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In [1]: import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.linear_model import SGDRegressor
        import warnings
        warnings.filterwarnings('ignore')
In [2]: import scipy as sp
        import scipy.optimize
        def angles in ellipse(num,a,b):
            assert(num > 0)
            assert(a < b)</pre>
            angles = 2 * np.pi * np.arange(num) / num
            if a != b:
                e = (1.0 - a ** 2.0 / b ** 2.0) ** 0.5
                tot size = sp.special.ellipeinc(2.0 * np.pi, e)
                arc size = tot size / num
                arcs = np.arange(num) * arc size
                res = sp.optimize.root(
                    lambda x: (sp.special.ellipeinc(x, e) - arcs), angles)
                angles = res.x
            return angles
In [3]: a = 2
         b = 9
        n = 50
        phi = angles in ellipse(n, a, b)
        e = (1.0 - a ** 2.0 / b ** 2.0) ** 0.5
        arcs = sp.special.ellipeinc(phi, e)
        fig = plt.figure()
        ax = fig.gca()
        ax.axes.set aspect('equal')
        ax.scatter(b * np.sin(phi), a * np.cos(phi))
         plt.show()
```

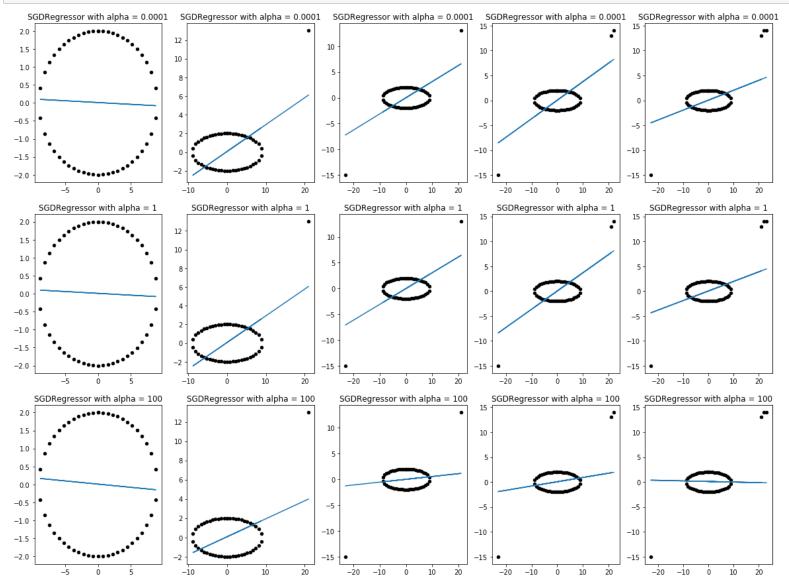
0

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-7.5 -5.0 -2.5 0.0 2.5 5.0 7.5
In [4]: X= b * np.sin(phi)
        Y= a * np.cos(phi)
In [5]: def draw line(coef, intercept, X):
            y pred = X * coef + intercept
            plt.plot(X,y pred)
In [6]: def applyRegressor(alpha range, outliers, X, Y):
            plot ctr = 0
            plt.figure(figsize=(20,15))
            for alpha in alpha range:
                x with outlier = X
                y with outlier = Y
                for outlier in outliers:
                    plot ctr += 1
                    x with outlier = np.append(x with outlier, outlier[0])
                    x with outlier = x with outlier.reshape(-1, 1)
                    y with outlier = np.append(y with outlier, outlier[1])
                    y with outlier = y with outlier.reshape(-1, 1)
                    clf = SGDRegressor(loss='squared loss', alpha=alpha, eta0=0.001, learning rate='constan
        t', random state=0)
                    clf.fit(x_with_outlier, y_with_outlier)
                    #print(clf.coef)
                    #print(clf.intercept )
                    plt.subplot(len(alpha range), len(outliers), plot ctr)
                    plt.scatter(x with outlier, y with outlier, s=20, color='black')
                    \#new x = np.array([np.min(x with outlier), np.max(x with outlier)])
                    \#new y = np.array([np.min(y with outlier), np.max(y with outlier)])
```

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draw_line(clf.coef_, clf.intercept_, x_with_outlier)
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plt.title('SGDRegressor with alpha = ' + str(alpha))

In [7]: alpha_range = [0.0001, 1, 100]
 outliers = [(0,2),(21, 13), (-23, -15), (22,14), (23, 14)]
 applyRegressor(alpha_range, outliers, X, Y)



Observation:

Here we can see that after train model with alpha = [0.0001, 1, 100] and outliers = [(0,2),(21, 13), (-23, -15), (22,14), (23, 14)] the best fit line I got is with al pha=100 which is in the last subplot.