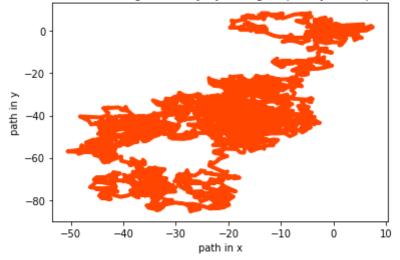
1a

```
In [2]: import matplotlib.pyplot as plt
        import numpy as np
        import random
        def path_finder(time_in_s, initial_x=0, initial_y=0):
            curr_pos_x = initial_x
            curr pos y = initial y
            global path_x
            global path y
            path_x = [curr_pos_x]
            path_y = [curr_pos_y]
            for n in range(0, time_in_s):
                curr_pos_x = curr_pos_x + np.random.normal()
                curr_pos_y = curr_pos_y + np.random.normal()
                path x.append(curr pos x)
                path_y.append(curr_pos_y)
        path_finder(3600)
        x = path_x
        y = path y
        fig = plt.figure()
        ax = plt.axes()
        #plt.axis([-50, 50, -50, 50])
        plt.plot(x, y, color='orangered', linewidth=4)
        plt.xlabel('path in x')
        plt.ylabel('path in y')
        plt.title('Path of ant wandering randomly by taking steps (x,y), one per
         second')
        #plt.scatter(x, y, color='orange', marker='^')
        #ax.xaxis.set(ticks=range(0,21),)
        plt.savefig('foo.png')
        plt.show()
```

Path of ant wandering randomly by taking steps (x,y), one per second



```
In [9]: import matplotlib.pyplot as plt
        import numpy as np
        import random
        from math import sqrt
        def path_finder(time_in_s, initial_x=0, initial_y=0):
            curr pos x = initial x
            curr pos y = initial y
            global path x
            global path y
            path_x = [curr_pos_x]
            path_y = [curr_pos_y]
            for n in range(0, time_in_s):
                curr_pos_x = curr_pos_x + np.random.normal()
                curr_pos_y = curr_pos_y + np.random.normal()
                path x.append(curr pos x)
                path y.append(curr pos y)
        def nest finder(time in s, nest dist, nest pos x=0, nest pos y=0):
            path finder(time in s)
            food pos x = path x[-1]
            food_pos_y = path_y[-1]
            curr pos x = food pos x
            curr pos y = food pos y
            found_nest = False
            for n in range(0, time in s):
                curr_pos_x = curr_pos_x + np.random.normal()
                curr pos y = curr pos y + np.random.normal()
                dist_from_nest = sqrt(curr_pos_x ** 2 + curr_pos_y ** 2)
                if dist from nest <= nest dist:</pre>
                     found nest = True
                    break
            if found nest:
                return True
            return False
        def prob_finder(time_in_s, num_trials, nest_dist):
            found = 0
            n = 0
            while n < num trials:</pre>
                if nest finder(time in s, nest dist) == True:
```

```
found += 1
n += 1
return found / num_trials

prob = prob_finder(3600, 1000, 10)
print(prob)
0.219
```

In [4]: ### The probability I get is 0.2x with 1000 trials. Obviously, this is n ot a good strategy because the ant is solely relying on a random steps t o find its way back to its nest. With 1000 trials, the probablility of f inding the nest is 0.2, which is certainly not a good number. Close to 8 0% of the time, the ant will end up wandering around unable to find its way back to the nest, therefore, this is not a good strategy.

1c

```
In [10]: def avg closest(time in s, nest pos x=0, nest pos y=0):
             path_finder(time_in_s)
             food_pos_x = path_x[-1]
             food pos y = path y[-1]
             curr pos x = food pos x
             curr_pos_y = food_pos_y
             closest_dist = sqrt(food_pos_x ** 2 + food_pos_y ** 2)
             for n in range(0, time_in_s):
                  curr_pos_x = curr_pos_x + np.random.normal()
                  curr_pos_y = curr_pos_y + np.random.normal()
                  curr_dist = sqrt(curr_pos_x ** 2 + curr_pos_y ** 2)
                  if curr_dist < closest_dist:</pre>
                      closest dist = curr dist
             return closest dist
         def prob finder(time in s, num trials):
             n = 0
             closest dist sum = 0
             while n < num trials:</pre>
                  closest dist sum += avg closest(time in s)
                  n += 1
             return closest dist sum / num trials
         closest_avg = prob_finder(3600, 1000)
         print(closest avg)
```

47.558947560537014

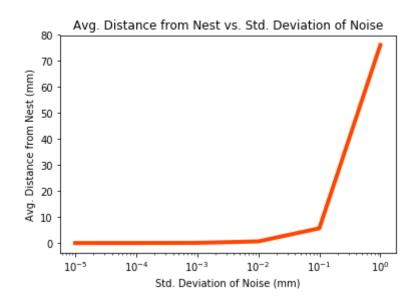
```
In [6]: ### Average closest distance would be 45 based on 1000 trial runs.
```

```
In [11]: import matplotlib.pyplot as plt
         import numpy as np
         import random
         from math import sqrt
         def memory vector finder(time in s, std dev, initial x=0, initial y=0):
             curr_pos_x = initial_x
             curr pos y = initial y
             memory_x = initial_x
             memory_y = initial_y
             for n in range(0, time_in_s):
                 x step = np.random.normal()
                 y_step = np.random.normal()
                 curr_pos_x = curr_pos_x + x_step
                 curr_pos_y = curr_pos_y + y_step
                 memory x = (memory x + x step + np.random.normal(scale=std_dev
         ))
                 memory y = (memory y + y step + np.random.normal(scale=std dev
         ))
             dist_x = curr_pos_x - memory_x
             dist_y = curr_pos_y - memory_y
             dist = sqrt(dist_x**2 + dist_y**2)
             return dist
         def avg_distance_finder(time_in_s, std_dev, num_trials):
             memory vector sum = 0
             for n in range(0, num trials):
                 memory_vector_sum += memory_vector_finder(time_in_s, std_dev)
             return memory vector sum / num trials
         t 1 = avg distance finder(3600, 1, 10)
         t_2 = avg_distance_finder(3600, 0.1, 10)
         t 3 = avg distance finder(3600, 0.01, 10)
         t = avg distance finder(3600, 0.001, 10)
         t = 5 = avg distance finder(3600, 0.0001, 10)
         t 6 = avg distance finder(3600, 0.00001, 10)
         print(t 1)
         print(t 2)
         print(t 3)
         print(t 4)
         print(t 5)
         print(t_6)
         y = [t 6, t 5, t 4, t 3, t 2, t 1]
         x = [0.00001, 0.0001, 0.001, 0.01, 0.1, 1]
         \#y = y[::-1]
         \#x = x[::-1]
         fig = plt.figure()
         ax = plt.axes()
```

```
ax.set_xscale('log')

plt.plot(x, y, color='orangered', linewidth=4)
plt.xlabel('Std. Deviation of Noise (mm)')
plt.ylabel('Avg. Distance from Nest (mm)')
plt.title('Avg. Distance from Nest vs. Std. Deviation of Noise')
plt.savefig('foo.png')
plt.show()
```

76.04009028314289 5.655185290515554 0.6183116161855854 0.06224101096903323 0.007462066447808923 0.0007538861241587004



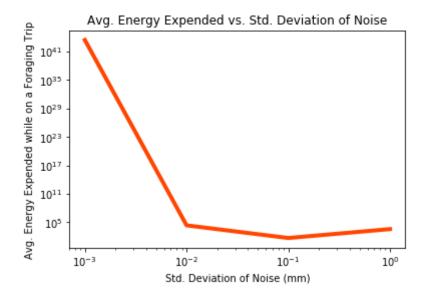
```
In [12]: import matplotlib.pyplot as plt
         import numpy as np
         import random
         from math import sqrt
         def memory vector finder(time in s, std dev, initial x=0, initial y=0):
             curr_pos_x = initial_x
             curr pos y = initial y
             memory_x = initial_x
             memory_y = initial_y
             for n in range(0, time_in_s):
                 x step = np.random.normal()
                 y_step = np.random.normal()
                 curr_pos_x = curr_pos_x + x_step
                 curr_pos_y = curr_pos_y + y_step
                 memory x = (memory x + x step + np.random.normal(scale=std_dev
         ))
                 memory y = (memory y + y step + np.random.normal(scale=std dev
         ))
             dist_x = curr_pos_x - memory_x
             dist_y = curr_pos_y - memory_y
             dist = sqrt(dist_x**2 + dist_y**2)
             return dist
         def avg distance finder(time in s, std dev, num trials):
             memory vector sum = 0
             for n in range(0, num trials):
                  memory_vector_sum += memory_vector_finder(time_in_s, std_dev)
             return memory vector sum / num trials
         t 1 = avg distance finder(3600, 1, 10)
         t_2 = avg_distance_finder(3600, 0.1, 10)
         t 3 = avg distance finder(3600, 0.01, 10)
         t = avg distance finder(3600, 0.001, 10)
         t = 5 = avg distance finder(3600, 0.0001, 10)
         t 6 = avg distance finder(3600, 0.00001, 10)
         t1_std = (t_1, 1)
         t2 std = (t 2, 0.1)
         t3 std = (t 3, 0.01)
         t4 \text{ std} = (t 4, 0.001)
         t5 \text{ std} = (t 5, 0.0001)
         t6 std = (t 6, 0.00001)
         def energy finder(t std dev):
             return (np.exp(0.1 / t std dev[1]) + (t std dev[0] ** 2))
         e1 = energy finder(t1 std)
         e2 = energy_finder(t2_std)
         e3 = energy_finder(t3_std)
         e4 = energy_finder(t4_std)
         e5 = energy finder(t5 std)
         e6 = energy finder(t6 std)
```

```
y = [e1, e2, e3, e4, e5, e6]
x = [1, 0.1, 0.01, 0.001, 0.0001, 0.00001]
#y.reverse()
#x.reverse()

fig = plt.figure()
ax = plt.axes()
ax.set_xscale('log')
ax.set_yscale('log')

plt.plot(x, y, color='orangered', linewidth=4)
plt.xlabel('Std. Deviation of Noise (mm)')
plt.ylabel('Avg. Energy Expended while on a Foraging Trip')
plt.title('Avg. Energy Expended vs. Std. Deviation of Noise')
plt.savefig('foo.png')
plt.show()
```

/Library/Frameworks/Python.framework/Versions/3.7/lib/python3.7/site-packages/ipykernel_launcher.py:47: RuntimeWarning: overflow encountered in exp



In []: ### The minimum energy is spent when the std. deviation is 0.1. From the evolutionary point of view, it makes perfect sense for ants to adapt to 0.1 std to spend least amount of energy because this is the point where there is optimal balance between the avg closest distance from a nest a nts end up at & the energy spent. If I compare the graph from 2 and 3a, when std=1.0, the distance from nest is significantly larger than when std=1.0 and energy expenditure is also farily larger, so it is unfeasib le for an ant to have evolved to adapt to std=1.0. When std=0.01, there is tiny gain to reduce distance from distance, but its advantage is ver y minimal compared to the significantly larger energy expenditure. There fore, std of 0.1 is the perfect number for an ant to have evolved to ada

ding its way back to the nest.

pt to balance the energy and the probabiltiy of finding the food and fin