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 Plan
 

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- 1 Complex vector spaces and Hermitian matrices
  - 2 Non-negative matrices and Perron-Frobenius theory
  - 3 Ranking models and Population growth models
- 

 ① Complex vector spaces and Hermitian matrices

Ex:  $A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$   $\lambda^2 + 1 = 0$   $\lambda^2 = -1$   $\lambda = \pm i$   $\lambda_1 = i, \lambda_2 = -i$  Complex eigenvalues

Eigenvectors:

$\lambda = i$ :  $E_i = \text{Null} \begin{pmatrix} -i & -1 \\ 1 & -i \end{pmatrix}$

$\begin{pmatrix} -i & -1 \\ 1 & -i \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{pmatrix} 1 & -i \\ -i & -1 \end{pmatrix} \xrightarrow{R_2 + iR_1} \begin{pmatrix} 1 & -i \\ 0 & 0 \end{pmatrix}$   $x - iy = 0$   
 $y$  free

$\underline{v} = \begin{pmatrix} iy \\ y \end{pmatrix} = y \cdot \begin{pmatrix} i \\ 1 \end{pmatrix}$   $\underline{v}_1 = \begin{pmatrix} i \\ 1 \end{pmatrix}$

$\lambda = -i$ :  $\begin{pmatrix} i & -1 \\ 1 & i \end{pmatrix} \rightarrow \begin{pmatrix} 1 & i \\ 0 & 0 \end{pmatrix}$   $x + iy = 0$   
 $y$  free

$\underline{v} = \begin{pmatrix} -iy \\ y \end{pmatrix} = y \cdot \begin{pmatrix} -i \\ 1 \end{pmatrix}$   $\underline{v}_2 = \begin{pmatrix} -i \\ 1 \end{pmatrix}$

Complex vectors:  $\underline{v} = (v_1, v_2, \dots, v_n)$  where  $v_i$  is a complex number

Inner product:  $\langle \underline{v}, \underline{w} \rangle = \overline{v_1} w_1 + \overline{v_2} w_2 + \dots + \overline{v_n} w_n$   
 $\underline{v} \cdot \underline{w}$   $\overline{a+bi} = a-bi$

Ex:  $(i, 1) \cdot (-i, 1) = \overline{i} \cdot (-i) + \overline{1} \cdot 1 = (-i)(-i) + 1 \cdot 1 = i^2 + 1 = 0$

$(i, 1) \cdot (i, 1) = \overline{i} \cdot i + \overline{1} \cdot 1 = (-i)i + 1 \cdot 1 = -i^2 + 1 = 1 + 1 = 2$   
 $\underline{v} \cdot \underline{v} = \|\underline{v}\|^2$

Note:  $\langle \underline{v}, \underline{v} \rangle = \|\underline{v}\|^2 \geq 0$   
 $\underline{v} \perp \underline{w} \Leftrightarrow \underline{v} \cdot \underline{w} = 0$

Defn:

$A = (a_{ij})$  complex matrix  
 $A^* = (\overline{a_{ij}})^T$

Ex:  $A = \begin{pmatrix} -i & -1 \\ 1 & -i \end{pmatrix}$   $A^* = \begin{pmatrix} \overline{-i} & \overline{-1} \\ \overline{1} & \overline{-i} \end{pmatrix}^T = \begin{pmatrix} i & -1 \\ 1 & i \end{pmatrix}^T = \begin{pmatrix} i & 1 \\ -1 & i \end{pmatrix}$

Defn:

$A$  is Hermitian if  $A^* = A$  (real: symmetric)  
and unitary if  $A^* = A^{-1}$  (orthogonal)

Ex:  $\begin{pmatrix} 2 & 1-i \\ 1+i & 3 \end{pmatrix}$  is Hermitian

Theorem: A Hermitian  $n \times n$  matrix

- i) A has  $n$  eigenvalues that are real (counted with mlt.)
- ii) If  $\underline{v}, \underline{w}$  are eigenvectors of A with distinct eigenvalues, then  $\underline{v} \cdot \underline{w} = 0$ .
- iii) There is a unitary matrix U such that  $U^* A U = D$  is diagonal.

Pf:

- i) If  $\lambda$  is an eigenvalue of A, with eigenvector  $\underline{v}$ , then

$$\begin{aligned} \langle A\underline{v}, \underline{v} \rangle &= \langle \lambda \underline{v}, \underline{v} \rangle \\ &= (\lambda \underline{v})^* \underline{v} \\ &= \bar{\lambda} \underline{v}^* \underline{v} \\ &= \bar{\lambda} \cdot \|\underline{v}\|^2 \end{aligned}$$

$$\begin{aligned} \langle \underline{v}, A\underline{v} \rangle &= \langle \underline{v}, \lambda \underline{v} \rangle \\ &= \underline{v}^* \lambda \underline{v} \\ &= \lambda \underline{v}^* \underline{v} \\ &= \lambda \cdot \|\underline{v}\|^2 \end{aligned}$$

$$\Downarrow \\ \lambda = \bar{\lambda}, \text{ i.e. } \lambda \text{ is real}$$

$$\text{ii) } \left. \begin{aligned} A\underline{v} &= \lambda_1 \underline{v} \quad (\lambda_1 \neq \lambda_2) \\ A\underline{w} &= \lambda_2 \underline{w} \end{aligned} \right\} \begin{aligned} \langle \underline{v}, A\underline{w} \rangle &= \langle \underline{v}, \lambda_2 \underline{w} \rangle = \lambda_2 \langle \underline{v}, \underline{w} \rangle \\ &= \end{aligned}$$

$$\begin{aligned} \langle A\underline{v}, \underline{w} \rangle &= \langle \lambda_1 \underline{v}, \underline{w} \rangle = \lambda_1 \langle \underline{v}, \underline{w} \rangle \\ &= \lambda_1 \langle \underline{v}, \underline{w} \rangle \end{aligned}$$

$$\Downarrow \\ (\lambda_2 - \lambda_1) \langle \underline{v}, \underline{w} \rangle = 0 \Rightarrow \langle \underline{v}, \underline{w} \rangle = 0$$

Note: i) Fundamental thm of algebra

$a_n z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0 = 0$   
has  $n$  complex roots where  
 $a_n, \dots, a_1, a_0$  are real/complex numbers.

- ii) A is Hermitian  $\Leftrightarrow \langle A\underline{v}, \underline{w} \rangle = \langle \underline{v}, A\underline{w} \rangle$

$$\begin{aligned} \langle \underline{v}, \underline{w} \rangle &= \bar{v}_1 w_1 + \dots + \bar{v}_n w_n \\ &= \underline{v}^* \cdot \underline{w} \\ &= (\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n) \cdot \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} \end{aligned}$$

$\Downarrow$

$$\begin{aligned} \langle A\underline{v}, \underline{w} \rangle &= (A\underline{v})^* \underline{w} = \underline{v}^* A^* \underline{w} \\ &= \underline{v}^* A \underline{w} \\ \langle \underline{v}, A\underline{w} \rangle &= \underline{v}^* A \underline{w} \end{aligned}$$

iii) A Hermitian: Let  $\underline{v}_1$  be an eigenvector of  $A$ , with eigenvalue  $\lambda_1$

Defn:  $V_2 = \{ \underline{w} : \langle \underline{v}_1, \underline{w} \rangle = 0 \} = \text{Null}(\underline{v}_1^*)$   
 $\underline{v}_1^* \cdot \underline{w} = 0$  (subspace of dim  $n-1$ )

Note:  $\underline{v}_2 \in V_2 \Rightarrow A\underline{v}_2 \in V_2$

$$\underline{v}_1^* \cdot \underline{v}_2 = 0$$

$$\langle \underline{v}_1, \underline{v}_2 \rangle = 0$$

$$\langle \underline{v}_1, A\underline{v}_2 \rangle = \langle A\underline{v}_1, \underline{v}_2 \rangle$$

$$= \langle \lambda_1 \underline{v}_1, \underline{v}_2 \rangle$$

$$= \overline{\lambda_1} \langle \underline{v}_1, \underline{v}_2 \rangle$$

$$= \lambda_1 \langle \underline{v}_1, \underline{v}_2 \rangle = 0$$

$$A: V_2 \rightarrow V_2 \quad \text{Hermitian}$$

$$\underline{v}_2 \mapsto A\underline{v}_2$$

} induction

You find eigenvectors  $\underline{v}_1, \underline{v}_2, \underline{v}_3, \dots, \underline{v}_n$  of  $A$  which are orthogonal.

$$U = \left( \frac{1}{\|\underline{v}_1\|} \underline{v}_1 \mid \frac{1}{\|\underline{v}_2\|} \underline{v}_2 \mid \dots \right) \quad D = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \end{pmatrix}$$



# Untitled9

September 28, 2020

```
[1]: import numpy as np
      from numpy import linalg as LA ✓
```

```
[2]: A = np.array([[0.75,0.02,0.1],[0.2,0.9,0.2],[0.05,0.08,0.7]])
      B = np.array([[0,0,6],[1/2,0,0],[0,1/3,0]]) }
```

```
[3]: LA.eig(A)
```

```
[3]: (array([1. , 0.7 , 0.65]),
      array([[-1.88144174e-01, -8.08122036e-01, -7.07106781e-01],
             [-9.40720868e-01,  5.05076272e-01, -1.59384710e-15],
             [-2.82216261e-01,  3.03045763e-01,  7.07106781e-01]]))
```

*Handwritten notes:*  
← eig. values  
bnd of  $E_1$  →  
↑  $\lambda < 0.7$   
↑  $\lambda = 0.65$

```
[4]: LA.matrix_power(A,5) ←  $A^5$ 
```

```
[4]: array([[0.28940219, 0.09010763, 0.17337312],
            [0.55462    , 0.72269    , 0.55462    ],
            [0.15597781, 0.18720238, 0.27200687]])
```

```
[5]: LA.matrix_power(A,10)
```

```
[5]: array([[0.16077148, 0.12365308, 0.14730873],
            [0.64783498, 0.67608251, 0.64783498],
            [0.19139354, 0.20026441, 0.20485628]])
```

```
[6]: LA.matrix_power(A,100)
```

```
[6]: array([[0.13333333, 0.13333333, 0.13333333],
            [0.66666667, 0.66666667, 0.66666667],
            [0.2       , 0.2       , 0.2       ]])
```

*Handwritten note:* ←  $\lambda$  eq. state

```
[7]: LA.eig(B)
```

```
[7]: (array([-0.5+0.8660254j, -0.5-0.8660254j,  1. +0.j      ]),
      array([[ 0.88465174+0.j      ,  0.88465174-0.j      ,
              -0.88465174+0.j      ],
             [-0.22116293-0.38306544j, -0.22116293+0.38306544j,
              -0.44232587+0.j      ]],
```

*Handwritten note:*  $\lambda = \sqrt{3} = i$

```
[-0.07372098+0.12768848j, -0.07372098-0.12768848j,
 -0.14744196+0.j      ]])
```

```
[8]: LA.matrix_power(B,5)
```

```
[8]: array([[0.      , 2.      , 0.      ],
            [0.      , 0.      , 3.      ],
            [0.16666667, 0.      , 0.      ]])
```

```
[9]: LA.matrix_power(B,10)
```

```
[9]: array([[0.      , 0.      , 6.      ],
            [0.5     , 0.      , 0.      ],
            [0.      , 0.33333333, 0.      ]])
```

```
[10]: LA.matrix_power(B,100)
```

```
[10]: array([[0.      , 0.      , 6.      ],
            [0.5     , 0.      , 0.      ],
            [0.      , 0.33333333, 0.      ]])
```

*k=3*

```
[11]: LA.matrix_power(B,101)
```

```
[11]: array([[0.      , 2.      , 0.      ],
            [0.      , 0.      , 3.      ],
            [0.16666667, 0.      , 0.      ]])
```

```
[12]: LA.matrix_power(B,102)
```

```
[12]: array([[1., 0., 0.],
            [0., 1., 0.],
            [0., 0., 1.]])
```

```
[ ]:
```

## ② Non-negative matrices and Perron-Frobenius th.

$A = (a_{ij})$   $n \times n$ -matrix

$\underline{v} = (v_i)$   $n$ -vector

Defn:

$A$  is positive  $A > 0$   
if  $a_{ij} > 0$  for all  $i, j$

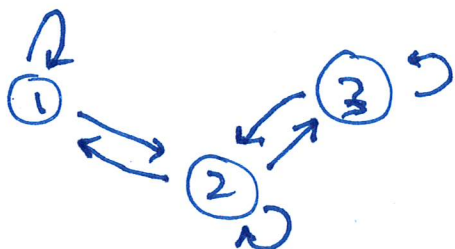
$A$  is non-negative  $A \geq 0$   
if  $a_{ij} \geq 0$  — " —

Interpretation: Graph of  $A$

nodes:  $1, 2, 3, \dots, n$

edges:  $i \rightarrow j$  if  $a_{ji} > 0$

Ex:  $A = \begin{pmatrix} 0.8 & 0.1 & 0 \\ 0.2 & 0.7 & 0.3 \\ 0 & 0.2 & 0.7 \end{pmatrix} \geq 0$



Strongly connected  
 $A$  irreducible

Defn:

The graph is strongly connected if there is a path from any node to any other node.

A non-negative matrix  $A \geq 0$  is irreducible if the graph is strongly connected, and  $A$  is primitive if  $A^m > 0$  for some positive integer  $m \geq 1$ .

Fact:

$A$  positive  $\Rightarrow A$  primitive  $\Rightarrow A$  irreducible

A primitive  $\Rightarrow$  there exist an integer  $m \geq 1$  such that you can get from any node to any other node in exactly  $m$  steps.

$$A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

irreducible, not primitive



$$A^1 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

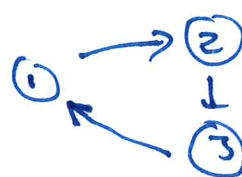
$$A^2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$A^3 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Note: For Markov chains, regular = primitive.

Ex:  $A = \begin{pmatrix} 0 & 0 & 6 \\ 1/2 & 0 & 0 \\ 0 & 1/3 & 0 \end{pmatrix}$

irreducible, not primitive



$$\begin{vmatrix} -\lambda & 0 & 6 \\ 1/2 & -\lambda & 0 \\ 0 & 1/3 & -\lambda \end{vmatrix} = -\lambda(\lambda^2 - 0) + 6 \cdot \left(\frac{1}{6} - 0\right) = 0$$

$$-\lambda^3 + 1 = 0$$

$$\lambda^3 = 1$$

$$\lambda_1 = 1 \quad \text{or} \quad \lambda^2 + \lambda + 1 = 0$$

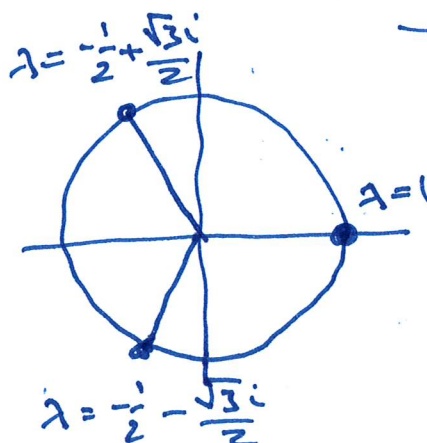
$$\lambda = \frac{-1 \pm \sqrt{1-4}}{2}$$

$$= -\frac{1}{2} \pm \frac{\sqrt{3}i}{2}$$

$$\frac{(\lambda^3 - 1) : (\lambda - 1) = \lambda^2 + \lambda + 1}{\lambda^3 - \lambda^2}$$

$$\frac{\lambda^2 - \lambda^2}{\lambda^2 - \lambda}$$

$$\lambda - 1$$







Thm (Perron - Frobenius)

Let  $A \geq 0$   $n \times n$  irreducible matrix. Then:

- i) There is a dominant eigenvalue  $\lambda_A > 0$  such that  $|\lambda| \leq \lambda_A$  for all other (real or complex) eigenvalues of  $A$ .
- ii) There is a positive eigenvector  $\underline{v}_A > 0$  with eigenvalue  $\lambda_A$ .
- iii)  $\lambda_A$  has multiplicity 1, and any eigenvector  $\underline{v} \geq 0$  of  $A$  is a positive multiple of  $\underline{v}_A$ .
- iv) all eigenvalues of  $A$  with  $|\lambda| = \lambda_A$  are all the solutions of  $\lambda^k = \lambda_A^k$  for some  $k \geq 1$  (index of primitivity)


Note:  $k=1$  if  $A$  is primitive  
 $k > 1$  otherwise

Ex:  $A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$   not primitive

$\lambda^2 - 1 = 0$   $\lambda_1 = 1$   $\lambda_2 = -1$  

$\lambda_A = 1$ :  $\lambda^2 = 1$   $E_{11} \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}$   
 $k=2$   $\underline{v}_A = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$   $\underline{v} = \begin{pmatrix} y \\ y \end{pmatrix} = y \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

$E_{-1} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$   
 $y \cdot \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

$A = \begin{pmatrix} 0 & 0 & 6 \\ 1/2 & 0 & 0 \\ 0 & 1/3 & 0 \end{pmatrix}$   not primitive

$\lambda = 1$  or  $\lambda = -\frac{1}{2} \pm \frac{\sqrt{3}}{2}i$  not primitive

$\lambda_A = 1$ :  $\underline{v}_A: \begin{pmatrix} -1 & 0 & 6 \\ 1/2 & -1 & 0 \\ 0 & 1/3 & -1 \end{pmatrix}$   
 $k=3 \rightarrow \begin{pmatrix} -1 & 0 & 6 \\ 0 & 1/3 & -1 \\ 0 & 0 & 0 \end{pmatrix}$

$\underline{v} = \begin{pmatrix} 6z \\ 3z \\ z \end{pmatrix} = z \cdot \begin{pmatrix} 6 \\ 3 \\ 1 \end{pmatrix}$   $-x + 6z = 0$   
 $1/3y - z = 0$   
 $z$  free

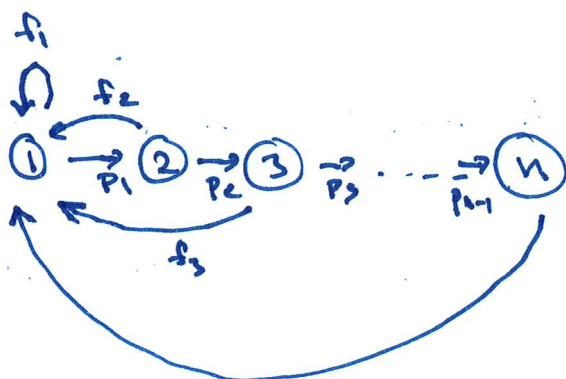
$\underline{v}_A = \begin{pmatrix} 6 \\ 3 \\ 1 \end{pmatrix}$

### ③ Population growth models

Leslie matrices:

$$A = \begin{pmatrix} f_1 & f_2 & f_3 & \dots & f_{n-1} & f_n \\ p_1 & 0 & 0 & \dots & 0 & 0 \\ 0 & p_2 & 0 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & p_{n-1} & 0 \end{pmatrix}$$

$n \times n$  matrix where  
 $f_1, \dots, f_n \geq 0$   $f_n \neq 0$   
 $0 < p_i < 1$



$x_1$ : 0-1 years old

$x_2$ : 1-2 years old

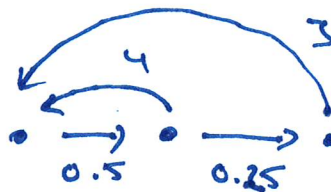
$\vdots$

$x_n$ :  $(n-1)$ - $n$  years old

A is irreducible

A is primitive  $\Leftrightarrow$  more than one  $f_i \neq 0$

Ex:  $A = \begin{pmatrix} 0 & 4 & 3 \\ 0.5 & 0 & 0 \\ 0 & 0.25 & 0 \end{pmatrix}$



$$\begin{vmatrix} -\lambda & 4 & 3 \\ 0.5 & -\lambda & 0 \\ 0 & 0.25 & -\lambda \end{vmatrix} = 0$$

$$\begin{aligned} -\lambda(\lambda^2 - 2) - 0.25(-1.5) &= 0 \\ -\lambda^3 + 2\lambda + 0.375 &= 0 \end{aligned}$$

i) Use Python to find eigenvalues/eigenvector

ii) Compute  $A^m \cdot \underline{v}_0$

$$\underline{v}_0 = \begin{pmatrix} 10 \\ 10 \\ 10 \end{pmatrix} \quad \underline{v}_1 = A \underline{v}_0 = \begin{pmatrix} 70 \\ 5 \\ 2.5 \end{pmatrix} \quad \underline{v}_2 = A \underline{v}_1 = \begin{pmatrix} 27.5 \\ 35 \\ 0.625 \end{pmatrix}$$

## Example Leslie matrices

October 6, 2020

```
[8]: import numpy as np
import pandas as pd
```

```
[9]: A = np.array([[0,4,3],[0.5,0,0],[0,0.25,0]])
v = np.array([[10],[10],[10]])
```

```
[3]: eval, evec = np.linalg.eig(A)
```

```
[4]: eval
```

```
[4]: array([ 1.5          , -1.30901699, -0.19098301])
```

```
[5]: w = evec[:,0]
```

```
[7]: w/w.sum()
```

```
[7]: array([0.72, 0.24, 0.04])
```

```
[10]: rows = v.shape[0]
for i in range(12):
    last = v.shape[1]
    w = np.reshape(v[:,last-1],(rows,1))
    v = np.append(v,A.dot(w),axis=1)
```

```
[11]: v
```

```
[11]: array([[1.00000000e+01, 7.00000000e+01, 2.75000000e+01, 1.43750000e+02,
            8.12500000e+01, 2.97812500e+02, 2.16406250e+02, 6.26093750e+02,
            5.44492188e+02, 1.33333984e+03, 1.32376953e+03, 2.87086426e+03,
            3.14754150e+03],
            [1.00000000e+01, 5.00000000e+00, 3.50000000e+01, 1.37500000e+01,
            7.18750000e+01, 4.06250000e+01, 1.48906250e+02, 1.08203125e+02,
            3.13046875e+02, 2.72246094e+02, 6.66669922e+02, 6.61884766e+02,
            1.43543213e+03],
            [1.00000000e+01, 2.50000000e+00, 1.25000000e+00, 8.75000000e+00,
            3.43750000e+00, 1.79687500e+01, 1.01562500e+01, 3.72265625e+01,
            2.70507812e+01, 7.82617188e+01, 6.80615234e+01, 1.66667480e+02,
```

```
1.65471191e+02]])
```

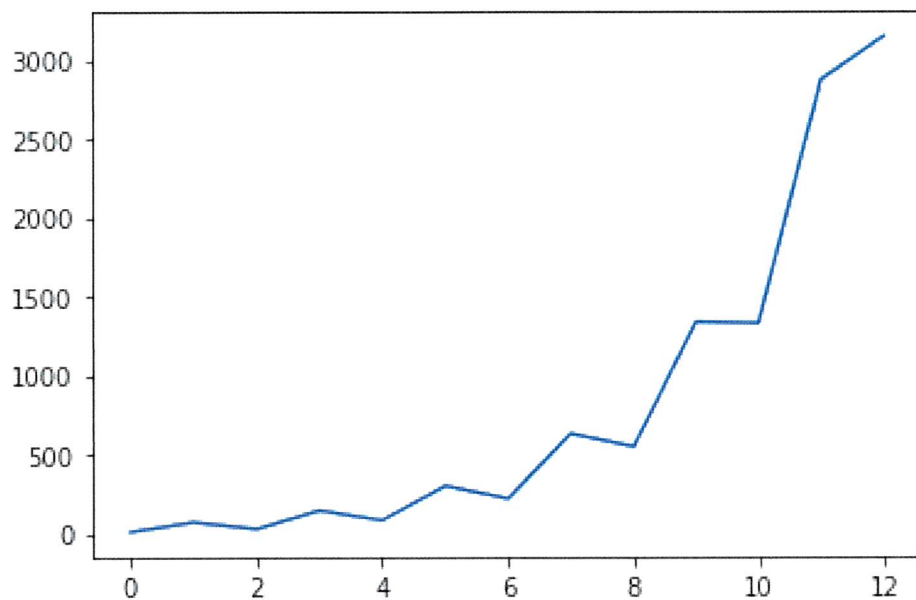
```
[12]: u = pd.Series(v[0])
```

```
[13]: u
```

```
[13]: 0      10.000000  
      1      70.000000  
      2      27.500000  
      3     143.750000  
      4      81.250000  
      5     297.812500  
      6     216.406250  
      7     626.093750  
      8     544.492188  
      9    1333.339844  
     10    1323.769531  
     11    2870.864258  
     12    3147.541504  
      dtype: float64
```

```
[14]: u.plot()
```

```
[14]: <matplotlib.axes._subplots.AxesSubplot at 0x26ca0cf7490>
```

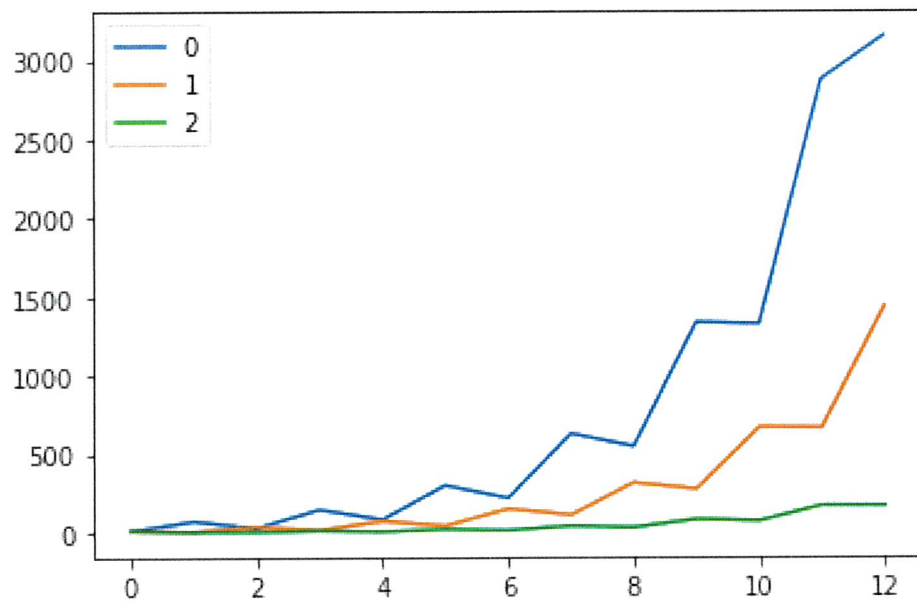


```
[17]: u = pd.DataFrame(np.transpose(v))
```



```
[18]: u.plot()
```

```
[18]: <matplotlib.axes._subplots.AxesSubplot at 0x26ca0ed6250>
```



```
[ ]:
```