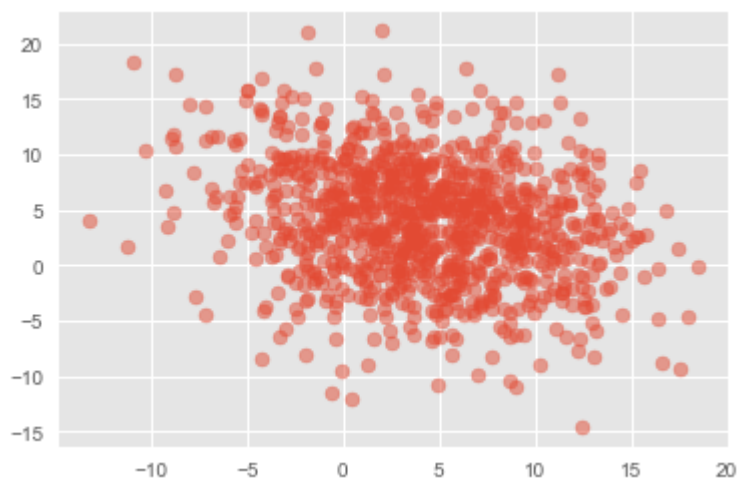
```
In [15]: xy = (Y[1] & Y[6]).sim(1000)
xy.plot()
xy.cov()
```

Out[15]: 6.1427686541344739



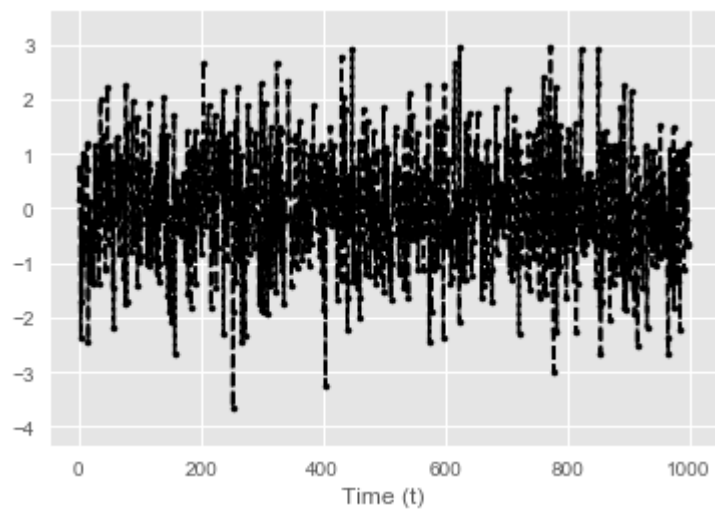
```
In [16]: xy = (Y[1] & Y[6 + 1/70]).sim(1000)
xy.plot()
xy.cov()
```

Out[16]: -6.799977650748569



Exer 23.4

```
In [32]: P = Normal(0, 1) ** inf  
RandomProcess(P).sim(1).plot(alpha=1,tmax=1000)
```



In []:

In []: