

# 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/2.0.2/api/pyplot\\_api.html#matplotlib.pyplot.errorbar](http://matplotlib.org/2.0.2/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

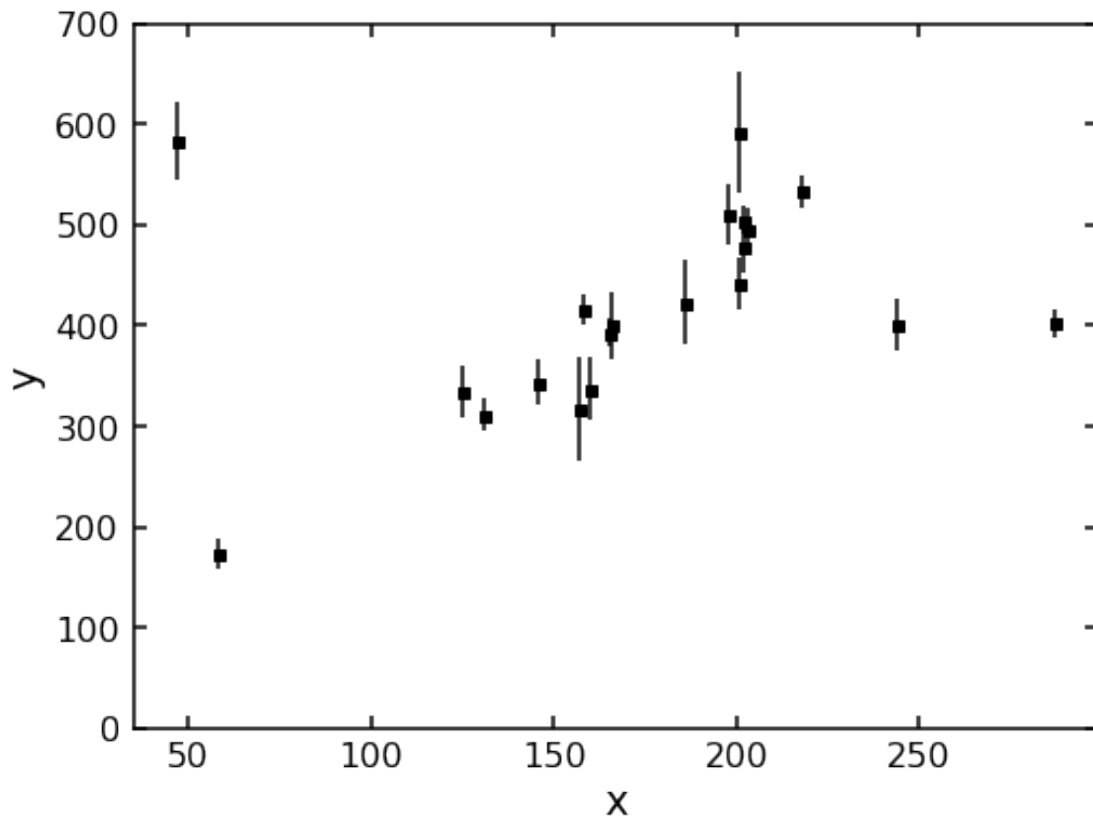
**Note** the full covariance matrix for each data point is: 
$$\begin{bmatrix} \sigma_x^2 & \rho_{xy}\sigma_x\sigma_y \\ \rho_{xy}\sigma_x\sigma_y & \sigma_y^2 \end{bmatrix}$$

### 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`:

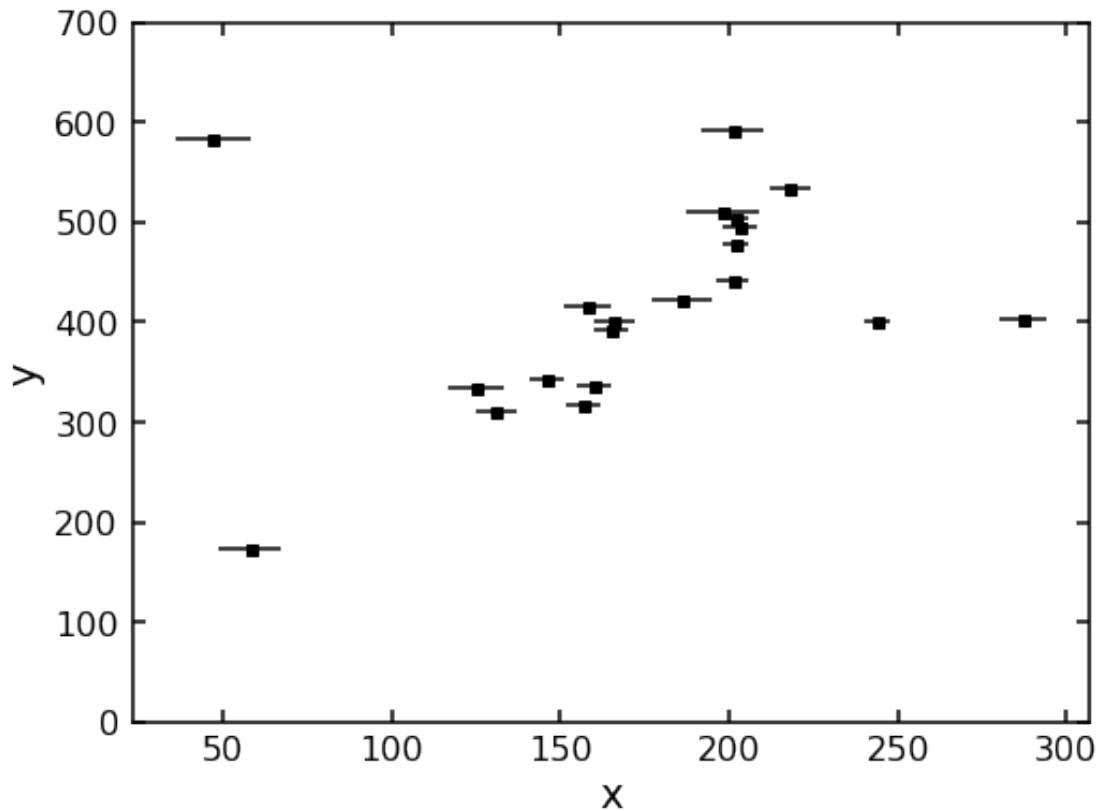
```
In [3]: plt.figure(1)
plt.errorbar(
    t['x'],
    t['y'],
    yerr=t['sy'],
    ls='None',
    mfc='k',
    mec='k',
    ms=5,
    marker='s',
    ecolor='k'
)
plt.xlabel('x')
plt.ylabel('y')
plt.ylim(0, 700)
```

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



```
In [4]: plt.figure(2)
plt.errorbar(
    t['x'],
    t['y'],
    xerr=t['sx'],
    ls='None',
    mfc='k',
    mec='k',
    ms=5,
    marker='s',
    ecolor='k'
)
plt.xlabel('x')
plt.ylabel('y')
plt.ylim(0, 700)
```

```
Out[4]: (0, 700)
```

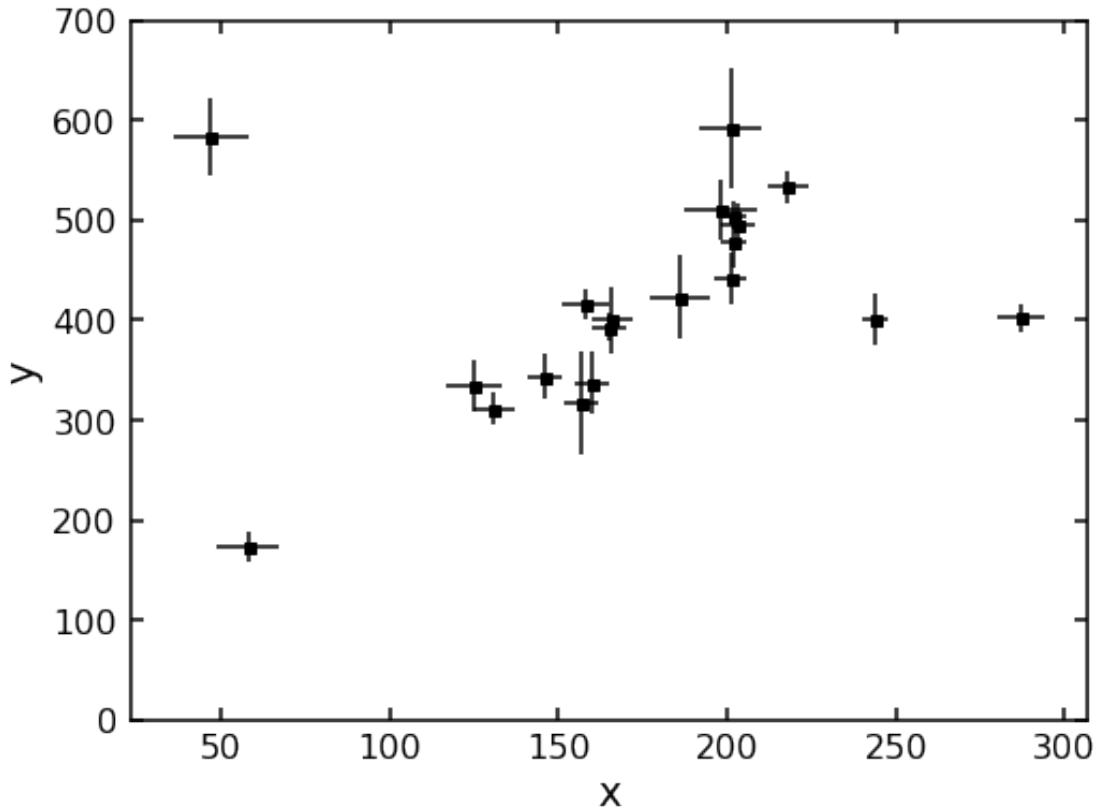


## 1.4 Uncertainties in both x and y with no cov

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',
    mec='k',
    ms=5,
    marker='s',
    ecolor='k'
)
plt.xlabel('x')
plt.ylabel('y')
plt.ylim(0, 700)
```

Out [5]: (0, 700)

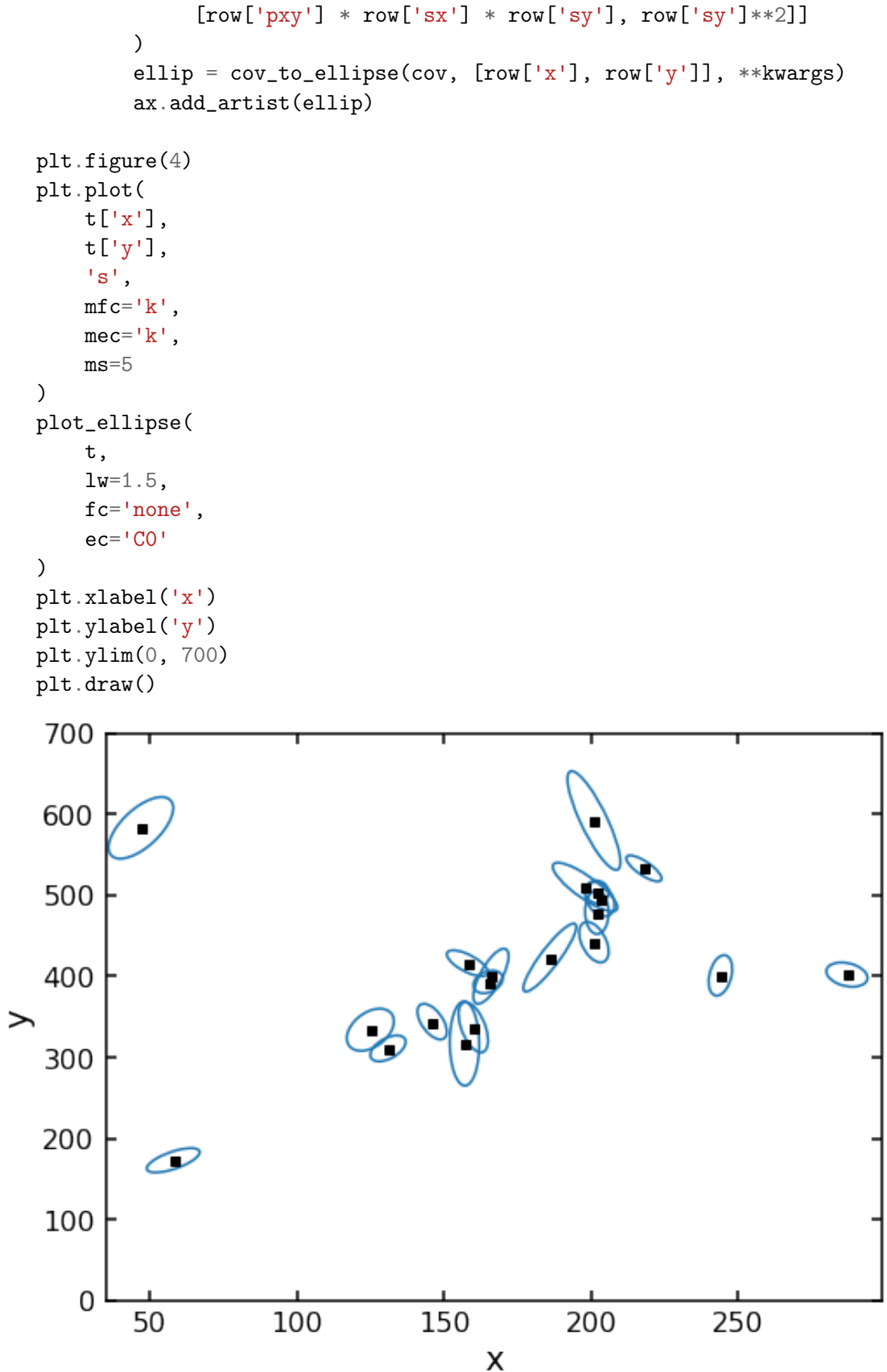


### 1.5 Uncertainties in both x and y with cov

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']],
```



```
In [ ]:
```