

Loglinearizált RBC modell - impulzus-vezáasz függvények Uhlig algoritmussal

Definialjuk a parametereket

```
In [1]: # Eloadason kalibrált parameterek  
alpha = 0.44  
beta = 0.964  
delta = 0.054  
# Preferencia parameterek a szakirodalomnak megfelelően  
theta = 1.5  
psi = 0  
rho = 0.85  
sigma = 0.05  
  
omega = alpha*beta/(1-beta*(1-delta));
```

Definialjuk a szükséges együttható mátrixokat

Az együttthatómátrixok:

$$A = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ -(1-\delta) \\ -\alpha \\ 0 \\ -\alpha\beta/\omega \\ 0 \end{bmatrix} \quad D = \begin{bmatrix} 0 \\ 0 \\ -1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$C = \begin{bmatrix} 1 & 0 & -(1-\delta\omega) & -\delta\omega & 0 & 0 \\ 0 & 0 & 0 & -\delta & 0 & 0 \\ 1 & -(1-\alpha) & 0 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 & -1 \\ \alpha\beta/\omega & 0 & 0 & 0 & -1 & 0 \\ 0 & \psi & \theta & 0 & 0 & -1 \end{bmatrix}$$

$$F = [0] \quad G = [0] \quad H = [0] \quad L = [0] \quad M = [0]$$

$$J = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ -1/\theta \\ 0 \end{bmatrix} \quad K = \begin{bmatrix} 0 \\ 0 \\ -1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad \text{eml: } y_t = \begin{bmatrix} y_t \\ l_t \\ c_t \\ i_t \\ \hat{r}_t \\ \hat{w}_t \end{bmatrix}$$

In [2]:

```
import numpy as np

A = np.matrix([0, 1, 0, 0, 0, 0]).T
B = np.matrix([0, -(1-delta), -alpha, 0, -alpha*beta/omega, 0]).T
C = np.matrix([[1, 0, -(1-delta*omega), -delta*omega, 0, 0],
               [0, 0, 0, -delta, 0, 0],
               [1, -(1-alpha), 0, 0, 0, 0],
               [1, -1, 0, 0, 0, -1],
               [alpha*beta/omega, 0, 0, 0, -1, 0],
               [0, psi, theta, 0, 0, -1]])
D = np.matrix([0, 0, -1, 0, 0, 0]).T
F = np.matrix([0])
G = np.matrix([0])
H = np.matrix([0])
J = np.matrix([0, 0, 1, 0, -1/theta, 0])
K = np.matrix([0, 0, -1, 0, 0, 0])
L = np.matrix([0])
M = np.matrix([0])
N = np.matrix([rho])
```

Számítsuk ki a megfelelő együttthatómátrixokat

$$\begin{aligned}\Psi &= F - JC^{-1}A \\ \Gamma &= JC^{-1}B + KC^{-1}A - G \\ \Theta &= KC^{-1}B - H \quad \Rightarrow \\ \Xi &= \begin{bmatrix} \Gamma & \Theta \\ I_m & 0_m \end{bmatrix} \\ \Delta &= \begin{bmatrix} \Psi & 0_m \\ 0_m & I_m \end{bmatrix}\end{aligned}$$

```
In [3]: # Matrixok merete
m, ee = F.shape
n, de = C.shape
k, xe = N.shape

# Matrixok definialasa
Cinv = C.I
PSI = F - J*Cinv*A;
GAMMA = J*Cinv*B + K*Cinv*A - G;
THETA = K*Cinv*B - H;
XI = np.matrix(np.concatenate((
    np.concatenate((GAMMA, THETA), axis = 1),
    np.concatenate((np.identity(m), np.zeros((m,m))), axis=1)), axis=0))
DELTA= np.matrix(np.concatenate((
    np.concatenate((PSI, np.zeros((m,m))), axis = 1),
    np.concatenate((np.zeros((m,m)), np.identity(m)), axis=1)), axis=0))
```

Generalizált sajátérték probléma megoldása

```
In [4]: from scipy.linalg import eig
eigvals, eigvecs = eig(XI, DELTA)
```

```
In [5]: # Ellenorizzuk
```

```
In [6]: XI*eigvecs
```

```
Out[6]: matrix([[ 1.7627244 , -1.21587394],
                [-0.75562017,  0.66864544]])
```

```
In [7]: DELTA*eigvecs*np.diag(eigvals)
```

```
Out[7]: matrix([[ 1.7627244 +0.j, -1.21587394+0.j],
                [-0.75562017+0.j,  0.66864544+0.j]])
```

```
In [8]: eigvals
```

```
Out[8]: array([1.15360091+0.j, 0.89922294+0.j])
```

```
In [9]: eigvecs
```

```
Out[9]: array([[ -0.75562017,  0.66864544],  
              [ -0.65501004,  0.74358139]])
```

```
In [10]: # 0.8992 lesz a stabil sajátérték  
LAMBDA = np.matrix([abs(eigvals[1])])  
OMEGA = np.matrix([abs(eigvecs[1,1])])
```

Parameterek kiszamitasa:

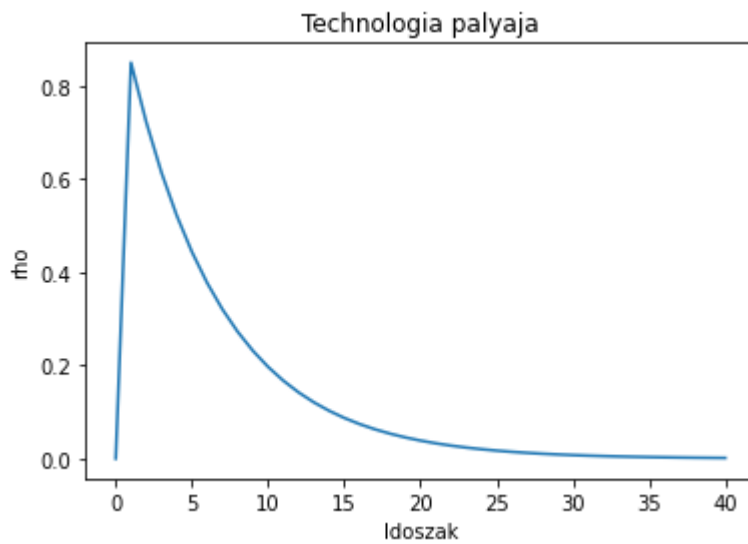
$$\begin{aligned}0 &= AP + B + CR \\0 &= (FP + G + JR)P + H + KR \\0 &= AQ + CS + D \\0 &= (FP + G + JR)Q + (FQ + JS + L)N + KS + M\end{aligned}$$

```
In [11]: P = OMEGA*LAMBDA/OMEGA  
R = -Cinv*(A*P+B)  
  
V = np.concatenate((  
    np.concatenate((np.kron(np.identity(k),A),  
        np.kron(np.identity(k),C)), axis=1),  
    np.concatenate((np.kron(N.T,F)+np.kron(np.identity(k),(F*P+J*R+G)),  
        np.kron(N.T,J)+np.kron(np.identity(k),K)), axis=1)), axis=0)  
LN_plus_M = L*N + M;  
QS_vec = (-V).I*np.concatenate((D, LN_plus_M), axis=0)  
Q = QS_vec[0]  
S = QS_vec[1:]
```

Impulzus-valasz fuggvenyek

```
In [12]: import matplotlib.pyplot as plt  
  
# A technologia egy szorasnyi sokkot kap  
nper = 40;  
z = [0, rho]  
for i in range(nper-1):  
    z.append(rho*z[-1])  
  
fig = plt.figure()  
axs = plt.axes()  
axs.plot(range(len(z)),z)  
axs.set_title('Technologia palyaja')  
axs.set_xlabel('Idoszak')  
axs.set_ylabel('rho')
```

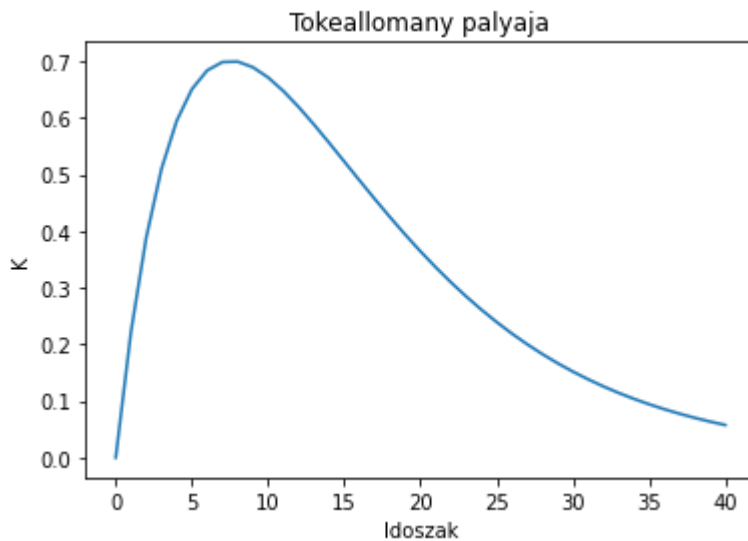
```
Out[12]: Text(0, 0.5, 'rho')
```



```
In [16]: # Az allapotvaltozo (k) palyaja
K = [0]
for t in range(nper):
    K.append(P*K[t] + Q*z[t+1])

fig = plt.figure()
axs = plt.axes()
axs.plot(range(len(K)),K)
axs.set_title('Tokeallomany palyaja')
axs.set_xlabel('Idoszak')
axs.set_ylabel('K')
```

Out[16]: Text(0, 0.5, 'K')



```
In [14]: # A tobbi endogen valtozo palyaja
y = np.zeros((nper+1, n))
y[1,:] = R.T*K[0] + S.T*z[1]
for t in range(nper-1):
    y[t+2,:] = R.T*K[t+1][0,0] + S.T*z[t+2]

legend_y = ['y', 'n', 'c', 'i', 'r', 'w']
```

```

for i in range(n):
    fig = plt.figure()
    axs = plt.axes()
    axs.plot(range(len(y[:,i])),y[:,i])
    axs.set_title(legend_y[i])
    axs.set_xlabel('Idoszak')
    axs.set_ylabel(legend_y[i])

```

