Lab 11 - Curve Fitting

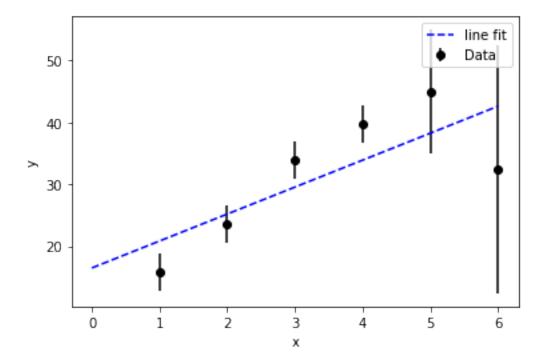
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```
[1]: %pylab inline
```

Populating the interactive namespace from numpy and matplotlib

```
[10]: from scipy import optimize
      #define fitting function to be a linear function
      def line_func(x, a, b):
          return x*a+b
      #Sample data for straight line fine
      x_{data} = np.array([1.0,2.0,3.0,4.0,5.0,6.0])
      y_{data} = np.array([15.9, 23.6, 33.9, 39.7, 45.0, 32.4])
      y_{unc} = np.array([3.0,3.0,3.0,3.0,10.0,20.0])
      #Plotting sample data
      plt.errorbar(x_data, y_data,yerr=y_unc,fmt="ko",label="Data")
      #initial quesses for parameter values and calculating best fit curve for x_data_
      \rightarrow and y_data
      guess_a = 1.0
      guess_b = 0.0
      par, cov = optimize.curve_fit(line_func, x_data, y_data, p0=[guess_a,guess_b])
      #Fitted values of a and b
      fit_a = par[0]
      fit_b = par[1]
      print("best fit value of a: ", fit_a)
      print("best fit value of b: ", fit_b)
      #Plotting the best fit line
      xf = np.linspace(0.0,6.0,100)
      yf = line_func(xf,fit_a,fit_b)
      plt.plot(xf, yf,"b--",label="line fit")
      plt.xlabel("x")
      plt.ylabel("y")
      plt.legend()
```

best fit value of a: 4.3571428997566874 best fit value of b: 16.499999850851594

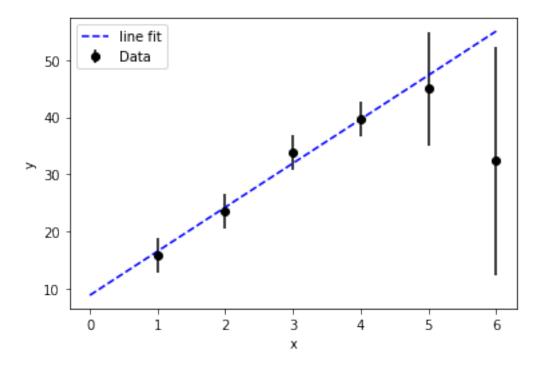


Jupyter Notebook 11.1: Plot data including error bars and best-fit function to obtain

a result like that of Fig 11.2 in lab manual version 6. Function is biased towards poorly measured points.

```
#Plotting sample data
plt.errorbar(x_data, y_data,yerr=y_unc,fmt="ko",label="Data")
#initial guesses for parameter values and calculating best fit curve for x_{-}data_{\sqcup}
\rightarrow and y_data
guess_a = 1.0
guess_b = 0.0
#Fitted values of a and b
fit_a2 = par2[0]
fit_b2 = par2[1]
print("best fit value of a: ", fit_a)
print("best fit value of b: ", fit_b)
#Plotting the best fit line
xf = np.linspace(0.0,6.0,100)
yf = line_func(xf,fit_a2,fit_b2)
plt.plot(xf, yf,"b--",label="line fit")
plt.xlabel("x")
plt.ylabel("y")
plt.legend()
plt.show()
print("""Jupyter Notebook 11.2: Plotting the correct uncertainty y_unc to⊔
→produce new plot with data and fit - result
is a fit with no more bias
""")
```

best fit value of a: 4.3571428997566874 best fit value of b: 16.499999850851594



Jupyter Notebook 11.2: Plotting the correct uncertainty y_unc to produce new plot with data and fit - result is a fit with no more bias

```
[4]: def constant_func(x,a):
         return a
     #choose y2\_data independent of x-values from Gaussian with a mean of 50 and
     →uncertainty on the y-values to 10
     x2_{data} = np.arange(100)
     y2_data = np.random.normal(50.0, 10.0, size=100)
     #Best fit constant values
     guess_a = 0.0
     par3, cov3 = optimize.curve_fit(constant_func, x2_data, y2_data,__
      →p0=[guess_a],absolute_sigma=True)
     #Best fit parameter and it's uncertainty
     unc = np.sqrt(np.diag(cov3))
     fit_a = par3[0]
     unc_a = unc[0]
     print("""Jupyter Notebook 11.3: Set uncertainty on the y values to 10 and ⊔
      \hookrightarrowabsolute_sigma = True in the fit. Record the
```

```
""")

#Printing the results
print("mean of y data: ", np.mean(y2_data))
print("fitted constant: ", fit_a)
print("uncertainty: ", unc_a)
```

Jupyter Notebook 11.3: Set uncertainty on the y values to 10 and absolute_sigma = True in the fit. Record the uncertainty.

mean of y data: 50.13170251076923 fitted constant: 50.13170251085707 uncertainty: 0.1000000001752797

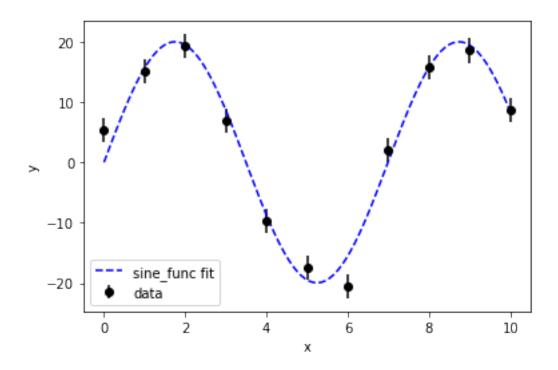
```
[5]: #fit y3_data to a constant value a:
     def constant_func(x,a):
        return a
     #choose y3_data independent of x-values from Gaussian with a mean of 50 and \Box
     ⇔sigma of 10
     x3_{data} = np.arange(100)
     y3_data = np.random.normal(50.0, size=100)
     #Best fit constant value
     guess_a = 0.0
     par4, cov4 = optimize.curve_fit(constant_func, x3_data, y3_data,__
      →p0=[guess_a],absolute_sigma=True)
     #Best fit parameter and it's uncertainty
     unc = np.sqrt(np.diag(cov3))
     fit_a = par4[0]
     unc_a = unc[0]
     print("""Jupyter Notebook 11.4: Leave the uncertainties on the y values⊔
     →unspecified and absolute_sigma = True in the fit.
     """)
     #Printing the results
     print("mean of y data: ", np.mean(y3_data))
     print("fitted constant: ", fit_a)
     print("uncertainty: ", unc_a)
```

Jupyter Notebook 11.4: Leave the uncertainties on the y values unspecified and absolute_sigma = True in the fit.

mean of y data: 50.12609604415761 fitted constant: 50.126096044245415 uncertainty: 0.1000000001752797

```
[6]: def sine_func (x, a, b):
         return a*np.sin(b*x)
     #Sample data and setting y uncertainty to sigma = 2
     x_sample = np.array([0.0,1.0,2.0,3.0,4.0,5.0,6.0,7.0,8.0,9.0,10.0])
     y_{sample} = np.array([5.3,15.0,19.2,6.8,-9.7,-17.4,-20.5,2.1,15.7,18.5,8.6])
     y_unc_sine = 2*np.ones(11)
     #Plotting the sample data
     plt.errorbar(x_sample, y_sample,yerr=y_unc_sine,fmt="ko",label="data")
     #initial quesses for parameter values and calculating best fit curve for x_data_
      \rightarrow and y_data
     guess_a = 20.0
     guess_b = 2*pi/7
     par_sine, cov_sine = optimize.curve_fit(sine_func, x_sample, y_sample, u
      →p0=[guess_a,guess_b],absolute_sigma=True)
     #Fitted values of a and b
     fit_a_sine = par_sine[0]
     fit_b_sine = par_sine[1]
     print("best fit value of a: ", fit_a)
     print("best fit value of b: ", fit_b)
     #Plotting the best fit line
     xf_sine = np.linspace(0.0,10.0,100)
     yf_sine = fit_a_sine*np.sin(fit_b_sine*xf_sine)
     plt.plot(xf_sine, yf_sine,"b--",label="sine_func fit")
     plt.xlabel("x")
     plt.ylabel("y")
     plt.legend()
     plt.show()
     print("""Jupyter Notebook 11.5: Plot sample data listed in lab manual version 6⊔
      \rightarrowin section 11.4 with error bars and
     best-fit sine wave.
     """)
```

best fit value of a: 50.126096044245415 best fit value of b: 16.499999850851594



Jupyter Notebook 11.5: Plot sample data listed in lab manual version 6 in section 11.4 with error bars and best-fit sine wave.

[]: