Given's rotations and the QR decomposition

In this lecture we introduce Given's rotations as a way to calculate the QR decomposition of a matrix.

We derived an algorithm for calculating the LU Decomposition by applying lower triangular operations on 2 rows at time: thus it came down to the matrix

$$L = \begin{pmatrix} 1 & \\ -\frac{b}{a} & 1 \end{pmatrix}$$

so that

$$L\begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} a \\ 0 \end{pmatrix}$$

```
In [57]:
```

```
x=rand(2)
a,b=x;
L=[1  0;
   -b/a 1]
L*x
```

```
Out[57]:
```

```
2-element Array{Float64,1}:
    0.398033
    1.11022e-16
```

This had the issue that it degenerated as *a* becomes small.

As an alternative, we will *rotate* the vector to introduce a zero. Recall that the angle of a point (a, b) with the origin is given by $a \tan \frac{b}{a}$, thus we want to rotate by $\theta = -a \tan \frac{b}{a}$:

```
In [59]:
```

```
 \theta = -\operatorname{atan2}(b,a) \quad \# \; equivalent \; to \quad \theta = \operatorname{atan}(b/a)   Q = [\cos(\theta) - \sin(\theta); \\ \sin(\theta) \cos(\theta)]   Q*x
```

```
Out[59]:

2-element Array{Float64,1}:

0.959959

5.55112e-17
```

Unlike lower triangularizations, this works well even for points on the y-axis:

```
In [61]:
```

```
x=[0.,2.]
a,b=x

θ=-atan2(b,a)
Q=[cos(θ) -sin(θ);
sin(θ) cos(θ)]

Q*x
```

```
Out[61]:
2-element Array{Float64,1}:
2.0
1.22465e-16
```

Effect of error

On a computer, we don't know exactly where a point is: every point can have a small ϵ of error. Thus to understand the robustness of an algorithm, we need to understand what happens to balls of radius ϵ around where we think of the point. This can be used to demonstrate why rotations are preferred to lower triangular operations.

In the following, we design a function that plots a circle around a point [a,b] of size ϵ , both before and after a matrix L is applied:

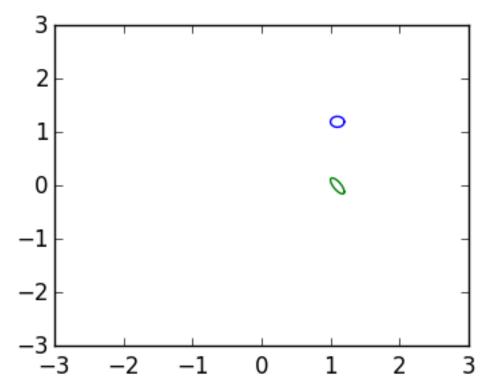
In [62]:

```
Out[62]:
plotmat (generic function with 1 method)
```

Here we demonstrate that the effect of L is to stretch the circles: our error can be amplified:

```
In [64]:
```

```
a,b=[1.1,1.2]
L=[1 0;
-b/a 1]
ε=0.1
plotmat(a,b,ε,L)
```



```
Out[64]:
```

4-element Array{Int64,1}:

-3

3

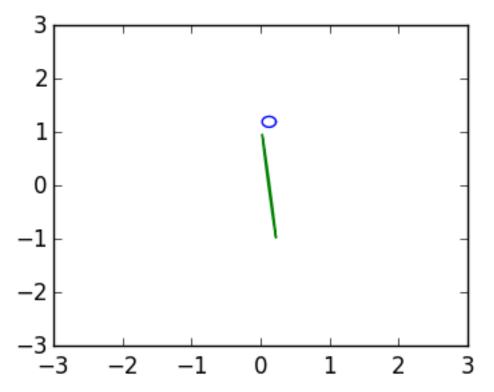
-3

3

As a becomes small, this error amplification becomes greater: in the following, we go from knowing the true point with accuracy 0.1 to only knowing it with about accuracy 1:

```
In [66]:
```

```
a,b=[0.125,1.2]
L=[1    0;
    -b/a 1]
ε=0.1
plotmat(a,b,ε,L)
```



Out[66]:

4-element Array{Int64,1}:
-3
3
-3
3

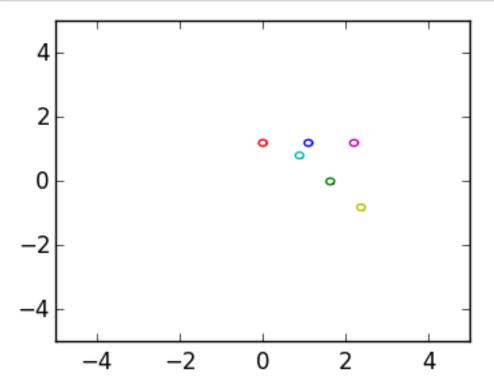
Rotations perform much better: the circles are only rotated, and are not magnified at all:

```
In [67]:
a,b=[1.1,1.2]

θ=-atan2(b,a)
Q=[cos(θ) -sin(θ);
    sin(θ) cos(θ)]

ε=0.1

plotmat(a,b,ε,Q)
plotmat(0,b,ε,Q)
```



```
Out[67]:
4-element Array{Int64,1}:
   -5
   5
   -5
   5
```

plotmat($2a,b,\varepsilon,Q$) axis([-5,5,-5,5])

Simpler definition

We can construct a simpler expression for Q as

$$\frac{1}{\sqrt{a^2 + b^2}} \begin{pmatrix} a & b \\ -b & a \end{pmatrix}$$

```
In [52]:

Q=[cos(θ) -sin(θ);
sin(θ) cos(θ)]

Out[52]:

2x2 Array{Float64,2}:
    0.675725   0.737154
    -0.737154   0.675725

In [54]:

[a b; -b a]/sqrt(a^2+b^2)

Out[54]:

2x2 Array{Float64,2}:
    0.675725   0.737154
    -0.737154   0.675725
```