

Lecture 6: Linear algebra

In this lecture we introduce linear algebra. We first consider the problem of calculating the equilibrium of a system of four springs and three balls, affixed to walls at 0 and 4, where 0 is at the top and 4 is at the bottom.

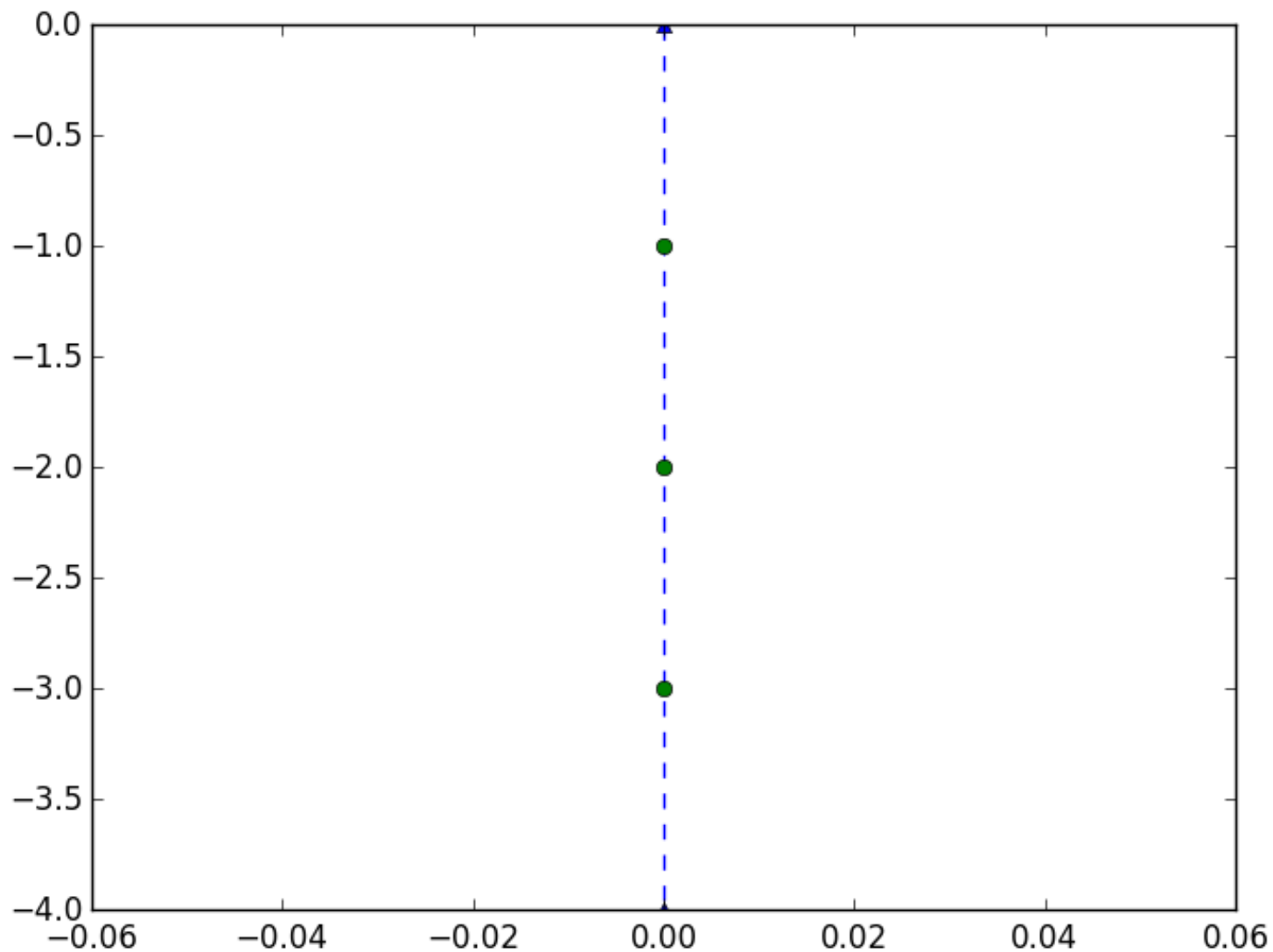
Here is a plot of our system, where the green circles represent the balls and the blue dotted lines the spring:

In [76]:

```
using PyPlot # load the plot command

n=3
plot([0.,0.],[0.,-4.];marker="^",linestyle="--") # plot triangles to represent the
                                                    # anchors at 0 and 4

p=1:n      # The resting position of the balls is equi spaced, so at 1, 2, and 3
plot(zeros(n),-p;marker="o",linestyle="")        # plot the three balls
```



Out[76]:

```
1-element Array{Any,1}:
 PyObject <matplotlib.lines.Line2D object at 0x3152dfdd0>
```

Denote the displacement of the balls by u_1, u_2, u_3 and the elongation of the springs by e_1, e_2, e_3, e_4 . Then we have

$$u_1 = e_1, u_2 - u_1 = e_2, u_3 - u_2 = e_3, -u_3 = e_4$$

Or in matrix form

$$A\mathbf{u} = \mathbf{e} \quad \text{for} \quad A = \begin{pmatrix} 1 & & \\ -1 & 1 & \\ & -1 & 1 \\ & & -1 \end{pmatrix}, \mathbf{u} = \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix}, \mathbf{e} = \begin{pmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{pmatrix}$$

Denote the forces of the springs by y_1, y_2, y_3, y_4 . As the spring becomes more elongated, the force becomes stronger, which we model by Hook's law: $y_k = c_k e_k$ where c_k are the stiffness of each spring. This also can be written in matrix form:

$$C\mathbf{e} = \mathbf{y} \quad \text{for} \quad C = \begin{pmatrix} c_1 & & & \\ & c_2 & & \\ & & c_3 & \\ & & & c_4 \end{pmatrix}$$

We finally have the external forces f_1, f_2, f_3 pulling down on the balls, these could be gravity or something else. The external force on each ball has to balance with the forces from the spring, giving us:

$$f_1 + y_2 = y_1, f_2 + y_3 = y_2, f_3 + y_4 = y_3$$

Or in matrix form

$$A^T \mathbf{y} = \mathbf{f} \quad \text{for} \quad \mathbf{f} = \begin{pmatrix} f_1 \\ f_2 \\ f_3 \end{pmatrix}$$

We thus want to find the displacement of each ball by solving:

$$A\mathbf{u} = \mathbf{e}, C\mathbf{e} = \mathbf{y}, A^T \mathbf{y} = \mathbf{f}$$

which is equivalent to

$$A^T C A \mathbf{u} = \mathbf{f}$$

In other words,

$$K\mathbf{u} = \mathbf{f}$$

for $K = A^T C A$.

We will use Julia to solve this equation and plot the results. To do this, we need to create Vectors and Matrices, do Matrix-Matrix multiplication, take transposes and solve linear systems.

Creating Vectors

The easiest way to create a vector is to use zeros to create a zero Vector and then modify its entries:

In [5]:

```
v=zeros(5)
v[2]=3.3423
v
```

Out[5]:

```
5-element Array{Float64,1}:
 0.0
 3.3423
 0.0
 0.0
 0.0
```

We can specify the type of the Vector by adding it as the first argument to zeros:

In [77]:

```
v=zeros{Int64,5}
v[2]=3
v
```

Out[77]:

```
5-element Array{Int64,1}:
 0
 3
 0
 0
 0
```

Note: we can't assign a Float to an integer vector:

In [78]:

```
v=zeros{Int64,5}
v[2]=3.5
```

```
LoadError: InexactError()
while loading In[78], in expression starting on line 2

in setindex! at array.jl:313
```

We can also create vectors with ones and rand:

In [12]:

```
ones(5)
```

Out[12]:

```
5-element Array{Float64,1}:  
 1.0  
 1.0  
 1.0  
 1.0  
 1.0
```

In [13]:

```
ones(Int64,5)
```

Out[13]:

```
5-element Array{Int64,1}:  
 1  
 1  
 1  
 1  
 1
```

In [79]:

```
rand(5)
```

Out[79]:

```
5-element Array{Float64,1}:  
 0.777089  
 0.75453  
 0.72074  
 0.94775  
 0.842468
```

In [15]:

```
rand(Int,5)
```

Out[15]:

```
5-element Array{Int64,1}:  
 -2711362773818306057  
  4463414055115793868  
 -4584630395923355567  
 -6681289071620779384  
 -6454145051917951527
```

We have already seen another way to create vectors directly:

In [16]:

```
[1,2,3,4]
```

Out[16]:

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

```
1
2
3
4
```

When the elements are of different types, they are mapped to a type that can represent every entry. For example, here we input a list of one `Float64` followed by three `Ints`, which are automatically converted to `Float64s`:

In [17]:

```
[1.0,2,3,4]
```

Out[17]:

```
4-element Array{Float64,1}:
```

```
1.0
2.0
3.0
4.0
```

In the event that the types cannot automatically be converted, it defaults to an `Any` vector. This is bad performance-wise so should be avoided.

In [18]:

```
[1.0,1,"1"]
```

Out[18]:

```
3-element Array{Any,1}:
```

```
1.0
1
"1"
```

We can also specify the type of the `Vector` explicitly by writing the desired type before the first bracket:

In [20]:

```
Float64[1,2,3,4]
```

Out[20]:

```
4-element Array{Float64,1}:  
 1.0  
 2.0  
 3.0  
 4.0
```

We can also create an array using brackets, a formula and a for command:

In [80]:

```
[k^2 for k=1:5]
```

Out[80]:

```
5-element Array{Int64,1}:  
 1  
 4  
 9  
16  
25
```

In [22]:

```
Float64[k^2 for k=1:5]
```

Out[22]:

```
5-element Array{Float64,1}:  
 1.0  
 4.0  
 9.0  
16.0  
25.0
```

Creating Matrices

Matrices are created similar to vectors, but by specifying two dimensions instead of one. Again, the simplest way is to zeros to create a matrix of all zeros:

In [23]:

```
zeros(5,5) # creates a 5x5 matrix of Float64 zeros
```

Out[23]:

```
5x5 Array{Float64,2}:  
 0.0  0.0  0.0  0.0  0.0  
 0.0  0.0  0.0  0.0  0.0  
 0.0  0.0  0.0  0.0  0.0  
 0.0  0.0  0.0  0.0  0.0  
 0.0  0.0  0.0  0.0  0.0
```

We can also have matrices of different types:

In [24]:

```
zeros(Int,5,5)
```

Out[24]:

```
5x5 Array{Int64,2}:  
 0  0  0  0  0  
 0  0  0  0  0  
 0  0  0  0  0  
 0  0  0  0  0  
 0  0  0  0  0
```

eye creates the identity matrix:

In [81]:

```
eye(5)
```

Out[81]:

```
5x5 Array{Float64,2}:  
 1.0  0.0  0.0  0.0  0.0  
 0.0  1.0  0.0  0.0  0.0  
 0.0  0.0  1.0  0.0  0.0  
 0.0  0.0  0.0  1.0  0.0  
 0.0  0.0  0.0  0.0  1.0
```

We can also create matrices by hand. Here, spaces delimit the columns and semicolons delimit the rows:

In [83]:

```
[1 2; 3 4; 5 6]
```

Out[83]:

3x2 Array{Int64,2}:

```
1  2
3  4
5  6
```

In [84]:

```
Float64[1 2; 3 4; 5 6]
```

Out[84]:

3x2 Array{Float64,2}:

```
1.0  2.0
3.0  4.0
5.0  6.0
```

We can also create matrices using brackets, a formula, and a `for` command:

In [85]:

```
[k^2+j for k=1:5,j=1:5]
```

Out[85]:

5x5 Array{Int64,2}:

```
2  3  4  5  6
5  6  7  8  9
10 11 12 13 14
17 18 19 20 21
26 27 28 29 30
```

Matrices are really Vectors in disguise. They are still stored in memory in a sequence of addresses. We can see the underlying vector using the `vec` command:

In [86]:

```
M=[1 2; 3 4; 5 6]
```

```
vec(M)
```

Out[86]:

6-element Array{Int64,1}:

```
1
3
5
2
4
6
```

The only difference between matrices and vectors from the computers perspective is that they have a size which changes the interpretation of whats stored in memory:

In [87]:

```
size(M)
```

Out[87]:

```
(3,2)
```

Matrices can be manipulated easily on a computer. We can easily take determinants:

In [88]:

```
M=rand(3,3)  
det(M)
```

Out[88]:

```
-0.14858065074498047
```

Or multiply:

In [89]:

```
M*M
```

Out[89]:

```
3x3 Array{Float64,2}:  
 0.615876  0.457927  0.19322  
 0.784475  1.06339   0.5475  
 0.100306  0.403361  0.312523
```

In [36]:

```
[1 2; 3 4]*[4 5; 6 7]
```

Out[36]:

```
2x2 Array{Int64,2}:  
 16  19  
 36  43
```

If you use `.*`, it does entrywise multiplication:

In [90]:

```
[1 2; 3 4].*[4 5; 6 7]
```

Out[90]:

2x2 Array{Int64,2}:

```
 4  10
18  28
```

Vectors are thought of as column vectors, and so `*` is not defined:

In [91]:

```
a=[1,2,3]
b=[4,5,6]
```

```
a*b
```

LoadError: MethodError: `*` has no method matching *(::Array{Int64,1}, ::Array{Int64,1})

Closest candidates are:

```
*(::Any, ::Any, !Matched::Any, !Matched::Any...)
*{T<:Union{Complex{Float32},Complex{Float64},Float32,Float64},S}(!Matched::Union{DenseArray{T<:Union{Complex{Float32},Complex{Float64},Float32,Float64},2},SubArray{T<:Union{Complex{Float32},Complex{Float64},Float32,Float64},2,A<:DenseArray{T,N},I<:Tuple{Vararg{Union{Colon,Int64,Range{Int64}}}},LD}}, ::Union{DenseArray{S,1},SubArray{S,1,A<:DenseArray{T,N},I<:Tuple{Vararg{Union{Colon,Int64,Range{Int64}}}},LD}})
*{TA,TB}(!Matched::Base.LinAlg.AbstractTriangular{TA,S<:AbstractArray{T,2}}, ::Union{DenseArray{TB,1},DenseArray{TB,2},SubArray{TB,1,A<:DenseArray{T,N},I<:Tuple{Vararg{Union{Colon,Int64,Range{Int64}}}},LD},SubArray{TB,2,A<:DenseArray{T,N},I<:Tuple{Vararg{Union{Colon,Int64,Range{Int64}}}},LD}})
...
```

while loading In[91], in expression starting on line 5

Whereas entry-wise multiplication works fine:

In [92]:

```
a.*b
```

Out[92]:

3-element Array{Int64,1}:

```
 4
10
18
```

Transposing a Vector gives a row vector, which is represented by a 1 x n matrix:

In [93]:

```
a'
```

Out[93]:

```
1x3 Array{Int64,2}:  
 1  2  3
```

Thus we can do dot products as follows:

In [42]:

```
a'*b
```

Out[42]:

```
1-element Array{Int64,1}:  
 32
```

This is a vector with one entry, because matrix-vector multiplication always returns a vector. If we use `dot`, we get the dot product as a scalar:

In [43]:

```
dot(a,b)
```

Out[43]:

```
32
```

One important note: a vector is not the same as an $n \times 1$ matrix:

In [106]:

```
v=zeros(Int,3,1)  
v[1:3,:]=1:3      # the : notation means all columns  
  
v
```

Out[106]:

```
3x1 Array{Int64,2}:  
 1  
 2  
 3
```

In [104]:

```
a
```

Out[104]:

```
3-element Array{Int64,1}:  
 1  
 2  
 3
```

In [105]:

```
v==a
```

Out[105]:

```
false
```

Finally, we can solve linear systems use \:

In [47]:

```
A=rand(5,5)  
b=ones(5)  
  
u=A\b
```

Out[47]:

```
5-element Array{Float64,1}:  
 0.0880726  
 0.184264  
-0.493238  
 0.606102  
 1.21375
```

In [48]:

```
A*u-b
```

Out[48]:

```
5-element Array{Float64,1}:  
 0.0  
 0.0  
 2.22045e-16  
 0.0  
 2.22045e-16
```

Back to the spring problem:

We now create the relevant matrices and vectors in Julia. We first create A by creating an $n+1 \times n$ matrix of zeros, and setting the diagonal to 1 and the subdiagonal to -1:

In [108]:

```
n=3  # the number of balls

A=zeros(n+1,n)

for k=1:n
    A[k,k]=1
    A[k+1,k]=-1
end

A
```

Out[108]:

```
4x3 Array{Float64,2}:
 1.0  0.0  0.0
-1.0  1.0  0.0
 0.0 -1.0  1.0
 0.0  0.0 -1.0
```

We now create C , where we assume the stiffness is equal. We then can create K and \mathbf{f} and solve the system using \backslash to find \mathbf{u} :

In [110]:

```
C=eye(n+1)

K=A'*C*A

f=[0,1,0.]

u=K\f  # solves for the displacement u
```

Out[110]:

```
3-element Array{Float64,1}:
 0.5
 1.0
 0.5
```

The masses are now shifted by u . We can plot the new locations in red versus the original locations in green as follows:

In [112]:

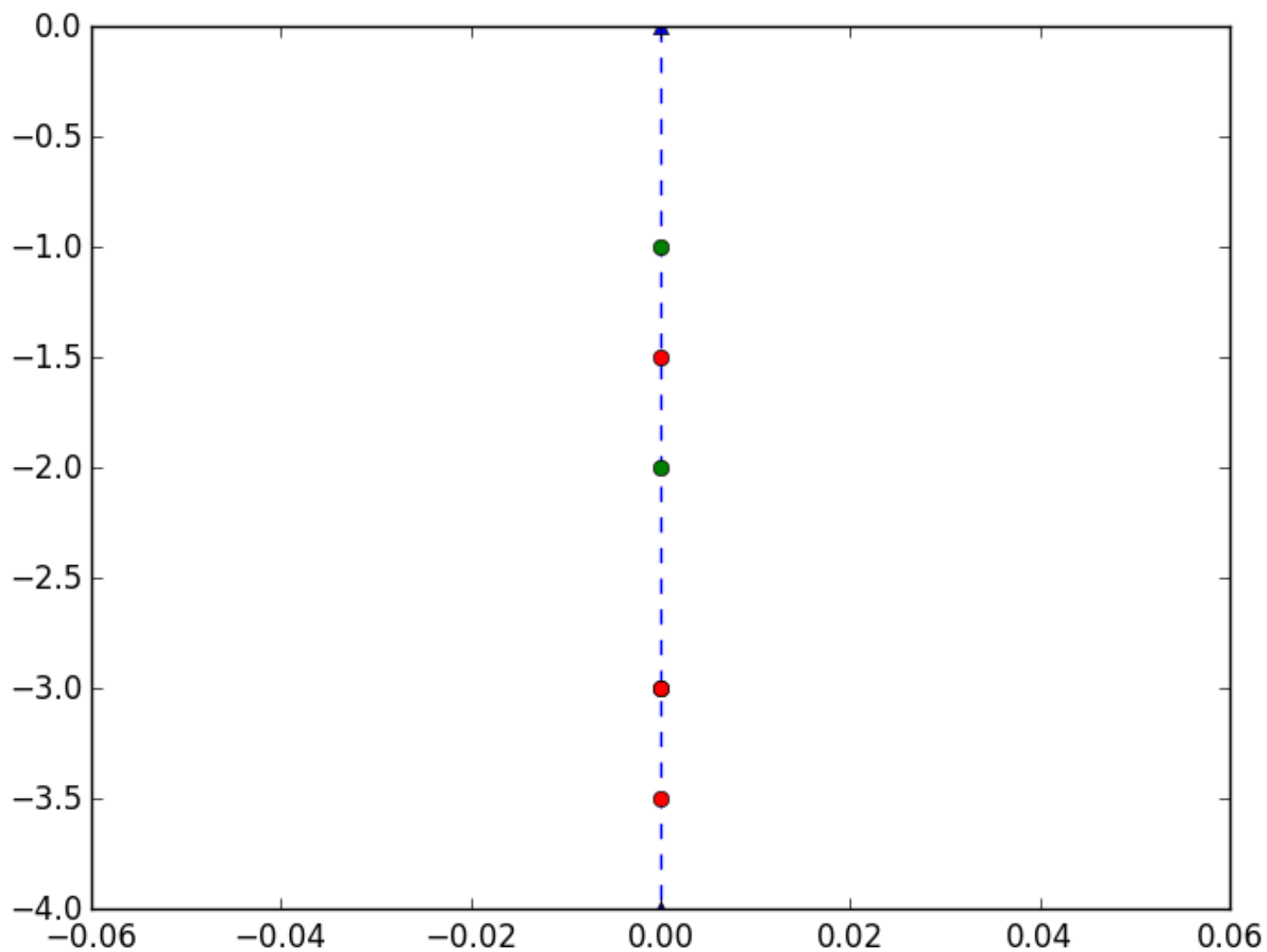
```
plot([0.,0.],[0.,-4.];marker="^",linestyle="--")
```

```
p=1:n
```

```
newp=p+u
```

```
plot(zeros(n),-p;marker="o",linestyle="")
```

```
plot(zeros(n),-newp;marker="o",linestyle="")
```



Out[112]:

```
1-element Array{Any,1}:
```

```
PyObject <matplotlib.lines.Line2D object at 0x319ef2f90>
```

Let's make this example interactive!

In [67]:

```
using Interact # let's you use @manipulate
```

In [113]:

```
n=3 # the number of balls
```

```

# set up A
A=zeros(n+1,n)

for k=1:n
    A[k,k]=1
    A[k+1,k]=-1
end

# the original locations of the springs
p=1:n

# creates a figure to modify
fig=figure()

# @manipulate creates two sliders: one for F and one for c. F represents the force
# on the middle ball and c the stiffness of the springs
@manipulate for F=-2.:0.1:2., c=1.:10.
    withfig(fig) do
        C=c*eye(n+1) # this is the stiffnest matrix

        K=A'*C*A

        f=[0,F,0.]
        u=K\f
        newp=p+u

        plot([0.,0.],[0.,-4.];marker="^",linestyle="--")
        plot(zeros(n),-newp;marker="o",linestyle="")
    end
end
end

```

Out[113]:

