```
In [1]:
           1 import numpy as np
           import matplotlib.pyplot as plt
from scipy.stats import norm
           4 import pandas as pd
           5 from scipy.integrate import odeint
           6 from scipy.integrate import solve_ivp
```

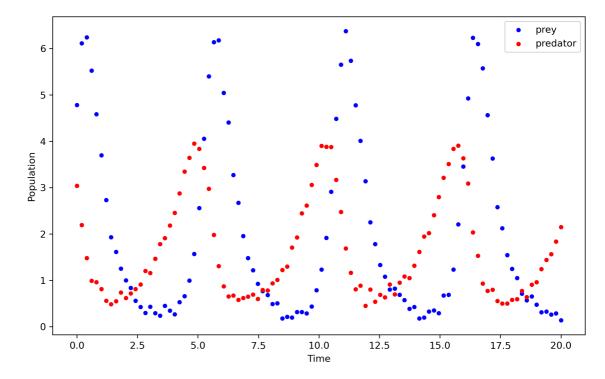
Opening predator-prey dataset

```
1 df = pd.read_csv('predator-prey-data.csv', index_col=False)
2 df.head()
```

Out[2]:		Unnamed: 0	t	x	У
	0	0	0.000000	4.781635	3.035257
	1	1	0.202020	6.114005	2.189746
	2	2	0.404040	6.238361	1.478907
	3	3	0.606061	5.520524	0.989836
	4	4	0.808081	4 582546	0 957827

```
In [3]:
            # Loading data into read-only numpy arrays
            data = df[['t','x','y']].values
          2
            # data[1], data[2] = data[2].copy(), data[1].copy()
            data.flags.writeable = False
          5
          6
          7
            # Plotting
          8 plt.figure(dpi =300, figsize=(10, 6))
          9
            point_width = 13
         10 plt.scatter(data[:,0], data[:,1], label = 'prey', color = 'blue', s =pc
         plt.scatter(data[:,0], data[:,2], label = 'predator', color = 'red', s=
         12 plt.ylabel('Population')
         13 plt.xlabel('Time')
         14 plt.legend()
         15
```

Out[3]: <matplotlib.legend.Legend at 0x13fc52f10>



Objective functions

Defining volterra equations function

```
In [5]: #Function that will return the data for predator and prey for a given s
def predator_prey_integration(time,initial_conditions,parameters):
    alpha,beta,delta,gamma = parameters
    results = odeint(predator_prey_odes,initial_conditions, time, args=
    predator_values,prey_values = results[:,0], results[:,1]
    return np.array([predator_values,prey_values]).T
```

Defining objective functions

```
In [101]:
               def weighted_sse(actual, predicted):
                   '''Weighted sum of squared erros'''
            2
            3
            4
                   sd = 0.01
            5
            6
                   sd_list = sd*actual #List of estimated standard deviations
            7
            8
                   inv_sd = 1/sd_list #List of inverted standard deviations from sd_li
            9
                   weighted_sse = np.sum(sd_list*((actual - predicted)**2))
           10
           11
           12
                   #weighted_sse =
           13
           14
                   return weighted_sse
           15
           16
           17
              def MAE(actual, predicted):
                   '''Mean absolute error'''
           18
           19
                   return np.mean(np.abs(actual - predicted))
           20
           21
           22
           23
              def MSE(actual, predicted):
                   '''Mean squared error'''
           24
           25
                   return np.mean((actual - predicted)**2)
 In [7]:
              test = np.array([1,2,3])
            1
            2
            3
              mult = 3*test
            4 print(mult)
            6
              sum = np.sum(mult*test)
            7
            8
              print(sum)
          [3 6 9]
          42
```

Algorithms & Optimisation

Simulated Annealing

```
In [8]:
                     1
                           def random_walk_annealing(parameters): #A random walk designed for anne
                     2
                                    lst = [parameter + np.random.normal(0, 0.5) for parameter in parameter
                     3
                                    # Ensure all elements are positive
                     4
                                   while any(x \leftarrow= 0 for x in lst):
                     5
                                            for indx in range(len(lst)):
                     6
                                                     if lst[indx] <= 0:</pre>
                     7
                                                             lst[indx] = max(0, parameters[indx] + np.random.normal(
                     8
                     9
                                    return 1st
                   10
                   11
                   12
                   13
                          def simulated_annealing(initial_temp,cooling_constant, data, time, init
                   14
                   15
                                   temp = initial_temp #Scaling factor for random movement. We square
                   16
                                    start = parameters #Initial starting parameters
                   17
                                   x_n = start
                   18
                                    scores = [] #A score is just the value of the objective function ev
                   19
                   20
                                    current_est = predator_prey_integration(time, initial_conditions, )
                                    current_score = objective(data, current_est) #The current value of
                   21
                                    scores.append(current_score) #Keeping track of the values of the objections of the objection of the objectio
                   22
                   23
                   24
                                    \#cur = function(x) \#The function value of the current x solution
                   25
                                   history = [x_n] #Stores previously searched x values
                   26
                   27
                                    for i in range (max_iterations):
                   28
                   29
                                            proposal = random walk annealing(x n) #A new proposal for the
                   30
                                            new_est = predator_prey_integration(time, initial_conditions, r
                   31
                                            new_score = objective(data, new_est) #Calculate new value of ob
                   32
                   33
                                            delta = new_score - current_score #Difference in objective fund
                   34
                                            #if proposal < 0 or proposal > 1:
                   35
                   36
                                                   #proposal = x n # Reject proposal by setting it equal to pre
                   37
                   38
                                            acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate
                   39
                   40
                                            #if delta < 0:
                   41
                                                  #x n = proposal ##Accept proposal
                   42
                                                     #current score = new score
                   43
                   44
                                            if np.random.rand() < acceptance_probability: #else if it is nd</pre>
                   45
                                                     x_n = proposal #Accept proposal
                   46
                                                     current_score = new_score
                   47
                   48
                                            scores.append(current_score)
                   49
                                            temp = cooling_constant**i * initial_temp #Cool temperature
                   50
                                            #print(temp)
                   51
                                            history.append(x_n) #Add to history
                   52
                   53
                                    return x n, scores
```

```
In [90]:
           1 | input data = data[:,1:3]
             initial_conditions = [input_data[0][0], input_data[0][1]]
           2
           3
             t = data[:,0]
           5 #Taking random draw for initial parameters (initial guess)
           6 | alpha = np.random.uniform(0,1)
           7 beta = np.random.uniform(0,1)
           8 delta = np.random.uniform(0,1)
           9 gamma = np.random.uniform(0,1)
          10 parameters = [alpha, beta, delta, gamma]
          11 | initial_temp = 20
          12 cooling_constant = 0.10
          13
          14 | # Using MSE
          15 | x_best, scores = simulated_annealing(initial_temp,cooling_constant, inc
```

/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in exp
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: divide by zero encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability

```
In [184]:
              input_data = data[:,1:3]
            2 initial_conditions = [input_data[0][0], input_data[0][1]]
            3 | t = data[:,0]
            4
            5
              #Taking random draw for initial parameters (initial guess)
            6 | alpha = np.random.uniform(0,1)
            7 beta = np.random.uniform(0,1)
            8 delta = np.random.uniform(0,1)
            9 gamma = np.random.uniform(0,1)
           10 parameters = [alpha, beta, delta, gamma]
           11 | initial_temp = 20
           12 cooling_constant = 0.10
           13
           14 # Using MSE
           15 x best, scores = simulated annealing(initial temp, cooling constant, inc
```

/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in exp
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: divide by zero encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability

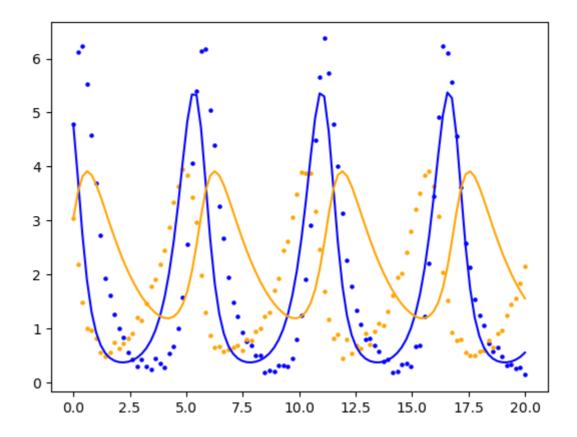
Solution discovery over iterations

LOWEST MSE: 1.3145243092280952

Curve fit

```
In [186]:
             # t, x, y = data
           1
             initial_conditions = [input_data[0][0], input_data[0][1]]
           2
           3
             t = data[:,0]
           4
             scaling = 2
           5
             parameters = x_best
           7
           8
             # Using MSE
           9
             x = predator_prey_integration(t,initial_conditions,parameters)
          10
          11 #plt.figure(dpi =300, figsize=(6, 5))
          12 point width = 13
             plt.plot(t, x[:,0],color = "b")
          13
             plt.plot(t, x[:,1],color = "orange")
          14
          15
          plt.scatter(t, data[:,1], color= 'blue', s = 5)
             plt.scatter(t, data[:,2], color= 'orange', s = 5)
          17
          18
          19 | #mse_prey = MSE(data[:,1],x[:,0]) #MSE for fitted curve
          21
             #mse_total = mse_prey + mse_predator
             print("MAE: " +str(scores[-1]))
```

MAE: 1.3145243092280952



Calculation of mean and variance

```
In [96]:
           1
              ### Distribution of parameters for multiple runs
           2
           3
              def multiple_runs_annealing(initial_temp,cooling_constant,input_data,t,
           4
           5
                  mse total list = []
           6
           7
           8
                  for i in range(n_runs):
           9
          10
                      x_best, scores = simulated_annealing(initial_temp,cooling_const
          11
          12
                      mse = scores[-1]
          13
                      mse_total_list.append(mse) #Add total MSE for this simulation t
          14
          15
          16
                  return mse_total_list
          17
                                                                                      Þ
```

```
In [106]:
               input_data = data[:,1:3]
            2
               initial_conditions = [input_data[0][0], input_data[0][1]]
            3 | t = data[:,0]
            4 | #Taking random draw for initial parameters (initial guess)
            5 alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7
              delta = np.random.uniform(0,1)
              gamma = np.random.uniform(0,1)
              parameters = [alpha, beta, delta, gamma]
            9
              parameters = [alpha, beta, delta, gamma]
           10
           11
           12 | initial_temp = 20
           13
              cooling_constant = 0.10
           14
           15
              mse_total_list = multiple_runs_annealing(initial_temp,cooling_constant,
           16
           17
              #print(mse total list)
```

/var/folders/rc/tn2ys5g55157vhhlmn0 vfwr0000gq/T/ipykernel 58892/385309489 9.py:38: RuntimeWarning: overflow encountered in exp acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /var/folders/rc/tn2ys5g55157vhhlmn0 vfwr0000gq/T/ipykernel 58892/385309489 9.py:38: RuntimeWarning: overflow encountered in scalar divide acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /var/folders/rc/tn2ys5g55157vhhlmn0 vfwr0000gq/T/ipykernel 58892/385309489 9.py:38: RuntimeWarning: divide by zero encountered in scalar divide acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /Users/alex 1/anaconda3/lib/python3.11/site-packages/scipy/integrate/ odep ack_py.py:248: ODEintWarning: Excess work done on this call (perhaps wrong Dfun type). Run with full_output = 1 to get quantitative information. warnings.warn(warning_msg, ODEintWarning)

```
In [107]: 1 print(mse_total_list)
```

[2.6473227546360665, 2.7464957827676564, 2.8201901313461293, 2.68480725254 8282, 2.6589224841462658, 2.5022908556330266, 2.4583578426621084, 2.439019 9669793287, 2.460643513633285, 2.543829159819853]

```
In [108]: 1
2    mean_mse_annealing = np.mean(mse_total_list)
3    std_mse_annealing = np.std(mse_total_list)
4    print("Average MSE = " + str(mean_mse_annealing))
6    print("Standard deviation of MSE = " + str(std_mse_annealing))
```

Average MSE = 2.5961879744172007 Standard deviation of MSE = 0.12680792857239243

FOR WEIGHTED SSE

```
In [189]:
            1
            2
            3 input_data = data[:,1:3]
            4 | initial_conditions = [input_data[0][0], input_data[0][1]]
            5 t = data[:,0]
            6 #Taking random draw for initial parameters (initial quess)
            7 alpha = np.random.uniform(0,1)
            8 beta = np.random.uniform(0,1)
            9 delta = np.random.uniform(0,1)
           10 | gamma = np.random.uniform(0,1)
           11 parameters = [alpha, beta, delta, gamma]
           12 parameters = [alpha, beta, delta, gamma]
           13
           14 | initial_temp = 20
           15 cooling_constant = 0.10
           16
           17 | mse_total_list = multiple_runs_annealing(initial_temp,cooling_constant,
           18
```

/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in exp
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: divide by zero encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability

```
In [190]: 1 print(mse_total_list)
2 print(np.std(mse_total_list))
```

[1.4283946885408745, 1.3067741402910857, 1.4442398586927112, 1.28940092054 71197, 1.4204590248705167, 1.4440930616053458, 1.3031423818937582, 1.35896 96758657763, 1.3465599587990993, 1.3043258816810026] 0.06051240976134452

Comparison of cooling schedules

```
input_data = data[:,1:3]
In [221]:
              initial_conditions = [input_data[0][0], input_data[0][1]]
            2
            3 | t = data[:,0]
            4 #Taking random draw for initial parameters (initial guess)
            5 alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7
              delta = np.random.uniform(0,1)
            8 gamma = np.random.uniform(0,1)
            9
              parameters = [alpha, beta, delta, gamma]
           10
              parameters = [alpha, beta, delta, gamma]
           11
           12 | initial_temp = 20
           13
           14
              cooling_constants = np.arange(0.10,1,0.10)
           15
           16 mean_mse_per_simulation = [] #The mean MSE per simulation
           17
              sd mse per simulation = [] #The standard deviation per simulation
           18
           19
              for constant in cooling constants:
           20
           21
                  mse_total_list = multiple_runs_annealing(initial_temp,constant,inpu
           22
                  mean_mse_annealing = np.mean(mse_total_list)
           23
                  std_mse_annealing = np.std(mse_total_list)
                  mean_mse_per_simulation.append(mean_mse_annealing)
           24
           25
                   sd_mse_per_simulation.append(std_mse_annealing)
           26
           27
```

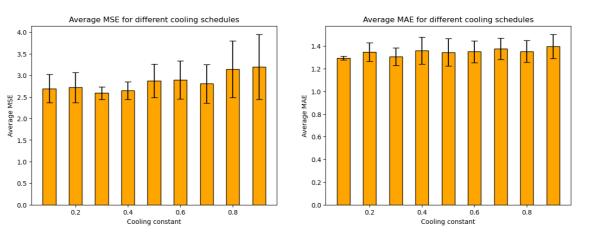
```
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in exp
 acceptance probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/Users/alex_1/anaconda3/lib/python3.11/site-packages/scipy/integrate/_odep
ack_py.py:248: ODEintWarning: Excess work done on this call (perhaps wrong
Dfun type). Run with full_output = 1 to get quantitative information.
  warnings.warn(warning msg, ODEintWarning)
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in scalar divide
  acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: divide by zero encountered in scalar divide
  acceptance probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
```

```
In [222]:
               mean_mae_per_simulation = []
               sd_mae_per_simulation = []
            2
            3
              for constant in cooling_constants:
            5
            6
                   mae_total_list = multiple_runs_annealing(initial_temp,constant,inpl
            7
                   mean_mae_annealing = np.mean(mae_total_list)
            8
                   std_mae_annealing = np.std(mae_total_list)
            9
                   mean_mae_per_simulation.append(mean_mae_annealing)
           10
                   sd_mae_per_simulation.append(std_mae_annealing)
           11
```

/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in exp
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: overflow encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability
/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489
9.py:38: RuntimeWarning: divide by zero encountered in scalar divide
 acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan
ce probability

```
In [223]:
            2
              fig, axes = plt.subplots(1, 2, figsize=(15, 5))
            3
            4
              bar_width = 0.05
            5
            6
               axes[0].bar(cooling_constants, mean_mse_per_simulation, yerr = sd_mse_p
            7
               axes[1].bar(cooling_constants, mean_mae_per_simulation, yerr = sd_mae_p
            8
            9
               axes[0].set_xlabel("Cooling constant")
           10
               axes[1].set_xlabel("Cooling constant")
           11
               axes[0].set_ylabel("Average MSE")
           12
           13
               axes[1].set_ylabel("Average MAE")
           14
              axes[0].set_title("Average MSE for different cooling schedules")
           15
               axes[1].set_title("Average MAE for different cooling schedules")
           16
           17
           18
           19
           20
               #plt.bar(cooling_constants, mean_per_simulation, yerr = sd_per_simulati
              #plt.ylabel("Average MSE")
           21
           22 #plt.xlabel("Cooling constant")
              #plt.title("Average MSE for different cooling schedules (30 simulations
           23
           24
           25
               #plt.show()
           26
```

Out[223]: Text(0.5, 1.0, 'Average MAE for different cooling schedules')



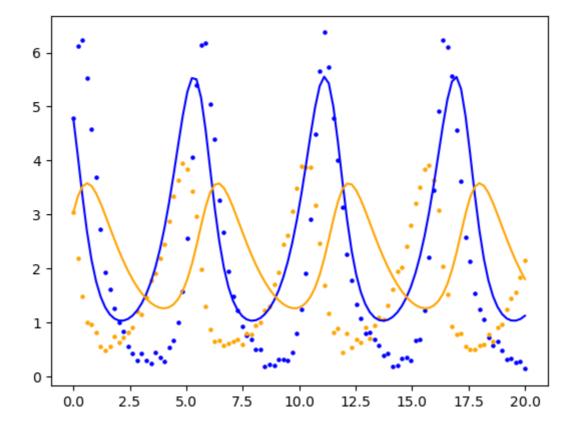
Hill climbing

```
In [238]:
            1
            2
            3
               def random_walk(parameters):
            4
                   lst = [parameter + np.random.normal(0, 0.5) for parameter in parame
            5
                   # Ensure all elements are positive
            6
                   while any(x \leftarrow 0 for x in lst):
            7
                       for indx in range(len(lst)):
            8
                            if lst[indx] <= 0:</pre>
            9
                                lst[indx] = max(0, parameters[indx] + np.random.normal(
           10
           11
                   return 1st
           12
              def hill_climbing(data, time, initial_conditions, parameters, objective
           13
                   '''Tries to find the best solution using random walker'''
           14
                   # Initialize starting parameter state
           15
           16
                   scores = []
           17
                   x n = parameters
           18
           19
                   current_est = predator_prey_integration(time, initial_conditions, )
           20
                   current_score = objective(data, current_est)
           21
                   scores.append(current_score)
           22
           23
                   for _ in range(max_iterations):
           24
                       # Generate a random walk for parameters
           25
                       x_n_1 = random_walk(x_n)
           26
           27
                       # Calculate the current and next estimations
                       current_est = predator_prey_integration(time, initial_conditior
           28
           29
                       new_estimation = predator_prey_integration(time, initial_condit
           30
                       new_score = objective(data, new_estimation)
           31
                       # If the next estimation is better, update the parameters
           32
           33
                       if new_score < current_score:</pre>
           34
                           current_score = new_score
           35
                           x n = x n 1
           36
                            scores.append(current score)
           37
           38
                   return x_n, scores
           39
In [239]:
            1
               input_data = data[:,1:3]
              initial_conditions = [input_data[0][0], input_data[0][1]]
            2
            3 | t = data[:,0]
              #Taking random draw for initial parameters (initial guess)
              alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7 delta = np.random.uniform(0,1)
            8 gamma = np.random.uniform(0,1)
            9
               parameters = [alpha, beta, delta, gamma]
           10
           11 # Using MSE
           12 x_best,scores = hill_climbing(input_data, t, input_data[0], parameters,
```

Curve fit

```
In [241]:
               # t, x, y = data
               initial_conditions = [input_data[0][0], input_data[0][1]]
            2
            3
              t = data[:,0]
            4
            5
              parameters = x_best
            7
              # Using MSE
            8
              x = predator_prey_integration(t,initial_conditions,parameters)
            9
           10
           11
           12 point_width = 13
               plt.plot(t, x[:,0],color = "b")
           13
              plt.plot(t, x[:,1],color = "orange")
           14
           15
           16 | plt.scatter(t, data[:,1], color= 'blue', s = 5)
              plt.scatter(t, data[:,2], color= 'orange', s = 5)
           17
           18
           19 | mse_prey = MSE(data[:,1],x[:,0]) #MSE for fitted curve
           20 mse_predator = MSE(data[:,2],x[:,1])
           21 | mse_total = mse_prey + mse_predator
              print("Mean Square Error: " +str(mse_total))
           22
           23
```

Mean Square Error: 5.385424572357688



```
In [195]:
              def multiple_runs_hill_climbing(input_data, t, initial_conditions, para
            1
            2
            3
                   mse_total_list = []
            4
            5
                   for i in range(n runs):
            6
            7
                       x_best = hill_climbing(input_data, t, input_data[0], parameters
            8
            9
                       x = predator_prey_integration(t,initial_conditions,x_best)
           10
                       mse\_prey = MSE(data[:,1],x[:,0])
           11
           12
                       mse_predator = MSE(data[:,2],x[:,1])
           13
                       mse_total = mse_prey + mse_predator
           14
                       mse_total_list.append(mse_total) #Add total MSE for this simula
           15
           16
           17
                   return mse_total_list
           18
                                                                                      Þ
 In [32]:
              mean_mse_hill_climbing = np.mean(mse_total_list)
            2
               std_mse_hill_climbing = np.std(mse_total_list)
            3
              print("Average MSE = " + str(mean_mse_hill_climbing))
            4
               print("Standard deviation of MSE = " + str(std_mse_hill_climbing))
          Average MSE = 7.020538795747848
```

Comparison of Optimisation Algorithms

Standard deviation of MSE = 0.525718488225756

MSE objective function

```
In [228]:
            1 | input data = data[:,1:3]
            2 initial_conditions = [input_data[0][0], input_data[0][1]]
            3 | t = data[:,0]
            4 #Taking random draw for initial parameters (initial guess)
            5 alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7 delta = np.random.uniform(0,1)
            8 gamma = np.random.uniform(0,1)
            9
               parameters = [alpha, beta, delta, gamma]
           10 parameters = [alpha, beta, delta, gamma]
           11
           12 | initial_temp = 20
           13 | cooling_constant = 0.10
           14
           15 | mse_total_list_annealing = multiple_runs_annealing(initial_temp,cooling
           16 | mean_mse_annealing = np.mean(mse_total_list_annealing)
           17
              std_mse_annealing = np.std(mse_total_list_annealing)
           18
```

/var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489 9.py:38: RuntimeWarning: overflow encountered in exp acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489 9.py:38: RuntimeWarning: overflow encountered in scalar divide acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489 9.py:38: RuntimeWarning: divide by zero encountered in scalar divide acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /Users/alex_1/anaconda3/lib/python3.11/site-packages/scipy/integrate/_odep ack_py.py:248: ODEintWarning: Excess work done on this call (perhaps wrong Dfun type). Run with full_output = 1 to get quantitative information. warnings.warn(warning msg, ODEintWarning)

```
In [229]:
            1 input data = data[:,1:3]
            2 initial_conditions = [input_data[0][0], input_data[0][1]]
            3 t = data[:,0]
            4 #Taking random draw for initial parameters (initial guess)
            5 | alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7
              delta = np.random.uniform(0,1)
             gamma = np.random.uniform(0,1)
            9
              parameters = [alpha, beta, delta, gamma]
           10
           11
           12 | mse_total_list_hill = multiple_runs_hill_climbing(input_data, t, initia
           13 | mean_mse_hill_climbing = np.mean(mse_total_list_hill)
              std mse hill climbing = np.std(mse total list hill)
```

MAE objective function

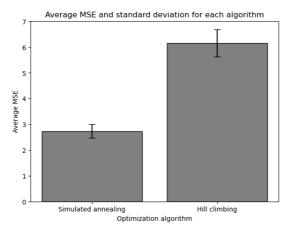
```
In [230]:
            1 | input data = data[:,1:3]
            2 initial_conditions = [input_data[0][0], input_data[0][1]]
            3 | t = data[:,0]
            4 #Taking random draw for initial parameters (initial guess)
            5 alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7 delta = np.random.uniform(0,1)
            8 gamma = np.random.uniform(0,1)
            9
               parameters = [alpha, beta, delta, gamma]
           10 parameters = [alpha, beta, delta, gamma]
           11
           12 | initial_temp = 20
           13 | cooling_constant = 0.10
           14
           15 | mae_total_list_annealing = multiple_runs_annealing(initial_temp,cooling
           16 | mean_mae_annealing = np.mean(mae_total_list_annealing)
              std_mae_annealing = np.std(mae_total_list_annealing)
           17
```

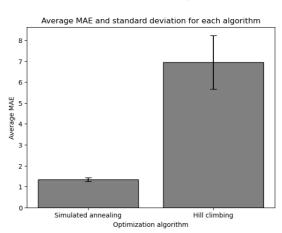
/var/folders/rc/tn2ys5g55157vhhlmn0 vfwr0000gg/T/ipykernel 70535/385309489 9.py:38: RuntimeWarning: overflow encountered in exp acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489 9.py:38: RuntimeWarning: overflow encountered in scalar divide acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability /var/folders/rc/tn2ys5g55157vhhlmn0_vfwr0000gq/T/ipykernel_70535/385309489 9.py:38: RuntimeWarning: divide by zero encountered in scalar divide acceptance_probability = min(np.exp(-(delta/temp)),1)#Calculate acceptan ce probability

```
In [231]:
            1 input data = data[:,1:3]
            2 initial_conditions = [input_data[0][0], input_data[0][1]]
            3 | t = data[:,0]
            4 #Taking random draw for initial parameters (initial quess)
            5 | alpha = np.random.uniform(0,1)
            6 beta = np.random.uniform(0,1)
            7 delta = np.random.uniform(0,1)
              gamma = np.random.uniform(0,1)
            9
              parameters = [alpha, beta, delta, gamma]
           10
           11
           12 mae_total_list_hill = multiple_runs_hill_climbing(input_data, t, initia
           13 | mean mae hill climbing = np.mean(mae total list hill)
              std mae hill climbing = np.std(mae total list hill)
```

```
In [237]:
               ##PLOTTING
            1
            2
            3
               fig, axes = plt.subplots(1, 2, figsize=(15, 5))
            4
            5
               x = ["Simulated annealing", "Hill climbing"]
            6
            7
               mse_list = [mean_mse_annealing,mean_mse_hill_climbing] #Stores the mear
               std_mse_list = [std_mse_annealing,std_mse_hill_climbing]
            8
            9
               mae_list = [mean_mae_annealing,mean_mae_hill_climbing] #Stores the mear
           10
               std_mae_list = [std_mae_annealing,std_mae_hill_climbing]
           11
           12
               axes[0].bar(x, mse_list, yerr = std_mse_list,color = "grey",ec = "black")
           13
               axes[1].bar(x, mae_list, yerr = std_mae_list,color = "grey",ec = "black")
           14
           15
           16
           17
           18
              axes[0].set_xlabel("Optimization algorithm")
           19
               axes[1].set_xlabel("Optimization algorithm")
           20
               axes[0].set_ylabel("Average MSE")
           21
           22
              axes[1].set_ylabel("Average MAE")
           23
           24
               axes[0].set_title("Average MSE and standard deviation for each algorith")
           25
               axes[1].set_title("Average MAE and standard deviation for each algorith")
           26
           27
           28
           29 #plt.bar(x, mse_list, yerr = std_list,color = "grey",ec = "black", ecol
           30 #plt.title("Bar plot comparing the means and errors of the MSE")
           31 #plt.ylabel("MSE")
           32 #plt.xlabel("Optimization algorithm")
```

Out[237]: Text(0.5, 1.0, 'Average MAE and standard deviation for each algorithm')





```
In [ ]:
In [ ]:
           1
```