

Uncertainty_plotting

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1 Plotting uncertainty

In this example we will go over plotting uncertainties in various ways: + y errorbars + x errorbars + x and y errorbars (no covariance) + x and y error-ellipse (covariance)

1.1 Packages being used

- `matplotlib`: all the plotting
- `astropy`: read in the data table
- `numpy` and `scipy`: convert cov matrix to ellipse params

1.2 Relevant documentation

- `matplotlib`: http://matplotlib.org/1.5.1/api/pyplot_api.html#matplotlib.pyplot.errorbar

```
In [1]: from astropy.table import Table
import scipy.linalg as sl
import numpy as np
from matplotlib import pyplot as plt
from matplotlib.patches import Ellipse
import mpl_style
%matplotlib inline
plt.style.use(mpl_style.style1)
```

Our data contains (x, y) positions with $1\text{-}\sigma$ uncertainties and covariance values:

```
In [2]: t = Table.read('data.csv', format='ascii.csv')
print(t)
```

ID	x	y	sy	sx	pxy
1	201	592	61	9	-0.84
2	244	401	25	4	0.31
3	47	583	38	11	0.64
4	287	402	15	7	-0.27
5	203	495	21	5	-0.33
6	58	173	15	9	0.67
7	202	479	27	4	-0.02
8	202	504	14	4	-0.05
9	198	510	30	11	-0.84
10	158	416	16	7	-0.69
11	165	393	14	5	0.3
12	201	442	25	5	-0.46
13	157	317	52	5	-0.03

```

14 131 311 16 6 0.5
15 166 400 34 6 0.73
16 160 337 31 5 -0.52
17 186 423 42 9 0.9
18 125 334 26 8 0.4
19 218 533 16 6 -0.78
20 146 344 22 5 -0.56

```

Note the full covariance matrix for each data point is:
$$\begin{bmatrix} \sigma_x^2 & \rho_{xy}\sigma_x\sigma_y \\ \rho_{xy}\sigma_x\sigma_y & \sigma_y^2 \end{bmatrix}$$

1.3 y-uncertainties or x-uncertainties only

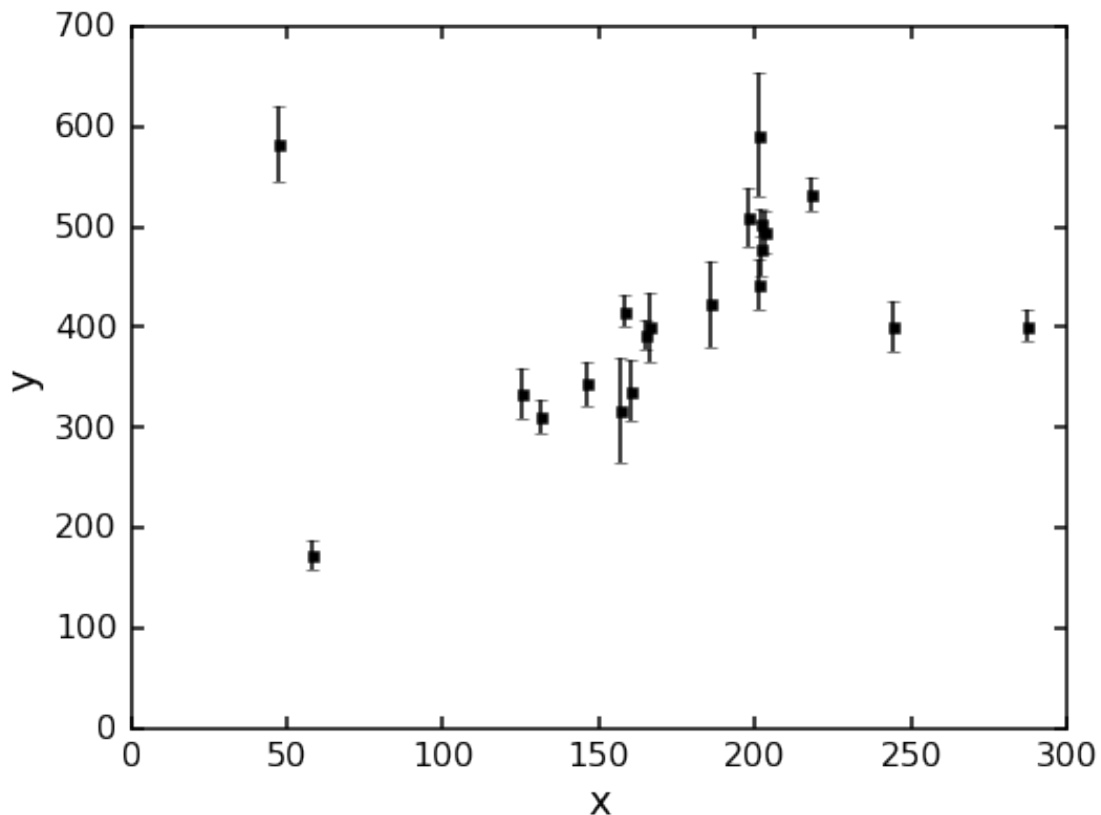
The most common type of data you will work with will only have (significant) uncertainties in one direction. In this case it is very easy to plot using `errorbar`:

```

In [3]: plt.figure(1)
        plt.errorbar(t['x'], t['y'], yerr=t['sy'], ls='None', mfc='k', ms=5, marker='s', ecolor='k')
        plt.xlabel('x')
        plt.ylabel('y')
        plt.ylim(0, 700)

```

```
Out[3]: (0, 700)
```



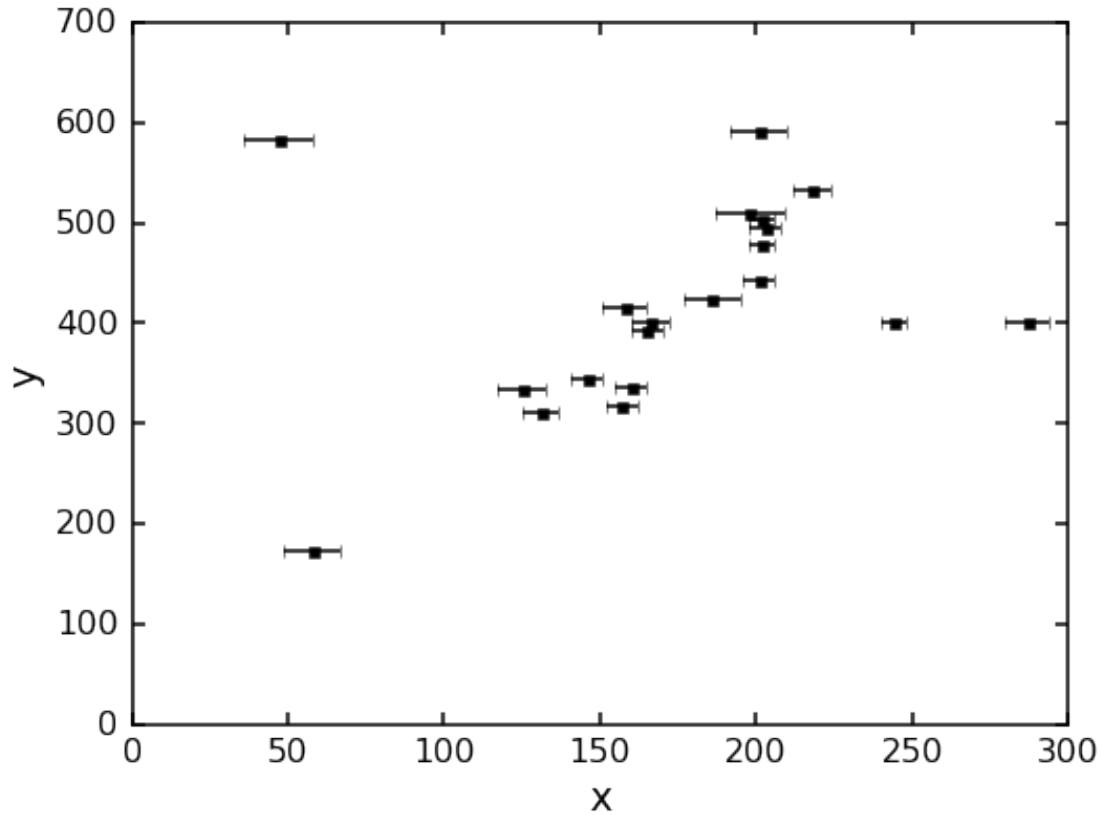
```

In [4]: plt.figure(2)
        plt.errorbar(t['x'], t['y'], xerr=t['sx'], ls='None', mfc='k', ms=5, marker='s', ecolor='k')

```

```
plt.xlabel('x')
plt.ylabel('y')
plt.ylim(0, 700)
```

Out[4]: (0, 700)

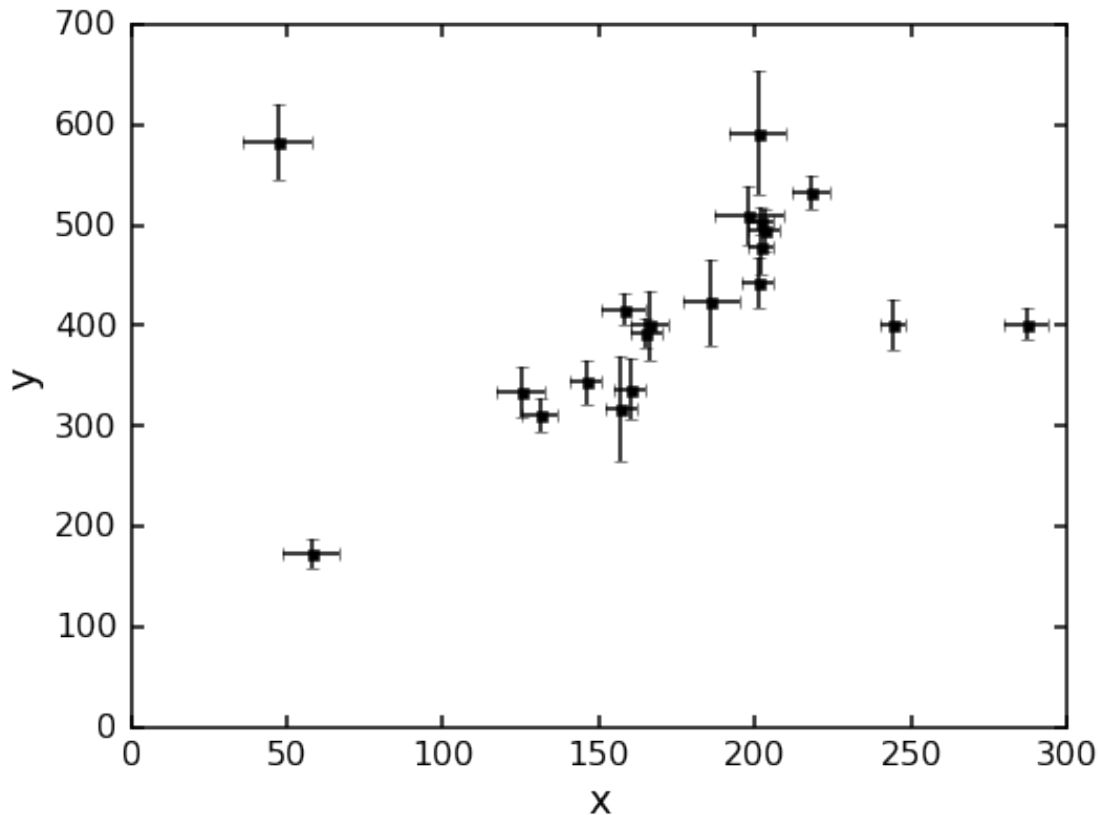


1.4 Uncertainties in both x and y with no cov

If your data has no cov you can still use `errorbar`:

```
In [5]: plt.figure(3)
plt.errorbar(t['x'], t['y'], yerr=t['sy'], xerr=t['sx'], ls='None', mfc='k', ms=5, marker='s', c=
plt.xlabel('x')
plt.ylabel('y')
plt.ylim(0, 700)
```

Out[5]: (0, 700)



1.5 Uncertainties in boty x and y with cov

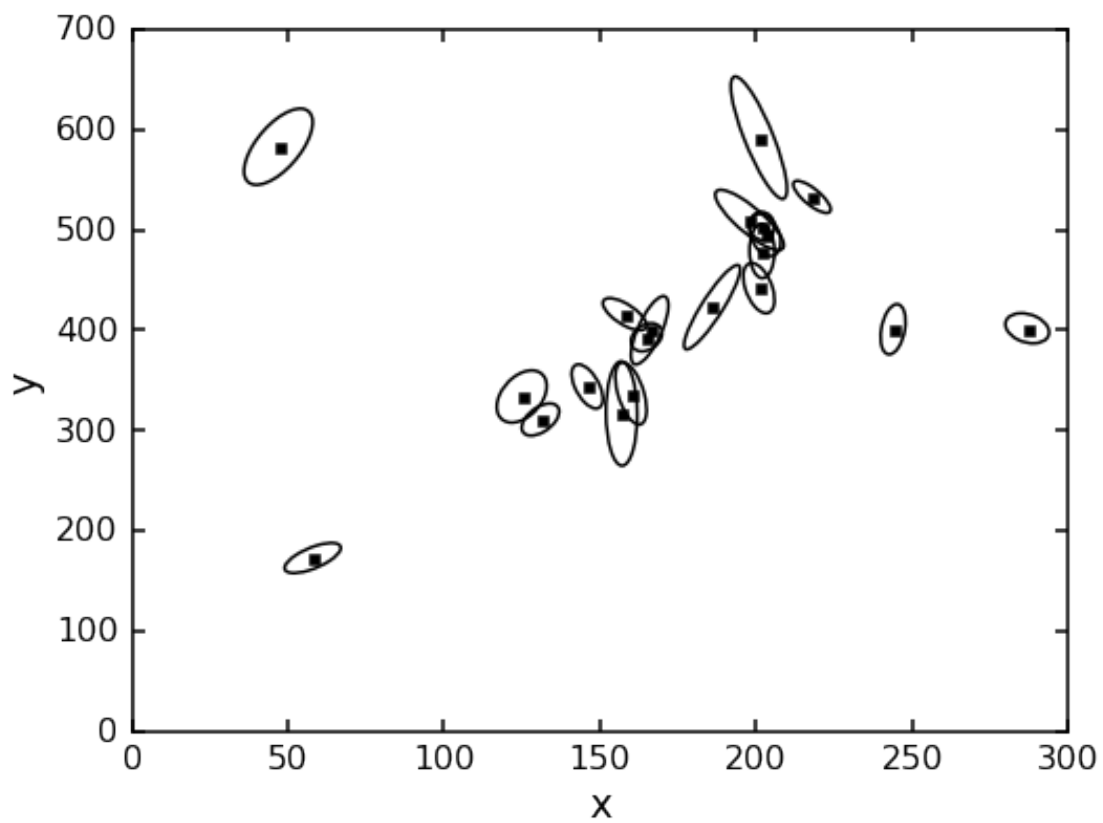
If your data does have cov you should plot a $1\text{-}\sigma$ ellipse around each point. There is no built in function to do this, so we will have to write our own. We will start by writing a function to turn a cov matrix into the parameters for an ellipse and draw it on a figure.

```
In [6]: def cov_to_ellipse(cov, pos, **kwargs):
    eigvec,eigval,V = sl.svd(cov,full_matrices=False)
    # the angle the first eigenvector makes with the x-axis
    theta = np.degrees(np.arctan2(eigvec[1, 0], eigvec[0, 0]))
    # full width and height of ellipse, not radius
    # the eigenvalues are the variance along the eigenvectors
    width, height = 2 * np.sqrt(eigval)
    return Ellipse(xy=pos, width=width, height=height, angle=theta, **kwargs)

def plot_ellipse(t, ax=None, **kwargs):
    if ax is None:
        ax = plt.gca()
    for row in t:
        cov = np.array([[row['sx']**2, row['pxy'] * row['sx'] * row['sy']], [row['pxy'] * row['sx'] * row['sy'], row['sy']**2]])
        ellip = cov_to_ellipse(cov, [row['x'], row['y']], **kwargs)
        ax.add_artist(ellip)

plt.figure(4)
```

```
plt.plot(t['x'], t['y'], 's', mfc='k', ms=5)
plot_ellipse(t, fc='none', lw=1.5)
plt.xlabel('x')
plt.ylabel('y')
plt.ylim(0, 700)
plt.draw()
```



In []: