

Problem set 2, Task 2
Diffusion driven instability
Computational Biology
FFR110/FIM740

MATTIAS BERG, KONSTANTINOS ZAKKAS

February 26, 2021

$$\frac{\partial u}{\partial t} = a - (b+1)u + u^2v + D_u \nabla^2 u \quad (1)$$

$$\frac{\partial v}{\partial t} = bu - u^2v + D_v \nabla^2 v \quad (2)$$

a.)

We want to find the steady states of the system. We neglect the diffusion part (setting D_u and D_v to zero). To do find the fixed point we sett $\frac{\partial u}{\partial t} = \frac{\partial v}{\partial t} = 0$. We start with finding v in terms of u using Equation 2

$$\begin{aligned} 0 &= bu - u^2v \\ u^2v &= bu \\ v &= \frac{b}{u} \end{aligned}$$

We put this into Equation 1

$$\begin{aligned} 0 &= a - (b+1)u + u^2 \left(\frac{b}{u} \right) \\ 0 &= a - bu - u + bu \\ u^* = a &\Rightarrow v^* = \frac{b}{a} \end{aligned} \quad (3)$$

The linear stability then is determined by the eigenvalues λ of the linearised Jacobian matrix

$$A = \begin{vmatrix} -1 - b + 2uv & u^2 \\ b - 2uv & -u^2 \end{vmatrix} \quad (4)$$

Thus for $u = u^*$ and $v = v^*$ we get

$$A = \begin{vmatrix} b-1 & a^2 \\ -b & -a^2 \end{vmatrix} \quad (5)$$

The steady state will be stable when the following conditions are satisfied

$$\begin{aligned} \text{tr} A &< 0 \\ \det A &> 0 \end{aligned}$$

For the first condition we have

$$\begin{aligned} b-1-a^2 &< 0 \\ b &< 1+a^2 \end{aligned} \quad (6)$$

and for the second condition

$$\begin{aligned} -(b-1)a^2 - (-ba^2) &> 0 \\ -2ba^2 + a^2 + ba^2 &> 0 \\ a^2 &> 0 \end{aligned} \quad (7)$$

This gives us that we have a stable steady state if $a \neq 0$ and $b < 1 + a^2$, as found in Equation 6 and 7

b.)

We want to find when D_v on the the homogeneous stable steady state have a diffusion-driven instability. We use equation (8) and (9) from lecture note 7 (CompBioLecture7.pdf) to find when the waves become unstable;

$$(dJ_{11} + J_{22})^2 - 4d \det \mathbb{J} \geq 0 \quad (8)$$

$$dJ_{11} + J_{22} > 0 \quad (9)$$

using equation 5 and 7 we get the following parameter values;

$$J_{11} = \frac{\partial \frac{\partial u^*}{\partial t}}{\partial u} = b-1 \Rightarrow b=8 \Rightarrow J_{22}=7 \quad (10)$$

$$J_{22} = \frac{\partial \frac{\partial v^*}{\partial t}}{\partial v} = -a^2 \Rightarrow a=3 \Rightarrow J_{22}=-9 \quad (11)$$

$$\det \mathbb{J} = a^2 \Rightarrow a=3 \Rightarrow \det \mathbb{J} = 9 \quad (12)$$

inserting the parameters in equation 8, gains the following (point when it becomes unstable);

$$\begin{aligned}
 (d * 7 - 9)^2 - 4d * 9 &= 0 \\
 49d^2 - 126d + 81 - 36d &= 0 \\
 49d^2 - 162d + 81 &= 0 \\
 d &= \frac{-(-162) \pm \sqrt{(-162)^2 - 4 * 49 * 81}}{2 * 49} \\
 d &= \frac{162 \pm \sqrt{26244 - 15876}}{98} \\
 d &= \frac{162 \pm \sqrt{10368}}{98} \\
 d &= \frac{162 \pm 101.823}{98} \\
 d_1 &\approx 2.692 \\
 d_2 &\approx 0.614
 \end{aligned}$$

Using the gained d_1 and d_2 we need to check if they fulfills equation 9 also;

$$\begin{aligned}
 d_1 * 7 - 9 &> 0 \Rightarrow 18.844 > 9 \Rightarrow \text{True} \\
 d_2 * 7 - 9 &> 0 \Rightarrow 4.298 > 9 \Rightarrow \text{Not true}
 \end{aligned}$$

Hence the diffusion-driven instability start at d_1 and goes to ∞ . As D_u is set to 1, D_v value is the same as d_1 for this case.

c.)

For $D_v = (2.3, 5, 9)$ iterations numbers 100/1 000 was chosen as transient/stable state for the figures, larger iterations (tests up to 10 000 iteration gave no visible difference), see Figure 1, 3 and 4. $D_v = 3$ have some difference in the pattern between 1000 and 10 000 iterations, and we can see that some part have not yet formed after 1 000 iterations compared to the stable case after 10 000 iterations, and therefor 1 000/10 000 iterations was chosen for the Transient/stable state for Figure 2.

$D_v = 2.3$ in Figure 1 we see a homogeneous distribution, which is expected based on the calculation in b.) where the braking point was found to be $D_v \approx 2.692$. The other simulations. With D_v set above $D_v \approx 2.692$, we can see a clear inhomogeneous distribution in Figure 2, 3 and 4 at their stable state. We can also see that large D_v gives a larger variance between the two chemicals.

The close one is to the stable diffusion, the longer it seems to take for the system to become stable. In Figure 5, showing a D_v value close to where the diffusion-driven instability starts, the stable state pattern wont be clearly seen

until iterations somewhere between 10 000 and 100 000, compared to the systems with higher values of D_v that only need somewhere between 100 and 1 000 iterations.

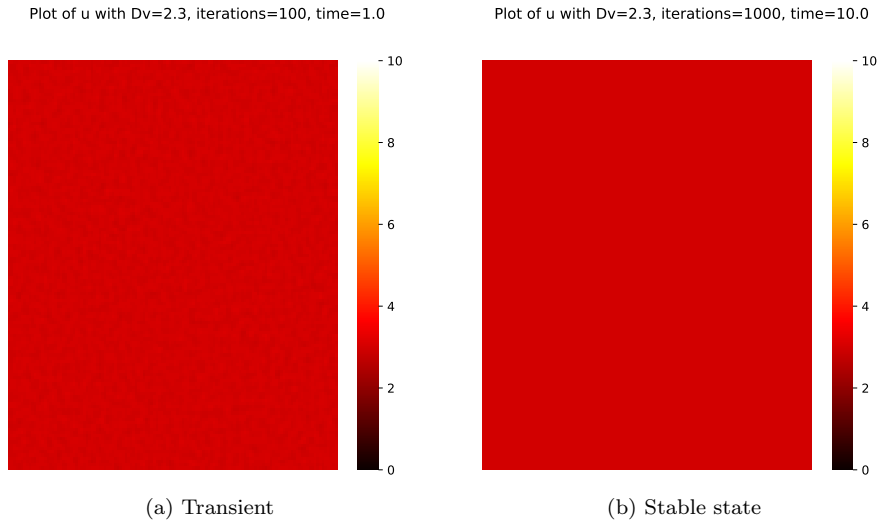
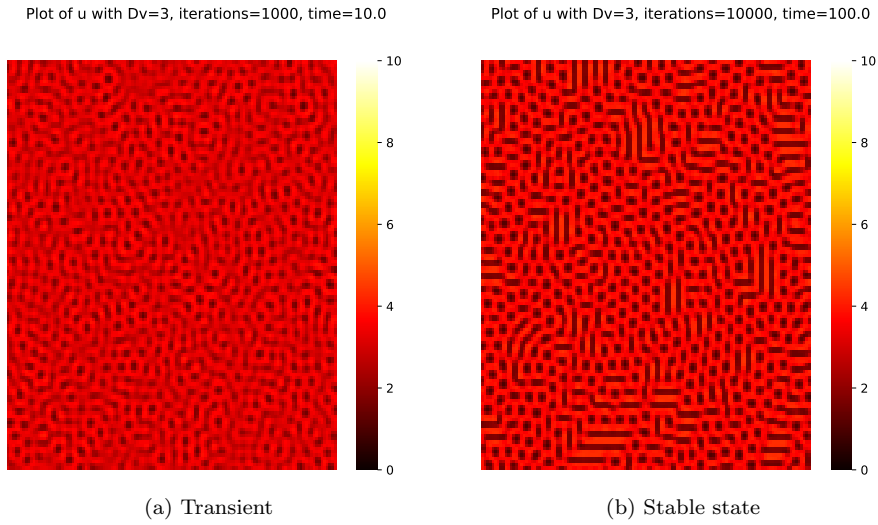
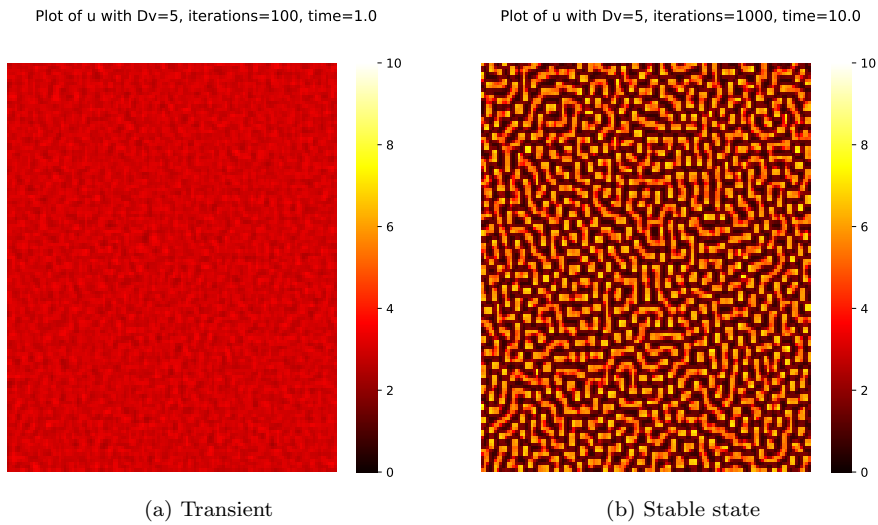


Figure 1: Spatial distribution of u in form of heatmaps for $D_v = 2.3$

Figure 2: Spatial distribution of u in form of heatmaps for $D_v = 3$ Figure 3: Spatial distribution of u in form of heatmaps for $D_v = 5$

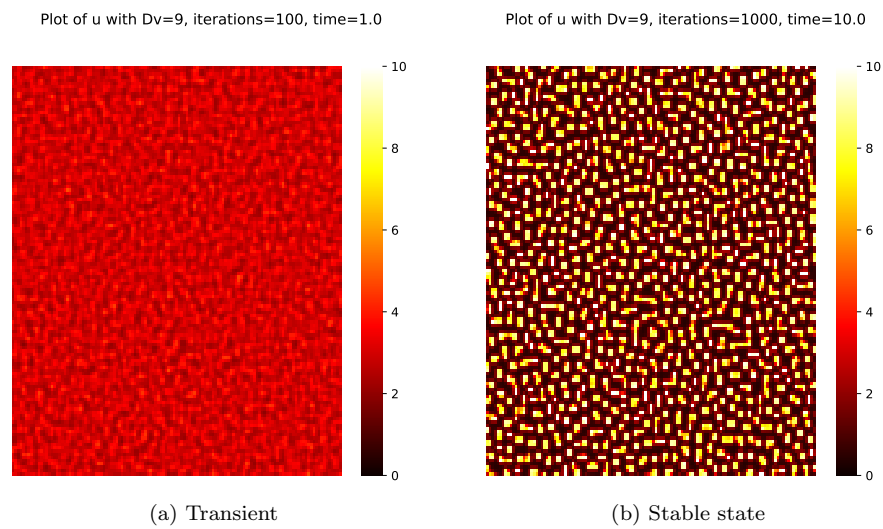


Figure 4: Spatial distribution of u in form of heatmaps for $D_v = 9$

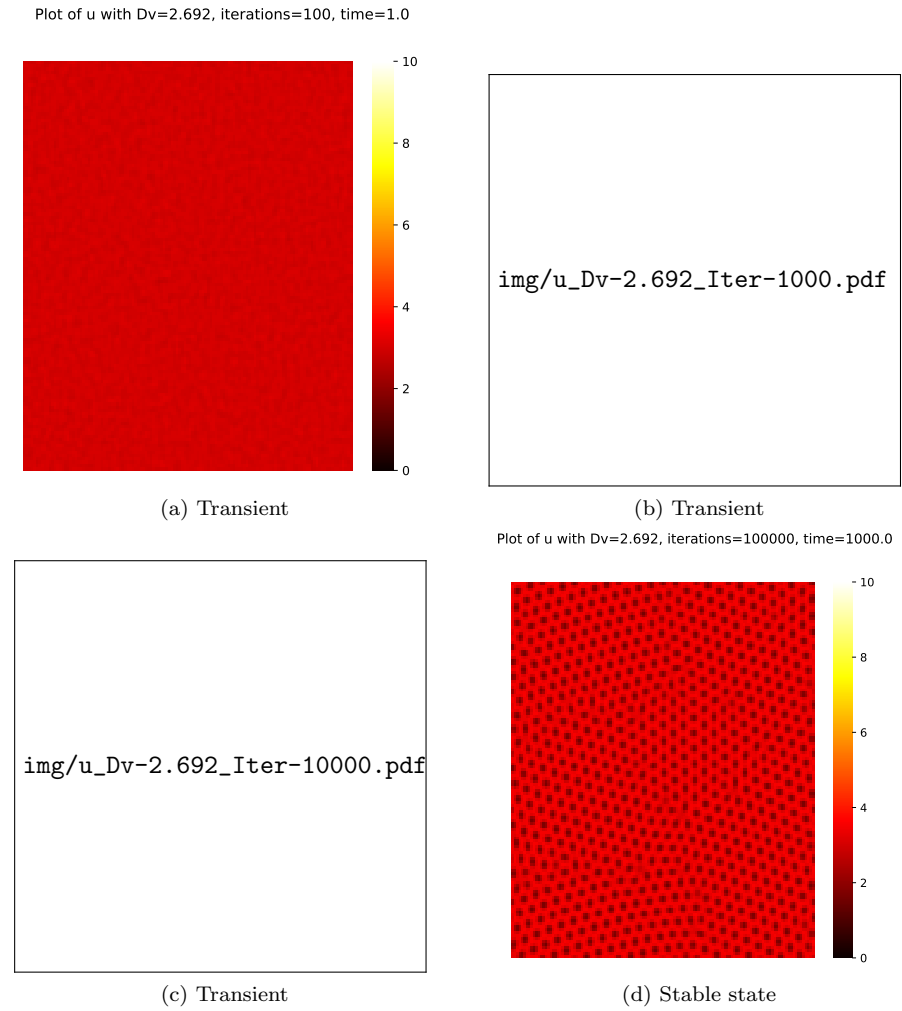


Figure 5: Spatial distribution of u in form of heatmaps at point close to diffusion-driven instability ($D_v = 2.692$)

Appendix


```

#Course : FFR110/FIM740 Computational Biology
#Problem : Problem set 2, Task 2. Diffusion driven instability
#Code : Python 3.8.5

import numpy as np
import matplotlib.pyplot as plt
import math
import seaborn as sns
from numpy import genfromtxt

import time

def Calc_BelousovZhabotinsky(a, b, Du, Dv, u, v, dTime = 0.01):
    yLenght, xLenght = u.shape
    u_new = np.zeros((yLenght, xLenght))
    v_new = np.zeros((yLenght, xLenght))

    h = 1

    for y in range(1, yLenght-1):
        for x in range(1, xLenght-1):
            LaplacianU = (u[y, x+h] + u[y, x-h] + u[y+h, x] + u[y-h, x] - 4*u[y, x]) / (h**2)
            LaplacianV = (v[y, x+h] + v[y, x-h] + v[y+h, x] + v[y-h, x] - 4*v[y, x]) / (h**2)

            dudt = a - (b+1)*u[y, x] + v[y, x]*u[y, x]**2 + Du*LaplacianU
            dvdt = b*u[y, x] - v[y, x]*u[y, x]**2 + Dv*LaplacianV

            u_new[y, x] = u[y, x] + dTime*dudt
            v_new[y, x] = v[y, x] + dTime*dvdt

            if x == 1:
                u_new[y, -1] = u_new[y, x]
                v_new[y, -1] = v_new[y, x]
            elif x == yLenght-2:
                u_new[y, 0] = u_new[y, x]
                v_new[y, 0] = v_new[y, x]

            if y == 1:
                u_new[-1, x] = u_new[y, x]
                v_new[-1, x] = v_new[y, x]
            elif y == yLenght-2:
                u_new[0, x] = u_new[y, x]
                v_new[0, x] = v_new[y, x]

    return u_new, v_new

def SaveU(Dv, u, Iteration, dTime = 0.01):
    fileName = "u_Dv-%s_Iter-%s.csv" % (Dv, Iteration)
    np.savetxt(fileName, u, delimiter=";")

    vmin=0
    vmax=10

    fig, ax = plt.subplots(figsize=(6,6))

    titleString = "Plot_of_u_with_Dv=%s, iterations=%s, time=%s" % (Dv, Iteration, Iteration*dTime)
    fig.suptitle(titleString)
    ax = sns.heatmap(u[1:-2, 1:-2], cmap="hot", center=(vmin + vmax) / 2, vmin=vmin, vmax=vmax, yticklabels=False, xticklabels=False)

    graphName = "u_Dv-%s_Iter-%s.pdf" % (Dv, Iteration)
    plt.savefig(graphName, bbox_inches='tight')

def SaveV(Dv, v, Iteration, dTime = 0.01):
    fileName = "v_Dv-%s_Iter-%s.csv" % (Dv, Iteration)
    np.savetxt(fileName, v, delimiter=";")

    vmin=0
    vmax=10

    fig, ax = plt.subplots(figsize=(6,6))

    titleString = "Plot_of_v_with_Dv=%s, iterations=%s, time=%s" % (Dv, Iteration, Iteration*dTime)
    fig.suptitle(titleString)
    ax = sns.heatmap(v[1:-2, 1:-2], cmap="jet", center=(vmin + vmax) / 2, vmin=vmin, vmax=vmax, yticklabels=False, xticklabels=False)

    graphName = "v_Dv-%s_Iter-%s.pdf" % (Dv, Iteration)
    plt.savefig(graphName, bbox_inches='tight')

def RunProgram(Dv, Iteration):
    xLenght = 128
    yLenght = 128
    a=3
    b=8

```

```
Du=1

dTime = 0.01

#Add two to the lenght to create space on each side of the matrix (for edge cases)
u = a+(np.random.rand(yLenght+2,xLenght+2)*2-1)*0.1
v = b/a+(np.random.rand(yLenght+2,xLenght+2)*2-1)*0.1
for t in range(Iteration):
    u, v = Calc_BelousovZhabotinsky(a, b, Du, Dv, u, v, dTime=dTime)
    if not t%10:
        tString = "Dv=%s, _iterations_%s_out_of_%s" % (Dv, t, Iteration)
        print(tString)
    SaveU(Dv, u, Iteration)

Iteration = 100
Dv=2.3
RunProgram(Dv, Iteration)

Iteration = 1000
Dv=2.3
RunProgram(Dv, Iteration)

Iteration = 10000
Dv=2.3
RunProgram(Dv, Iteration)
```