Condition number

```
In [1]:
#keep
import numpy as np
import numpy.linalg as la
import matplotlib.pyplot as pt
%matplotlib inline
Let's grab a 2 \times 2 matrix A:
In [2]:
#keep
if 0:
    np.random.seed(17)
    A = np.random.randn(2, 2)
else:
    A = np.array([[3, 0], [0,1]], dtype=np.float64)
Α
Out[2]:
array([[ 3., 0.],
       [ 0., 1.]])
And its inverse:
In [3]:
Ainv = la.inv(A)
Ainv
Out[3]:
array([[ 0.33333333, 0.
                                  ],
       [ 0.
```

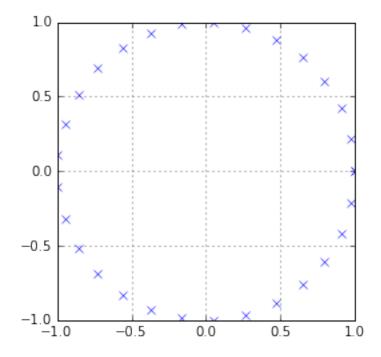
Now we would like to figure out where that matrix puts all the vectors with 2-norm 1.

]])

To do so, let's make an array of vectors with vectors with norm 1:

1.

```
In [4]:
```



Now apply \boldsymbol{A} to all those vectors...:

```
In [5]:
```

```
#keep
Axs = A.dot(xs)
Axs.shape
```

Out[5]:

(2, 30)

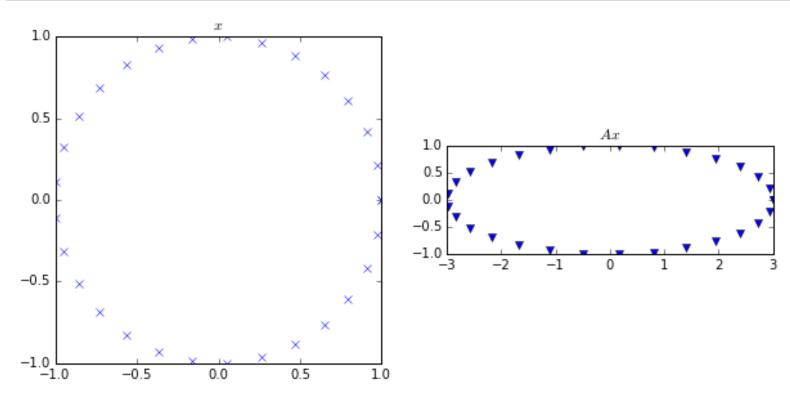
...and plot:

```
In [12]:
```

```
#keep
pt.figure(figsize=(10, 5))

pt.subplot(121)
pt.title("$x$")
pt.plot(xs[0], xs[1], "x")
pt.gca().set_aspect("equal")

pt.subplot(122)
pt.title("$Ax$")
pt.plot(Axs[0], Axs[1], "v")
pt.gca().set_aspect("equal")
```



Next, let's see what happens to small perturbations at each of the x and Ax points.

To that end, let's make an array ys of shape $2 imes N_p imes N_p$, where N_p is the number of points above.

```
In [13]:
```

```
#keep
# ys has axes: XY x Npoints x Npoints

perturbation_size = 0.1
ys = perturbation_size * xs.reshape(2, -1, 1) + xs.reshape(2, 1, -1)

Ays = np.tensordot(A, ys, axes=1)
Ays.shape
```

```
Out[13]:
(2, 30, 30)
```

Side note: What does the argument -1 to reshape do?

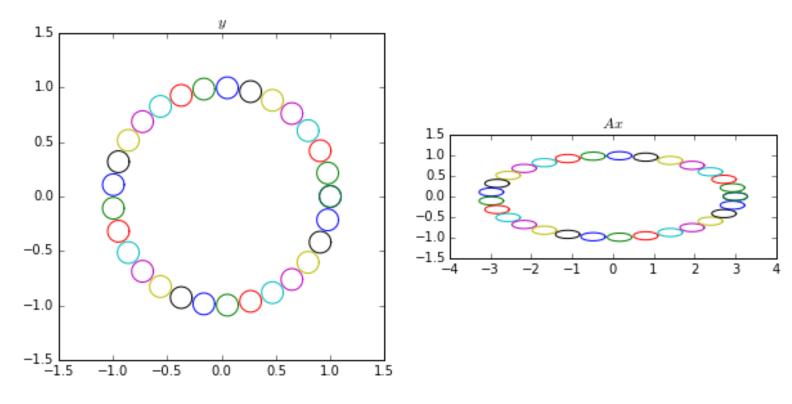
Let's plot what we've just made

```
In [14]:
```

```
#keep
pt.figure(figsize=(10, 5))

pt.subplot(121)
pt.title("$y$")
pt.plot(ys[0], ys[1])
pt.gca().set_aspect("equal")

pt.subplot(122)
pt.title("$Ax$")
pt.plot(Ays[0], Ays[1])
pt.gca().set_aspect("equal")
```



Let's compare this with $\|A\|$:

```
In [15]:
```

```
#keep
norm = la.norm(A, 2)
print(norm)

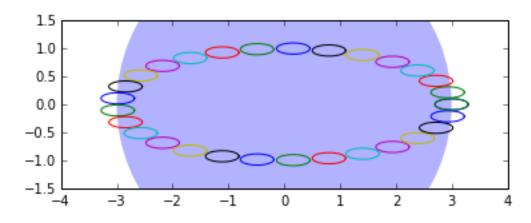
pt.plot(Ays[0], Ays[1])

ax = pt.gca()
ax.set_aspect("equal")
ax.add_artist(pt.Circle([0, 0], norm, alpha=0.3, lw=0))
```

3.0

Out[15]:

<matplotlib.patches.Circle at 0x10884cb90>



What we want now is a circle around each of the Ax that says,

"Because of the Δx variation, b is at most going to wiggle by this much, i.e. Δb will be at most this big."

Now we want a κ with $\frac{\|\Delta b\|}{\|b\|} \leq \kappa \frac{\|\Delta x\|}{\|x\|}$.

Assume $\|x\|=1$. Equivalent: $\|\Delta b\|\leq \kappa \|\Delta x\|\|b\|$.

Which κ does the job?

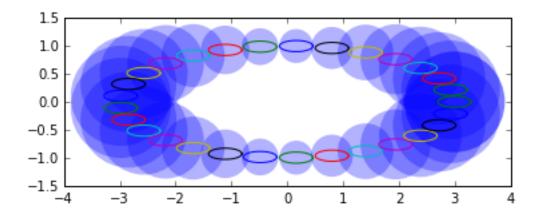
In [16]:

kappa = la.norm(A, 2)*la.norm(Ainv, 2)

```
In [17]:
```

```
#keep
pt.plot(Ays[0], Ays[1])

ax = pt.gca()
ax.set_aspect("equal")
for i in range(Ays.shape[2]):
   b = Axs[:, i]
   norm_delta_y = kappa * perturbation_size * la.norm(b)
   ax.add_artist(pt.Circle(b, norm_delta_y, alpha=0.3, lw=0))
```



In []:

In []:

In []: