HW₂

Avinash Joshi AMATH 422 Au 22 12 October 2022 In []: import matplotlib.pyplot as plt import numpy as np import scipy.optimize plt.rcParams['figure.dpi']= 200 plt.rcParams['axes.facecolor']='white' plt.rcParams['savefig.facecolor']='white' П In []: A = np.array([[0,1,5,.5], [0.5, 0, 0, 0], [0,0.9,0,0], [0,0,0.95,0]]) $n_{pop} = np.zeros((4, 51))$ n_pop[:,0] = np.array([100,100,100,100]) Tmax = 50iter_arr=np.arange(Tmax) for t in iter arr: n_pop[:,t+1]=np.matmul(A,n_pop[:,t]) In $[]: N_t = np.sum(n_pop, axis = 0)$ $w_a = np.zeros((4,51))$ for i in range(len(N t)): for j in range(4): $w_a[j,i] = n_pop[j,i]/N_t[i]$ logN = np.log(N t)lamb_lm = np.exp(np.polyfit(np.concatenate((iter_arr, np.array([50]))), logN, deg = 1)[0]) $print(f'\lambda \text{ is equal to about } \{np.round(lamb_lm, decimals = 4)\} \text{ using Leslie Matrices'})$ λ is equal to about 1.4626 using Leslie Matrices In []: label = ['\$a = 0\$', '\$a = 1\$', '\$a = 2\$', '\$a = 3\$'] fig, ax = plt.subplots(1,2) fig.tight_layout(pad = 3) $ax[0].plot(np.concatenate((iter_arr, np.array([50]))), N_t, 'k.-', ms = 3, label = '$N(t)$')$ for i in range(4): $ax[1].plot(np.concatenate((iter_arr, np.array([50]))), w_a[i,:], '.-', label = label[i])$ ax[1].legend() fig.supxlabel('Time') ax[0].set_yscale('log') ax[0].set_ylabel('log\$N(t)\$') $ax[1].set_ylabel('$w_a(t) = \dfrac{n_a(t)}{N(t)}$')$

ax[0].set_title('Total population', fontsize = 8)
ax[1].set_title('Fractional population', fontsize = 8)

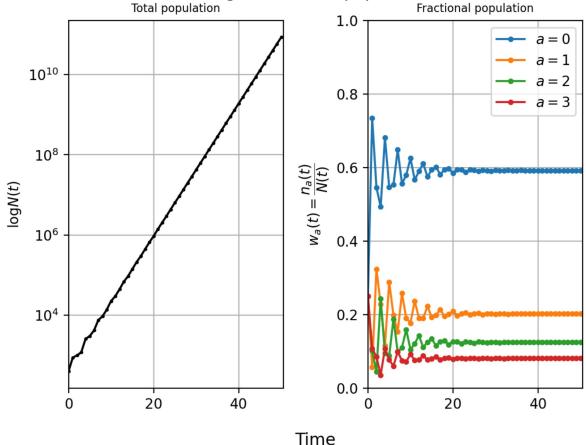
ax[0].grid()
ax[1].grid()

plt.show()

ax[0].set_xlim([0,50.5])
ax[1].set_xlim([0,50.5])
ax[1].set_ylim([0,1])

fig.suptitle('Evolution of age-structured population with \$A=3\$')

Evolution of age-structured population with A = 3



```
def eulot_func(lam,Ia_arr,fa_arr):
    """compute the Euler-Lotka sum, taking as arguments a scalar and two 1-D numpy arrays"""
    length_of_array=Ia_arr.size
    age_arr=np.arange(0,length_of_array)
    temp_arr=lam**(-(age_arr+1))*Ia_arr*fa_arr
    return sum(temp_arr) -1

Ia_arr=np.array([1,.5,.5*.9,.5*.9*.95])
fa_arr=np.array([0,1,5,.5])
```

```
In []: lambda_min=0.3
lambda_max=3

lambda_arr=np.linspace(lambda_min,lambda_max,100)

G_arr=np.zeros(lambda_arr.size)

iter_arr=np.arange(lambda_arr.size)

for j in iter_arr:
    G_arr[j]=eulot_func(lambda_arr[j],Ia_arr,fa_arr)

left_bracket=0.1
    right_bracket=5
    args=(Ia_arr,fa_arr)

lamb_el = scipy.optimize.brentq(eulot_func,left_bracket,right_bracket,args)

print(f'\lambda is equal to about {np.round(lamb_el, decimals = 4)} using the Euler-Lokta Formula')
```

 λ is equal to about 1.4624 using the Euler-Lokta Formula

My predictions for λ from both the Euler-Lokta formulas and the Leslie matrix simulations are accurate from each other to 3 decimal places, i.e., $\lambda=1.462$. This is to be expected because both methods are valid ways of numerically approximating the growth rate of the population.

It does not matter what specific p_a value I chose for $a \in [0,2]$ as long as $\prod_{a=0}^2 p_a = I_3$ because the survival probabilites are eventually multiplied together to get whatever probability that an owl will live to age 3. If $p_0=0.1$ and $p_1=0.5$, $I_2=0.05$, which is the same as $p_0=0.5$ and $p_1=0.1$, or if 10% of aged zero survives then 50% of that remaining 10% survive to age 2 is equal to 10% of the 50% of the surviving zero-year-olds surviving.

If $n_0(0)=N$, then after one year that same group will have $p_0\times N$ individuals, then after another year $p_1p_0\times N$ individuals. Looking at the multiplication being done, one could simply switch the numbers of p_1 and p_0 and they would still give the same product because of the commutative nature of multiplication.

```
In []: A = np.zeros((51,51))
        A[1,0] = 0.361
        A[2,1] = 0.4
        A[3,2] = 0.5
        for i in range(51):
            if (i<50) and (A[i+1,i] == 0):
                A[i+1,i] = 0.942
            if i>=3:
                A[0,i] = 0.24
        print(f'A matrix: \n {A}\n')
        print(f'p_0*p_1*p_2 = \{A[1,0]*A[2,1]*A[3,2]\}')
        A matrix:
         [[0. 0.
                     0. ... 0.24 0.24 0.24 ]
         [0.361 0. 0. 0. 0. 0. ]
         [0.
              0.4 0.
                            ... 0.
                                      0.
                                                 1
                0.
                            ... 0. 0. 0.
         [0.
                      0.
                            ... 0.942 0. 0.
         [0.
                0.
                      0.
                            ... 0.
                                      0.942 0.
        p_0*p_1*p_2 = 0.0722
In [ ]: 1,v = np.linalg.eig(A)
        lamb_owl = np.abs(1).max()
        print(f'The dominant \lambda, i.e., the long-term growth rate, is \lambda = \{\text{np.round}(\text{lamb\_owl}, \text{decimals} = 4)\}
        The dominant \lambda, i.e., the long-term growth rate, is \lambda = 0.9439
```