

estabilidad

June 22, 2018

1 Ecuación logística

Sea

$$x_{n+1} = \mu x_n (1 - x_n)$$

la ecuación logística, la familia de funciones es dada por $F_\mu(x) = \mu x(1 - x)$, entonces

$$x_{n+1} = F_\mu(x_n)$$

Analizaremos las bifurcaciones partiendo el espacio en una grilla de resolución $4/n$ para el eje x y $1/n$ para el y , ya que la ecuación está definida en $[0, 4] \times [0, 1]$. En ese caso encenderemos todos los $(\mu, F_\mu(x))$ para un x dado.

```
In [84]: import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

In [85]: def f(mu, x):
    return np.multiply(np.multiply(mu,x),(1-x))

    def fn(mu, x, n):
        if n == 0:
            return f(mu, x)
        return fn(mu, f(mu, x), n-1)

    n = 250

In [98]: mus = np.linspace(0,4,n*4)

In [99]: def mu_fmu(x0, mu, n):
    return (mu, fn(mu, x0, n))

    def linspace_to_mcoords(array, xmin, samples):
        # mover las muestras y hacerlas un número entero
        return ((array - xmin)*samples).astype(int)

    def mcoords_to_linspace(array, xmin, samples):
        # hacer lo opuesto
        return ((array/samples)+xmin)
```

```

In [100]: # hacer la matriz
mat = np.zeros((n, n*4))

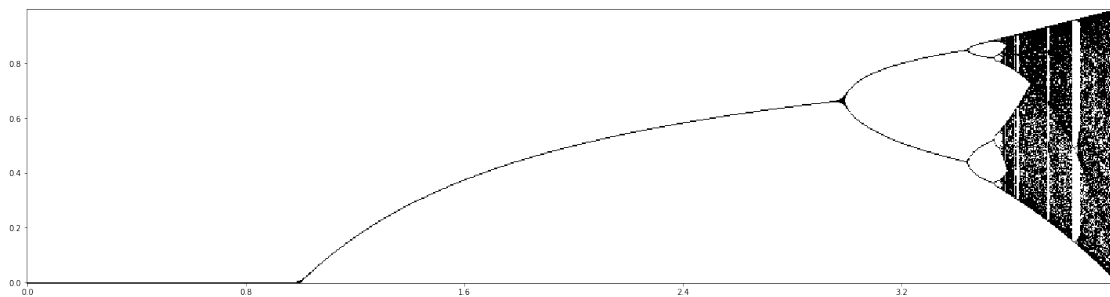
for i in range(100, 500):
    # calcular los  $f_{\mu}(x_0)$ 
    fmus = fn(mus, 1/2, i)

    # ponerlos en los índices de la matriz
    matfmus = linspace_to_mcoords(fmus, 0, n)

    for i in range(n*4):
        mat[matfmus[i], i] = 1

In [101]: fig = plt.figure(figsize=(25,7))
ax = fig.add_subplot(111)
ax.imshow(mat, cmap='Greys', interpolation='nearest', origin='lower')
_ = ax.set_xticklabels(mcoords_to_linspace(ax.get_xticks(), 0, n))
_ = ax.set_yticklabels(mcoords_to_linspace(ax.get_yticks(), 0, n))
fig.savefig('diagBifur.pdf', bbox_inches='tight')
plt.show()

```



```

In [11]: import matplotlib.animation as animation
from matplotlib import animation, rc
from IPython.display import HTML

```

1.1 Animación sobre el eje x

```

In [102]: fig = plt.figure(figsize=(25,7))
ax = fig.add_subplot(111)
im = ax.imshow(np.concatenate((mat[:,0:1], np.zeros((n,n*4-1)))), axis=1, cmap='Greys')
_ = ax.set_xticklabels(mcoords_to_linspace(ax.get_xticks(), 0, n))
_ = ax.set_yticklabels(mcoords_to_linspace(ax.get_yticks(), 0, n))

def animate(i):
    A = np.concatenate((mat[:,0:i], np.zeros((n,n*4-i))), axis=1)
    im.set_data(A) # update the data

```

```

    return im,

# Init only required for blitting to give a clean slate.
def init():
    return im,

ani = animation.FuncAnimation(fig, animate, np.arange(1, n*4), init_func=init,
                             interval=25, blit=True)

HTML(ani.to_html5_video())

```

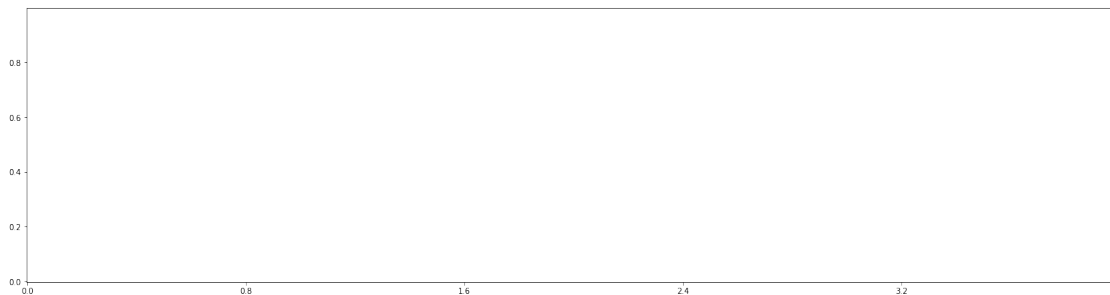
NameError Traceback (most recent call last)

```

<ipython-input-102-4bd2c53c052a> in <module>()
    15     return im,
    16
---> 17 ani = animation.FuncAnimation(fig, animate, np.arange(1, n*4), init_func=init,
    18                               interval=25, blit=True)
    19 HTML(ani.to_html5_video())

```

NameError: name 'animation' is not defined



1.2 Animación de iteración de la función

```

In [95]: fig = plt.figure()
        ax = fig.add_subplot(111)
        im = ax.scatter(mus, fn(mus, 1/2, 0))

def animate(i):
    mat = np.zeros((n, n*4))
    # calcular los f_mu(x_0)

```

```

    fmus = fn(mus, 1/2, i)
    data = np.hstack((mus[:,np.newaxis], fmus[:,np.newaxis]))
    im.set_offsets(data)
    return im,

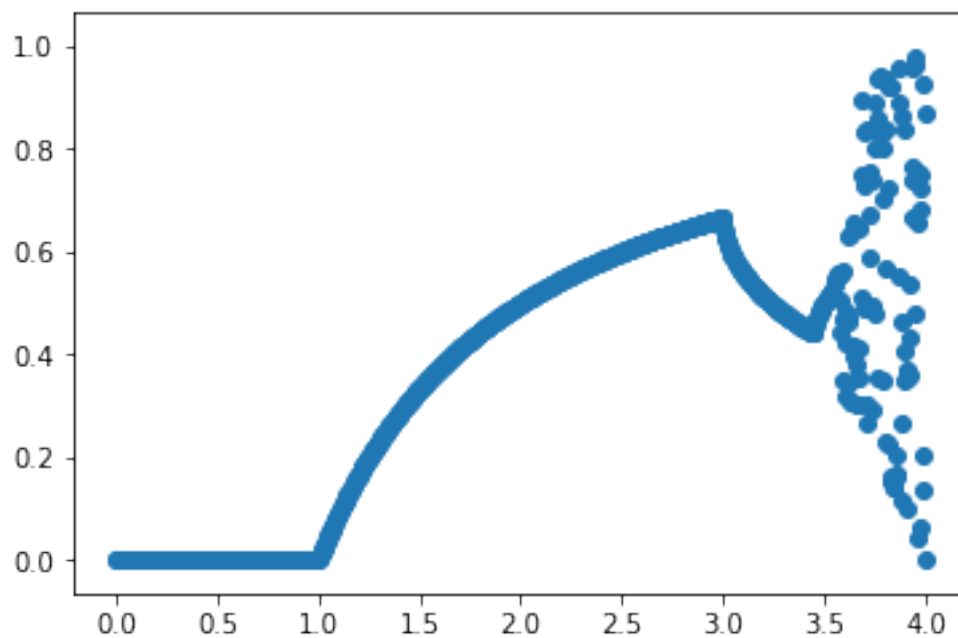
# Init only required for blitting to give a clean slate.
def init():
    return im,

ani = animation.FuncAnimation(fig, animate, np.arange(100,500), init_func=init,
                             interval=25, blit=True)

HTML(ani.to_html5_video())

```

Out [95]: <IPython.core.display.HTML object>



1.3 Generador de números aleatorios

```

In [142]: mu = 3.999
          x0 = 1/2

          x = fn(mu, x0, 400)
          #x = x0
          nums = []
          for i in range(100000):

```

```

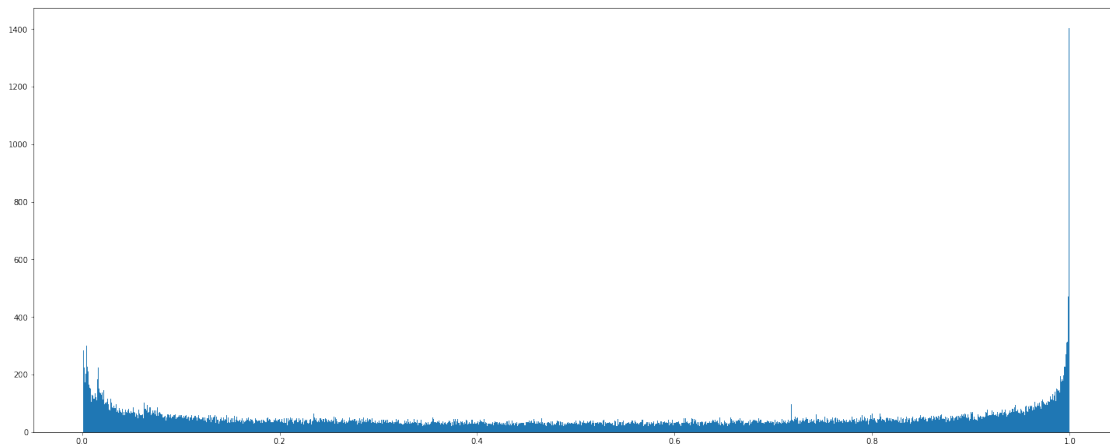
x = f(mu, x)
nums.append(x)

```

```

In [147]: fig = plt.figure(figsize=(25,10))
          ax = fig.add_subplot(111)
          ax.hist(nums, bins=2000)
          plt.savefig("dist.pdf")
          plt.show()

```



```

In [151]: nums[-10:]

```

```

Out[151]: [0.9613495492744186,
           0.14858921694377886,
           0.5059153357457021,
           0.9996100702032594,
           0.0015587212282252552,
           0.006223610173815383,
           0.024733322524028843,
           0.09646221953852391,
           0.3485418817011619,
           0.9080146931670994]

```

Usando la prueba de χ^2 podemos determinar la probabilidad de que esta distribución es tomada de una distribución uniforme

```

In [16]: import scipy.stats

```

```

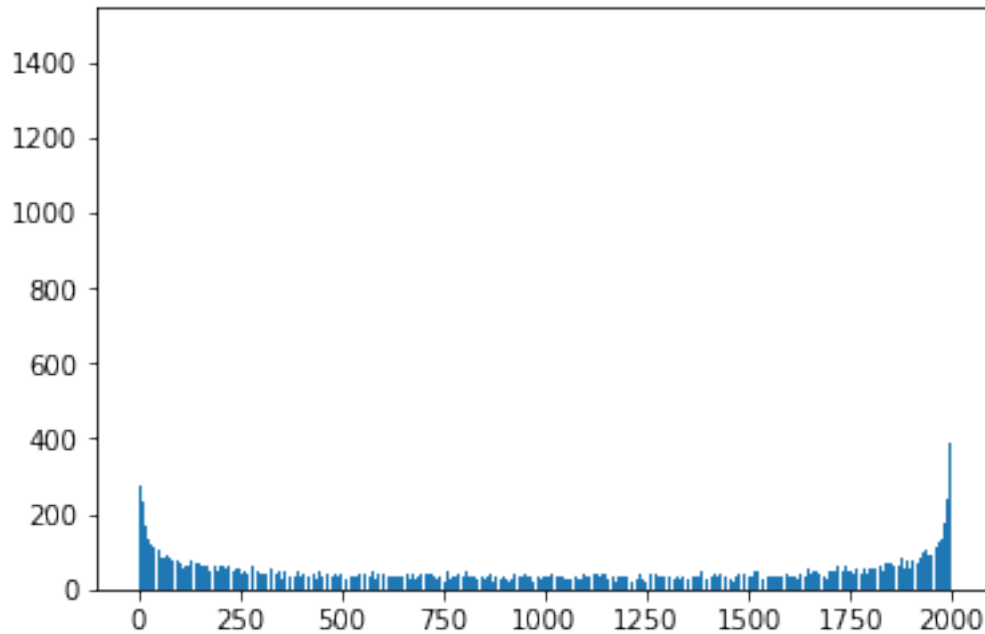
In [41]: # make a histogram
          (hist, bin_edges) = np.histogram(nums, bins=2000)

```

```

In [42]: plt.bar(range(len(hist)),hist)
          plt.show()

```



```
In [49]: observed = hist
         expected = np.ones(2000)*100000/2000
```

```
In [63]: chi_squared_stat = (((observed-expected)**2)/expected).sum()
         crit = scipy.stats.chi2.ppf(q = 0.95, df = 1999)
         p_value = 1 - scipy.stats.chi2.cdf(x=chi_squared_stat, df=1999)
         print(chi_squared_stat)
         print(crit)
         print(p_value)
```

```
115732.16000000002
2104.128222359781
0.0
```

```
In [61]: scipy.stats.chisquare(f_obs= observed,  # Array of observed counts
                               f_exp= expected)  # Array of expected counts
```

```
Out[61]: Power_divergenceResult(statistic=115732.16000000002, pvalue=0.0)
```

```
In [141]: fn(4,1/32,20)
```

```
Out[141]: 0.7030706048273905
```