

Last name _____

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LARSON—MATH 310—BONUS WORKSHEET 01

Numpy Basics.

1. Start the Chrome browser.
2. Go to <https://cocalc.com>
3. Log in.
4. You should see an existing Project for our class. Click on that.
5. Click “New”, then “Jupyter Notebook”, then call it **310-b01**.
6. Make sure you have PYTHON as the *kernel*.

Instructions: Run all of the code chunks. **Annotate** your Jupyter notebook cells so that it is clear to anyone what problem your work corresponds to.

Numpy is the standard library for numerical linear algebra computing. It is based on *arrays* holding uniform type data (floats). Under the hood are the well-researched BLAS and Lapack libraries. Read more here: <https://en.wikipedia.org/wiki/NumPy>.

Matrix basics

1. A matrix can be represented as an array of arrays.

```
1 from numpy import array
2 A = array([[1, 2, 3], [4, 5, 6]])
3 print(A)
4 type(A)
5
```

2. We can add matrices.

```
1 from numpy import array
2 A = array([[1, 2, 3], [4, 5, 6]])
3 B = array([[1, 3, 5], [2, 4, 6]])
4 print(A + B)
5
```

3. We can multiply a matrix times a vector.

```
1 from numpy import array, dot
2 A = array([[1, 2], [3, 4]])
3 A.dot([5, 6])
4
```

4. We can multiply a 2×3 matrix times a 3×2 matrix.

```
1 import numpy as np
2 A = [[1, 2, 3], [4, 5, 6]]
3 B = [[1, 2], [3, 4], [5, 6]]
4 np.dot(A, B)
5
```

5. We can also multiply these matrices using the overloaded “`*`” operator

```
1 import numpy as np
2 A = [[1, 2, 3], [4, 5, 6]]
3 B = [[1, 2], [3, 4], [5, 6]]
4 np.matrix(A) * np.matrix(B)
5
```

6. It can be useful to initialize a vector or matrix with all 0’s.

```
1 import numpy as np
2 np.zeros((2,3))
3
```

7. We can create and *identity matrix* of any size.

```
1 import numpy as np
2 np.identity(3)
3
```

Matrix inversion

8. One way to find the inverse of an invertible matrix is with the “`inv`” function.

```
1 A = [[1, 2], [2, 1]]
2 from numpy.linalg import inv
3 inv(A)
4
```

9. Let’s check that what we got really is the inverse.

```
1 A = [[1, 2], [2, 1]]
2 from numpy.linalg import inv
3 B=inv(A)
4 np.matrix(A) * np.matrix(B)
5
```

10. We can also find inverses with the overloaded Python exponentiation operator “`**`”.

```
1 A = [[1, 2], [2, 1]]
2 from numpy import matrix
3 matrix(A)**(-1)
4
```

Matrix rank

11. We can find the rank (which is the same as both the column rank and row rank). The matrix here is 3×3 , so if the rank is less than 3 we know that both the rows are linearly dependent and the columns are linearly dependent.

```
1 from numpy import *
2 A = array([[1, 2, 3], [2, 3, 4], [3, 4, 5]])
3 linalg.matrix_rank(A)
4
```

Solving matrix equations

12. Given a matrix A and a vector \hat{b} we want to find a vector \hat{x} so that $A\hat{x} = \hat{b}$. You should check (by hand) that the produced solution works.

```
1 import numpy as np
2 A = np.array([[1, 2], [3, 4]])
3 b = np.array([8, 18])
4 print("Solution of linear equations:", np.linalg.solve(A, b))
5
```

Eigenvalues

13. We can find eigenvalues of a square matrix. The eigenvalues are, in general complex, so the output is represented as complex numbers.

```
1 from numpy.linalg import eig
2 A = [[3, -2, 2], [2, -1, 4], [2, -2, 1]]
3 lmd, vec = eig(A)
4 lmd
5
```

14. We can also print the corresponding eigenvectors.

```
1 print(vec)
2
```

15. Let's check that the first listed eigenvalue and first listed eigenvector actually work. We should have, for instance, that $A * v_0 = \lambda v_0$.

```
1 from numpy.linalg import eig
2 A = [[3, -2, 2], [2, -1, 4], [2, -2, 1]]
3 lmd, vec = eig(A)
4 lambda0 = lmd[0]
5 v0 = vec[0]
6 print(np.dot(A, v0))
7 D = lambda0*np.identity(3)
8 print(np.dot(D, v0))
9
```

Getting your homework recorded

When you are done,

1. Click the “Print” menu choice (under “File”) and make a pdf of this worksheet (html is OK too).
2. Send me an email (cclarson@vcu.edu) with an informative header like “Math 310 - b01 worksheet attached” (so that it will be properly recorded).
3. Remember to attach your homework worksheet!