# CapitalistConsumption

December 3, 2019

#### 1 TODO

- ☐ Plug inflation in housing stock☐ Add other autonomous expenditure
- [1]: #!pip install pysolve3

Time checking:

[2]: from datetime import datetime
t1 = datetime.now()

#### 2 Introduction

Jupyter notebook for the simplest model scratch. By convention, all exogenous variables are presented with a overline line ( $e.g. \bar{a}$ ) and the parameters are represented by greek letters ( $e.g. \alpha$ ). The code in this document is executable and is strongly recommend to do the follow to ensure that the output is corrected and updated:

Run time > Restart and run all...

# 3 Changes

In this version, the households are splitted in workers and capitalists where workers spend all their income white capitalists spend a share o distributed profits.

# 4 Equations

#### 4.1 General Equations

$$Y = C + I$$

$$C = C_w + C_k$$

$$I = I_f + I_h$$

$$\omega=\overline{\omega}$$

$$W = \omega \cdot Y$$

$$Y_K = \frac{K_f}{\overline{v}}$$

$$u = \frac{Y}{Y_K}$$

$$g_k = \frac{h \cdot u}{v}$$

$$Z = I_h$$

$$K = K_f + K_{HD}$$

### 4.2 Households

$$C_w = \alpha \cdot W$$

$$C_k = \alpha_2 \cdot FD$$

$$YD_w = W$$

$$YD_k = FD + \overline{r}_{m-1} \cdot M_{-1} - \overline{r}_{mo-1} \cdot MO$$

$$Sh_w = YD_w - C_w = 0$$

$$Sh_k = YD_k - C_k = \Delta M$$

$$\Delta MO = I_h$$

$$V_h = M + K_H - MO$$

$$NFW_h = S_h - I_h$$

$$Residual = \Delta M - \Delta L_f$$

# 4.3 Firms

$$\Delta L_{f} = I_{f} - FU$$

$$FT = Y - W = FU + FD$$

$$FU = \gamma_{F} \cdot (FT - r_{L-1} \cdot L_{f-1})$$

$$FD = (1 - \gamma_{F}) \cdot (FT - r_{L-1} \cdot L_{f-1})$$

$$I_{f} = h \cdot Y$$

$$\Delta K_{f} = I_{f}$$

$$\Delta h = h_{-1} \cdot \gamma_{u} \cdot (u - \overline{u}_{n})$$

$$V_{f} = K_{f} - L_{f}$$

$$NFW_{f} = FU - I_{f}$$

### 4.4 Banks

$$L = L_f$$
 
$$\Delta M = \Delta L + \Delta MO$$
 
$$NFW_b = rl_{-1} \cdot L_{-1} + rmo_{-1} \cdot MO_{-1} - rm_{-1} \cdot M_{-1}$$
 
$$r_l = r_m + spread_l (= 0)$$
 
$$r_{mo} = r_m + spread_{mo} (= 0)$$
 
$$V_b = L + MO - M$$

#### 4.5 Residential Investment

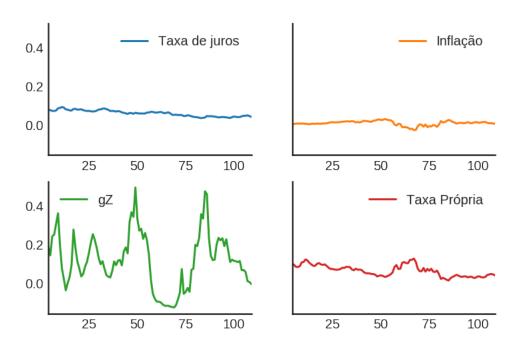
$$K_{HS} = K_{HD}$$
 $I_h^S = I_h$ 
 $\Delta K_{HD} = I_h$ 
 $I_h = (1 + g_Z) \cdot I_{h-1}$ 
 $k = \frac{K_h \cdot p_h}{K}$ 
 $g_Z = \phi_0 - \phi_1 \cdot own$ 
 $own = \left(\frac{1 + r_{mo}}{1 + \dot{p}}\right) - 1$ 

## 5 Simulation setup

#### 5.1 Loading libraries

```
[3]: %config InlineBackend.figure_format = 'retina'
    import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import matplotlib.patheffects as pe
    import seaborn as sns
    import networkx as nx
    import sympy as sp
    from sympy import pprint, cse
    import ipywidgets as widgets
    from ipywidgets import interact, interactive, fixed, interact_manual
    plt.style.use('seaborn-white')
    from pysolve3.model import Model
    from pysolve3.utils import SolveSFC, ShockModel, SummaryShock, SFCTable
[4]: data = pd.read_csv('../SeriesTemporais/Dados_yeojohnson.csv')
    data = data[[
        "Taxa de juros",
        "Inflação",
```

/home/gpetrini/.local/lib/python3.6/sitepackages/pandas/plotting/\_matplotlib/\_\_init\_\_.py:62: UserWarning: To output
multiple subplots, the figure containing the passed axes is being cleared
 plot\_obj.generate()



## 6 Creating model function

```
[7]: def model(
        alpha = 1,
        alpha_2 = .3,
        gamma_F = 0.9,
        gamma_u = 0.01,
        g_Z = 0.05,
        omega = 0.3,
        rm = 0.02,
        spread 1 = 0,
        spread_mo = 0,
        un = 0.8,
        v = 2.5,
        phi 0 = 0.04,
        phi_1 = 0.02,
        infla = 0.0,
        phparam=1.0
    ):
      phparam: 1.0 means no inflation
     model = Model()
     model.set_var_default(0)
     model.var('C', desc='Total Consumption')
     model.var('C w', desc='Workers Consumption')
     model.var('C_k', desc='Capitalists Consumption')
     model.var('FD', desc='Distributed profits')
     model.var('Fn', desc='Net profits')
     model.var('FT', desc='Total Profits')
     model.var('FU', desc='Retained profits')
     model.var('gk', desc='Capital growth rate', default=0.01)
     model.var('g_Z', desc='Autonomous grouth rate', default=g_Z)
      model.var('h', desc='Marginal propensity to invest (non-residential)', u
     \rightarrowdefault=0.30)
      model.var('I_t', desc='Investment', default = 100) # 200
     model.var('I f', desc='Non-residential investment') # 100
     model.var('I_h', desc='Residential investment', default = 100) # 100
      model.var('Is', desc='Residential investment (Supply)', default = 100) # 100
     model.var('K_HS', desc='Houses supply', default=500) # 500
     model.var('K_HD', desc='Houses demand', default=500) # 500
     model.var('K_f', desc='Non-residential capital', default = 1000) # 10000
     model.var('Knom', desc='Nominal Capital', default=1500)
     model.var('K', desc='Real Capital', default=1500)
     model.var('K_k', desc="% of Kf in total")
      model.var('L', desc='Total Loans') # 100
      model.var('Lf', desc='Firms Loans') # 100
```

```
model.var('M', desc='Money deposits', default = 310) # 300
model.var('MO', desc='Mortgages', default = 300) # 200
model.var('NFW_h', desc='Households Net Financial Wealth')
model.var('NFW_f', desc='Firms Net Financial Wealth')
model.var('NFW_b', desc='Banks Net Financial Wealth')
model.var('own', desc='Own interest rate', default = 0.02)
model.var('ph', desc='House price', default = 1)
model.var('rl', desc='Interests rates on loans')
model.var('rmo', desc='Interests rates on mortgages')
model.var('Sh_w', desc='Workers savings', default = 0)
model.var('Sh_k', desc='Capitalist savings', default = 100)
model.var('u', desc='Capacity utilization ratio', default=0.7)
model.var('V_h', desc='Household net financial wealth', default = 500)
model.var('V_f', desc='Firms net financial wealth', default = 1000)
model.var('V b', desc='Banks net financial wealth')
model.var('W', desc='Wages')
model.var('Y', desc='GDP')
model.var('Yk', desc='Capacity')
model.var('YD_w', desc='Workers disposable income')
model.var('YD_k', desc='Capitalists disposable income')
model.var('Z', desc='Autonomous expenditures')
model.var('Residual', desc='Unecessarily equation. Should be zero')
model.param('alpha', desc='Propensity to consume out of wages',,,
→default=alpha) # 1
model.param('alpha_2', desc='Propensity to consume out of profits', __

default=alpha_2) # 0.6

model.param('gamma_F', desc='% of undistributed profits', default=gamma_F) #__
→0.4
model.param('gamma_u', desc='Adjustment parameter for the marginal propensity_
→to invest', default=gamma_u) # 0.01
model.param('omega', desc='Wage-share', default = omega)
model.param('rm', desc='Interest rates on money deposits', default=rm) # 0.02
model.param('spread_1', desc='Spread for loans', default=spread_1) # 0.01
model.param('spread mo', desc='Spread for mortgages', default=spread mo) # 0.
→005
model.param('un', desc='Normal capacity utilization ratio', default=un) # 0.8
model.param('v', desc='Capitl-Output ratio', default=v) # 2.5
model.param('phi 0', desc='Autonomous housing investment component',default = ___
→phi_0)
model.param('phi_1', desc='Housing investment sensitivity to own interest⊔
→rate', default = phi_1)
model.param('phparam', desc='ph parameter', default = phparam)
model.param('infla', desc='infla value', default = infla)
model.param('gZn', desc='Growth rate new autonomous expenditure',
            default=phi_0 - phi_1*data['Taxa Própria'].iloc[-1])
```

```
# General equations
model.add('Y = C + I_t') # Eq1
model.add('C = C_w + C_k')
model.add('I_t = I_f + I_h') # Eq2
model.add('Yk = K_f(-1)/v') # Eq 4
model.add('u = Y/Yk') # Eq 5
model.add('W = omega*Y') # Eq 6
model.add('gk = h*u/v') # Eq 7
model.add('Knom = K HD*ph + K f')
model.add('K = K_HD + K_f')
model.add('Z = I_h') # Eq 9
# Household equations
model.add('YD_k = FD + rm*M(-1) - rmo*MO(-1)') # Eq 10
model.add('d(MO) = I_h') # Eq 12
model.add('V_h = M + K_HD*ph - MO') # Eq 15
# Firms
model.add('d(Lf) = I_f - FU') # Eq 15
model.add('FT = (1-omega)*Y') # Eq 16
model.add('Fn = FT -rl*Lf(-1)')
model.add('FU = gamma_F*(Fn)') # Eq 17
model.add('FD = (1 - gamma_F)*(Fn)') # Eq 18
model.add('I_f = h*Y') # Eq 19
model.add('d(K_f) = I_f') # 20
model.add('h = h(-1)*gamma_u*(u-un) + h(-1)') # Eq 21 # Version without_l
\hookrightarrow corridor
model.add('V_f = K_f - Lf') # Eq 22
model.add("NFW_f = FU - I_f") # Eq 23
```

```
# Banks
 model.add('L = Lf')
 model.add('d(M) = d(L) + d(MO)') # Eq 24
 model.add('rmo = rm + spread_mo') # Eq 25
 model.add('rl = rm + spread_l') # Eq 26
 model.add('V_b = L + MO - M') # Eq 27
 model.add('NFW_b = rl*L(-1) + rmo*MO(-1) - rm*M(-1)') # Eq 28
 # Residential investment
 model.add('K_HS = K_HD') # Eq 29
 model.add('Is = I h')
 model.add('d(K_HD) = I_h') # Eq 30
 model.add('I_h = (1+g_Z)*I_h(-1)') # Eq 31
 model.add(
     'own = ((1+rmo)/(1+infla)) -1'
  model.add('Residual = d(M) - Sh_k - Sh_w')
 return model
t_{check} = 100
print('Evaluating consistenty at time = {}'.format(t_check))
test = model()
SolveSFC(test, time=t_check, table = False)
evaldf = pd.DataFrame({
   'Households' : test.evaluate('M - MO + K_HD*ph - V_h'),
'Firms' : test.evaluate('K f - Lf - V f'),
   'Banks' : test.evaluate('L + MO - M - V_b'),
   'Mortgages' : test.evaluate('MO(-1) - MO(-2) - I_h(-1)'),
   'Financial assets' : test.evaluate('d(MO) + d(L) - d(M)'),
   '[Total Wealth - K]' : test.evaluate('V_f + V_h + V_b - K'),
   "Firm's Funds": test.evaluate('Lf(-1) - Lf(-2) - I f(-1) + FU(-1)'),
   "Housing" : test.evaluate('K_HD - K_HS'),
   "Workers saving": test.evaluate('YD_w - C_w'),
   "Capitalist saving": test.evaluate('YD_k - C_k - Sh_k'),
   "Flow check" : test.evaluate('NFW_h + NFW_f + NFW_b'),
   "Investment" : test.evaluate('I_t - I_h - I_f'),
   "Profits" : test.evaluate('FT(-1) - rl*Lf(-2) - FU(-1) - FD(-1)'),
   "Wages" : test.evaluate('W - omega*Y'),
```

```
"Residual" : test.evaluate('Residual')
}, index = ['Sum'])
evaldf = evaldf.transpose()
evaldf.round(5)
```

Evaluating consistenty at time = 100

	$\operatorname{\mathtt{Sum}}$
Households	-0.00009
Firms	0.00000
Banks	0.00000
Mortgages	0.00000
Financial assets	0.00000
[Total Wealth - K]	0.00013
Firm's Funds	0.00008
Housing	0.00000
Workers saving	0.00000
Capitalist saving	-0.00000
Flow check	-0.00022
Investment	-0.00004
Profits	0.00000
Wages	0.00000
Residual	-0.19976
	Firms Banks Mortgages Financial assets [Total Wealth - K] Firm's Funds Housing Workers saving Capitalist saving Flow check Investment Profits Wages

# 7 Solving

```
[6]: base = model()
    df = SolveSFC(base, time=1000)
    df.transpose()
[6]:
                                        2
                                                     3
   C
                0.0
                      87.967573
                                   91.449946
                                                95.048891
                                                            98.770855
                                                                        102.621943
    C_w
                0.0
                      82.213358
                                   85.416921
                                                88.726708
                                                            92.148368
                                                                         95.687285
    C_k
                0.0
                       5.754891
                                   6.033472
                                                6.322641
                                                             6.622953
                                                                          6.934951
   FD
                0.0
                      19.183117
                                   20.111672
                                                21.075570
                                                            22.076614
                                                                         23.116568
   Fn
                0.0
                     191.831169
                                  201.116719
                                              210.755697
                                                           220.766137
                                                                        231.165679
    _h__1
                0.3
                                                0.299229
                       0.300000
                                    0.299655
                                                             0.298731
                                                                          0.298168
   _L__1
                0.0
                       0.000000
                                  -90.528459 -186.335588 -287.663933 -394.767401
    _K_HD__1
              500.0
                     500.000000
                                  603.960000
                                              712.036816
                                                           824.393474
                                                                        941.199455
   _I_h__1
              100.0
                     100.000000
                                  103.960000
                                               108.076816
                                                           112.356658
                                                                        116.805982
                        1.000000
                                                             1.000000
                                                                          1.000000
    _ph__1
                1.0
                                    1.000000
                                                 1.000000
                     6
                                   7
                                                 8
                                                              9
    С
               106.606201
                             110.729969
                                          114.999022
                                                        119.419298
    C_w
                99.347450
                                          107.053542
                             103.134458
                                                        111.110087
```

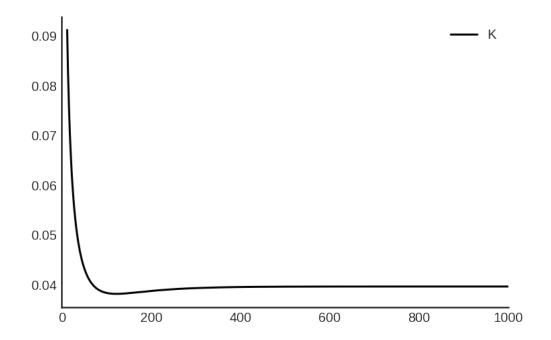
```
C_k
            7.259049
                        7.595814
                                    7.945787
                                                 8.309523
FD
                       25.319448
                                                 27.698478
           24.196894
                                    26.486024
          241.968943
                       253.194477
                                    264.860241
                                                 276.984779
. . .
                 . . .
                              . . .
                                           . . .
                                                        . . .
           0.297547
                        0.296875
_h__1
                                      0.296156
                                                  0.295396
_L__1
                     -627.370464
                                  -753.432140
         -507.911275
                                               -886.395504
_K_HD__1 1062.630954 1188.871140 1320.110437
                                                1456.546810
_I_h__1
         121.431498
                       126.240186
                                    131.239297
                                                 136.436373
_ph__1
            1.000000
                        1.000000
                                     1.000000
                                                   1.000000
                 991
                               992
                                             993
                                                           994
С
          3.208141e+18 3.335184e+18 3.467257e+18 3.604560e+18
C_w
          2.872427e+18 2.986175e+18 3.104427e+18 3.227363e+18
C_k
          3.357145e+17 3.490088e+17
                                     3.628295e+17 3.771975e+17
FD
          1.119048e+18 1.163363e+18
                                    1.209432e+18 1.257325e+18
Fn
          1.119048e+19 1.163363e+19 1.209432e+19 1.257325e+19
_h__1
         1.237514e-01 1.237514e-01
                                    1.237514e-01
                                                  1.237514e-01
_L__1
        -2.244076e+20 -2.332942e+20 -2.425326e+20 -2.521369e+20
_K_HD__1 1.308517e+20 1.360334e+20 1.414203e+20 1.470205e+20
_I_h__1
         4.984345e+18 5.181725e+18 5.386922e+18 5.600244e+18
          1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
_ph__1
                 995
                               996
                                             997
                                                           998
С
          3.747301e+18 3.895694e+18 4.049963e+18 4.210341e+18
C_w
          3.355166e+18 3.488031e+18 3.626156e+18 3.769752e+18
                                     4.238065e+17 4.405893e+17
C_k
          3.921346e+17 4.076631e+17
FD
         1.307115e+18 1.358877e+18 1.412688e+18 1.468631e+18
Fn
          1.307115e+19 1.358877e+19
                                     1.412688e+19 1.468631e+19
. . .
         1.237514e-01
_h__1
                       1.237514e-01
                                    1.237513e-01
                                                  1.237513e-01
_{\rm L}_{\rm -1}
         -2.621215e+20 -2.725016e+20 -2.832926e+20 -2.945110e+20
_K_HD__1 1.528426e+20 1.588951e+20
                                    1.651874e+20
                                                  1.717288e+20
_I_h__1
         5.822013e+18 6.052565e+18 6.292247e+18 6.541420e+18
_ph__1
         1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
                 999
                               1000
С
          4.377071e+18 4.550403e+18
C w
          3.919034e+18 4.074228e+18
C_k
          4.580366e+17 4.761748e+17
FD
          1.526789e+18 1.587249e+18
          1.526789e+19 1.587249e+19
. . .
                  . . .
        1.237513e-01 1.237513e-01
_h__1
       -3.061737e+20 -3.182981e+20
_{
m L}_{
m -1}
_K_HD__1 1.785292e+20 1.855990e+20
_{
m I}_{
m h}_{
m l}
         6.800460e+18 7.069758e+18
```

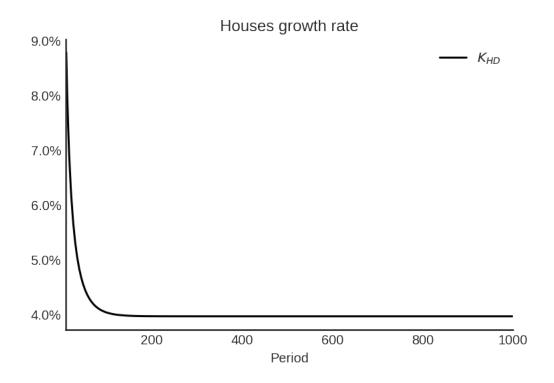
```
_ph__1 1.000000e+00 1.000000e+00 [67 rows x 1001 columns]
```

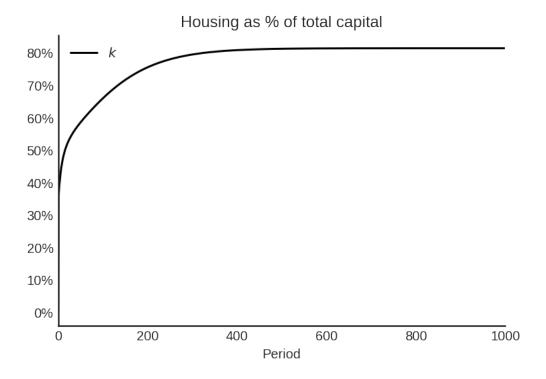
#### 8 Baseline Plots

```
[7]: fig, ax = plt.subplots()

df[["K"]].pct_change().rolling(12).mean().plot(ax=ax, color='black')
    sns.despine()
    plt.show()
```

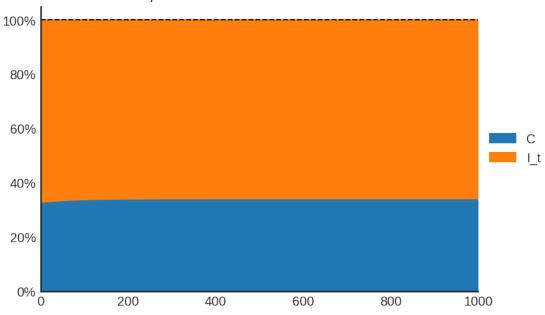




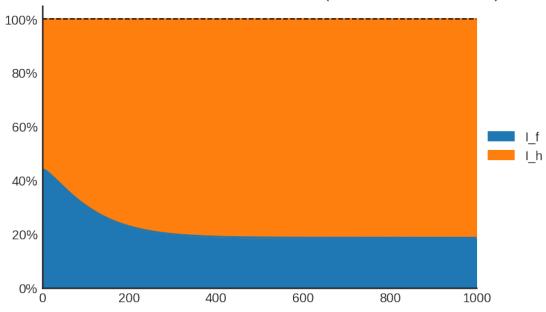


```
[10]: ax = df[['C', 'I_t']].apply(lambda x: x/df['Y']).plot(kind = 'area', stacked = True, legend = True, title = "Consumption and total investiment as % of GDP") ax.axhline(y=1, color = "black", ls = "--", lw=1) ax.legend(loc='center left', bbox_to_anchor=(1, 0.5)) ax.set_yticklabels(['{:,.0%}'.format(x) for x in ax.get_yticks()]) sns.despine() plt.show()
```

#### Consumption and total investiment as % of GDP







```
[12]: ax = df[['W', 'FT']].apply(lambda x: x/df['Y']).plot(kind = 'area', stacked = True, legend = True, title = "Wage and profit share (as % of GDP)")

ax.axhline(y=1, color = "black", ls = "--", lw=1)

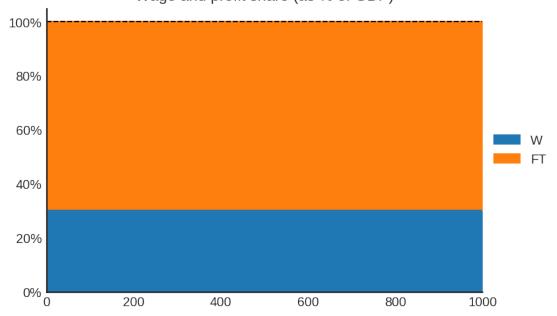
ax.legend(loc='center left', bbox_to_anchor=(1, 0.5))

ax.set_yticklabels(['{:,.0%}'.format(x) for x in ax.get_yticks()])

sns.despine()

plt.show()
```





```
[13]: ax = df[['V_b', 'V_h', 'V_f']].apply(lambda x: np.abs(x)/df['K']).plot(kind = \( \to '\) area', stacked = True, title = "Net wealth (as % K)")

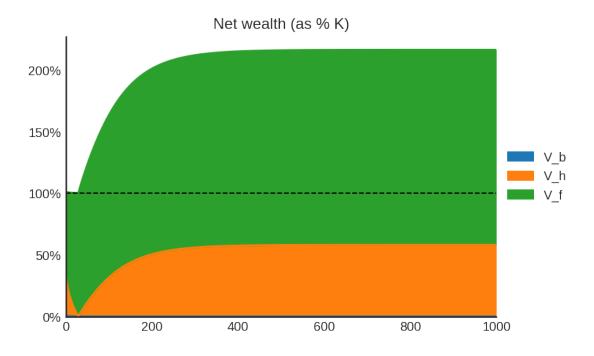
ax.set_yticklabels(['{:,.0%}'.format(x) for x in ax.get_yticks()])

ax.axhline(y=1, color = "black", ls = "--", lw=1)

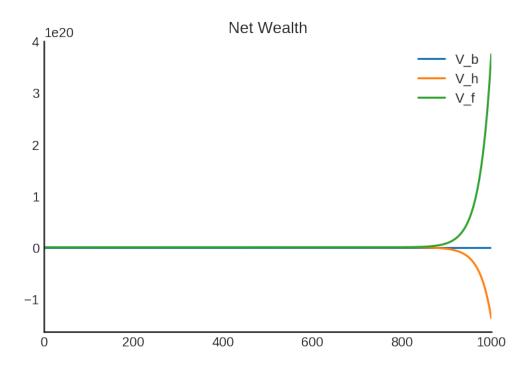
ax.legend(loc='center left', bbox_to_anchor=(1, 0.5))

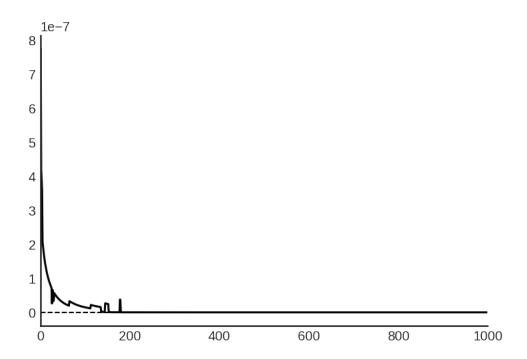
sns.despine()

plt.show()
```

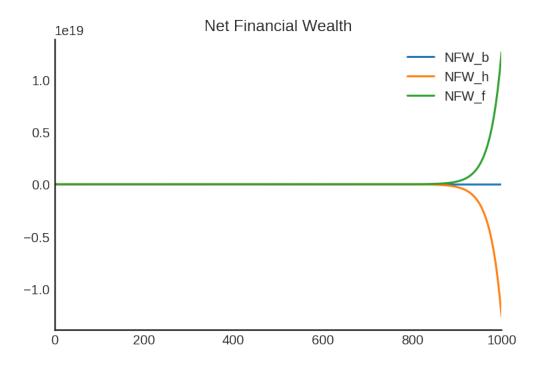


```
[14]: df[["V_b", "V_h", "V_f"]].plot(title = "Net Wealth")
sns.despine()
plt.show()
```

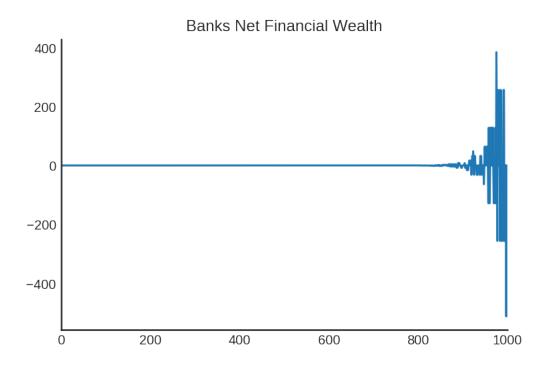




```
[16]: df[["NFW_b", "NFW_h", "NFW_f"]].plot(title = "Net Financial Wealth")
sns.despine()
plt.show()
```



```
[17]: df['NFW_b'].plot(title = "Banks Net Financial Wealth")
sns.despine()
plt.show()
```

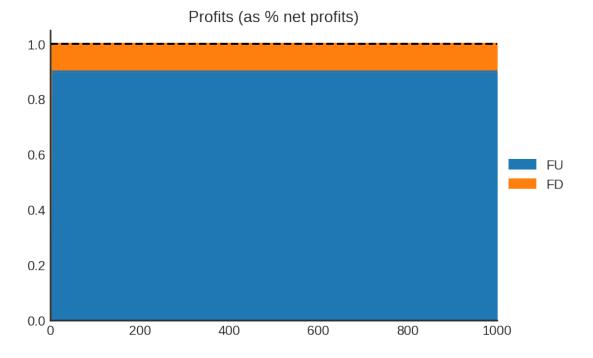


```
[18]: ax = df[["FU", "FD"]].apply(lambda x: x/df["Fn"]).plot(kind = "area", stacked = True, title = "Profits (as % net profits)")

ax.legend(loc='center left', bbox_to_anchor=(1, 0.5))

ax.axhline(y = 1, color = "black", ls = "--")

sns.despine()
plt.show()
```



```
[19]: ax = df[["MO", "L"]].apply(lambda x: x/df["M"]).plot(kind = "area",stacked = False ,title = "Credit (as % Deposits)")

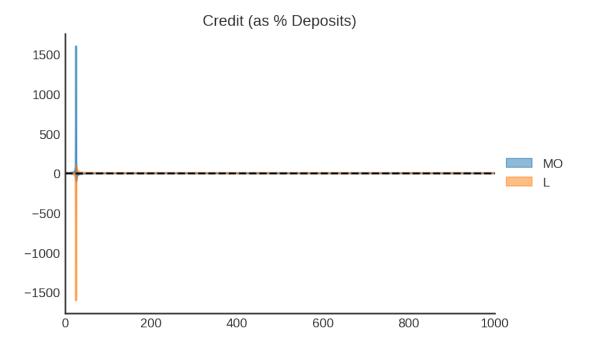
ax.legend(loc='center left', bbox_to_anchor=(1, 0.5))

ax.axhline(y = 1, color = "blue", ls = "--")

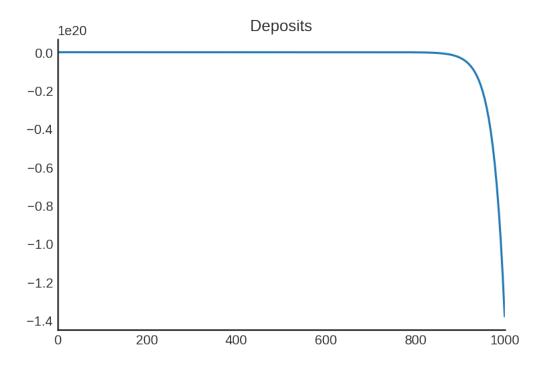
ax.axhline(y = 0, color = "black", ls = "--")

sns.despine()

plt.show()
```



```
[20]: df["M"].plot(title = "Deposits")
sns.despine()
plt.show()
```



```
[21]: ax = df['Y'].pct_change().plot(color = "black", title = "GDP growth rate", □

→label = "$Y_t$", legend = True)

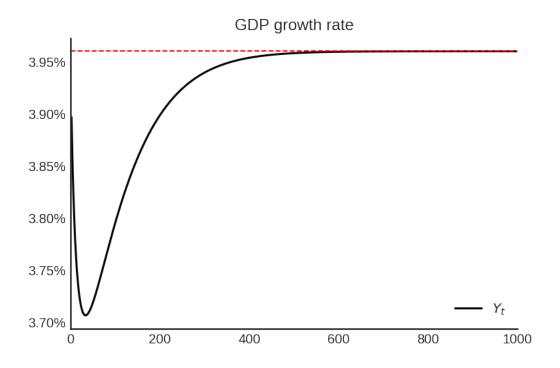
ax.set_yticklabels(['{:,.2%}'.format(x) for x in ax.get_yticks()])

ax.axhline(y=df["g_Z"].iloc[-1], color = "red", ls = "--", lw=1, label = □

→"Autonomous growth rate")

sns.despine()

plt.show()
```



```
[22]: ax = df['u'].plot(color = "black", title = "Capacity utilization ratio", label

⇒= "$u_t$", legend = True)

ax.set_yticklabels(['{:,.1%}'.format(x) for x in ax.get_yticks()])

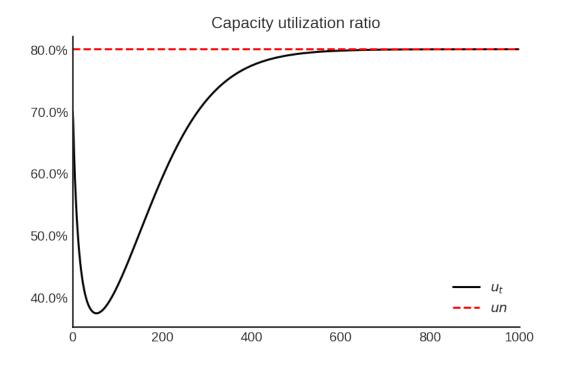
ax = df['un'].plot(color = "red", ls="--", title = "Capacity utilization

⇒ratio", label = "$un$", legend = True)

ax.set_yticklabels(['{:,.1%}'.format(x) for x in ax.get_yticks()])

sns.despine()

plt.show()
```



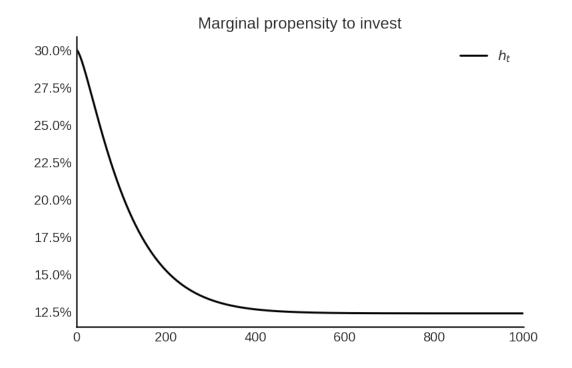
```
[23]: ax = df['h'].plot(color = "black", ls="-", title = "Marginal propensity to⊔

invest", label = "$h_t$", legend = True)

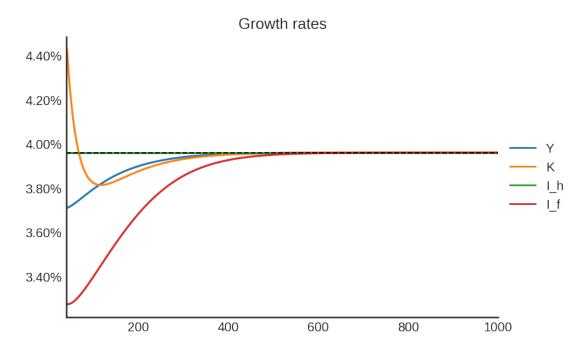
ax.set_yticklabels(['{:,.1%}'.format(x) for x in ax.get_yticks()])

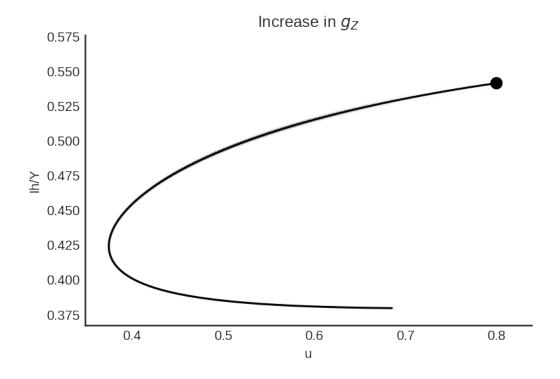
sns.despine()

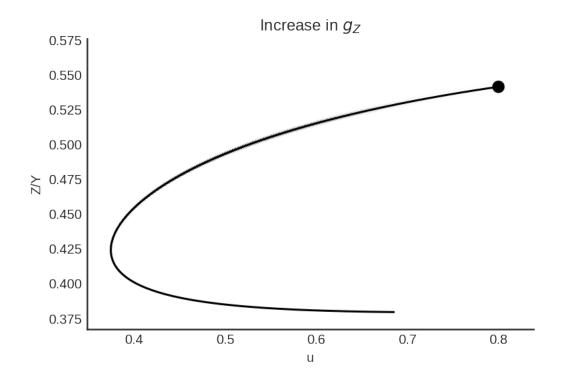
plt.show()
```

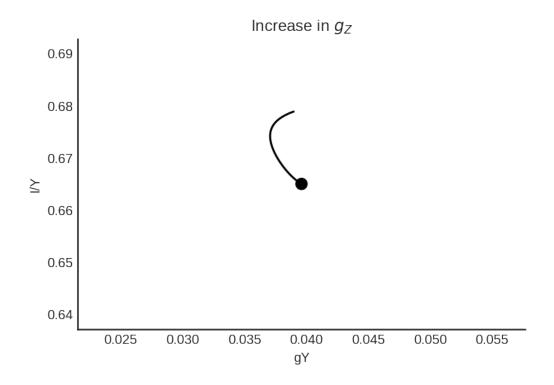


```
[24]: ax = df[["Y", "K", "I_h", "I_f"]][40:].pct_change().plot(title = "Growth rates")
    ax.set_yticklabels(['{:,.2%}'.format(x) for x in ax.get_yticks()])
    ax.axhline(y=df["g_Z"].iloc[-1], color = "black", ls = "--", lw=1)
    ax.legend(loc='center left', bbox_to_anchor=(1, 0.5))
    sns.despine()
    plt.show()
```









#### 9 Shocks

```
[26]: def plot_shock(filename, shock):
         This function plots some selected variables
         filename: name to save the plot (str)
         shock: df returned by ShockModel function
         sns.set_context('paper')
         fig, ax = plt.subplots(2,2)
         shock[["Y"]].pct change().plot(title = "Growth rates", ax = ax[0,0],
                                                             ls = ('--'),
         shock[["K"]].pct_change().plot(title = "Growth rates", ax = ax[0,0],
                                                             ls = (':'),
         shock[["I_h"]].pct_change().plot(title = "Growth rates", ax = ax[0,0],
                                                             ls = ('-'),
         shock[["I_f"]].pct_change().plot(title = "Growth rates", ax = ax[0,0],
                                                             ls = ('-.').
         ax[0,0].axhline(y=shock["g Z"].iloc[-1], color = "black", ls = "--", lw=1)
         \#ax[0,0].set\_yticklabels(['{:,.1%}'.format(x) for x in ax[0,0].
      \rightarrow get_yticks()])
         ax[0,0].legend(loc='upper center', bbox_to_anchor=(0.5, -0.06),
                         labels = ["$Y$", "$K$", "$I_h$", "$I_f$"],
                   fancybox=True, shadow=True, ncol=2)
         ax[0,0].ticklabel_format(useOffset=False)
         shock['K_k'].plot(color = "darkred", title = "Houses \nas % of Totalu
      \rightarrowcapital", label = "$\k$", legend = False, ax = ax[0,1], lw = 3, )
         ax[0,1].axhline(y = df['K_k'].iloc[-1], ls ='--', color = "gray")
         ax[0,1].ticklabel_format(useOffset=False)
         shock['u'].plot(title = 'Capacity utilization ratio', ax=ax[1,0], legend = ___
      →False, color = "darkred", lw = 3, )
         ax[1,0].axhline(y = shock['un'].iloc[-1], ls ='--', color = "gray")
         \#ax[1,0].set\_yticklabels(['\{:,.2\%\}'.format(x)) for x in <math>ax[1,0].
      \rightarrow get_yticks()])
         ax[1,0].ticklabel_format(useOffset=False)
         shock['h'].plot(title = 'Marginal propensity to invest', ax=ax[1,1], legend∪

⇒= False, color = "darkred", lw = 3, )
```

```
ax[1,1].axhline(y = df['h'].iloc[-1], ls ='--', color = "gray")
ax[1,1].ticklabel_format(useOffset=False)

sns.despine()
plt.tight_layout(rect=[0, 0.03, 1, 0.95])
plt.show()

fig.savefig(filename, dpi = 300)
```

#### 9.1 Increase in autonomous growth rate $(g_z)$

```
[27]: base = model()
    df = SolveSFC(base, time=900)
    shock = ShockModel(base_model=base, create_function=model(), variable='phi_0',__
     \rightarrowincrease=0.01, time = 1200)
    shock.transpose()
[27]:
                                    1
                                                  2
    C
              9.363219e+16 9.734002e+16 1.011947e+17 1.052020e+17
    C_w
              8.383412e+16 8.715394e+16 9.060523e+16 9.419319e+16
    C k
              9.798072e+15 1.018607e+16 1.058944e+16 1.100878e+16
    FD
              3.266024e+16 3.395358e+16
                                          3.529814e+16 3.669595e+16
    Fn
              3.266024e+17
                            3.395358e+17
                                          3.529814e+17
                                                        3.669595e+17
     . . .
                        . . .
    _h__1
              1.237546e-01 1.237546e-01 1.237545e-01 1.237545e-01
    _L__1
             -6.808832e+18 -6.808832e+18 -7.078462e+18 -7.358769e+18
    _K_HD__1 3.970222e+18 3.970222e+18 4.127443e+18 4.290890e+18
    _I_h__1
              1.512320e+17 1.512320e+17
                                          1.572208e+17 1.634467e+17
    _ph__1
              1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
                      4
                                    5
                                                                      \
    С
              1.093680e+17 1.136989e+17
                                          1.182014e+17 1.228821e+17
    C w
              9.792323e+16 1.018010e+17
                                          1.058323e+17
                                                        1.100232e+17
    C_k
              1.144473e+16 1.189794e+16
                                          1.236910e+16 1.285892e+16
    FD
              3.814911e+16 3.965981e+16
                                          4.123033e+16 4.286305e+16
                                          4.123033e+17 4.286305e+17
    Fn
              3.814911e+17
                            3.965981e+17
              1.237544e-01 1.237544e-01 1.237543e-01
                                                       1.237543e-01
     _h__1
    _L__1
             -7.650176e+18 -7.953123e+18 -8.268067e+18 -8.595483e+18
    _K_HD__1 4.460809e+18 4.637457e+18 4.821101e+18 5.012016e+18
    _I_h__1
              1.699192e+17 1.766480e+17
                                          1.836433e+17 1.909156e+17
              1.000000e+00 1.000000e+00
                                          1.000000e+00 1.000000e+00
    _ph__1
                      8
                                    9
                                                       1241
                                                                     1242 \
    С
              1.277483e+17
                            1.328071e+17
                                               7.397364e+42 7.764273e+42
    C w
              1.143801e+17
                            1.189096e+17
                                           ... 6.724489e+42 7.058024e+42
    C_k
              1.336813e+16 1.389751e+16
                                               6.728746e+41 7.062491e+41
```

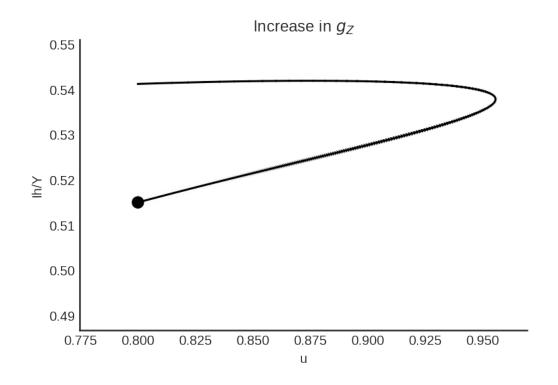
```
FD
               4.456043e+16 4.632502e+16
                                          ... 2.242915e+42 2.354164e+42
                            4.632502e+17
                                                2.242915e+43 2.354164e+43
    Fn
               4.456043e+17
     . . .
                        . . .
     _h__1
               1.237542e-01
                            1.237541e-01
                                                1.549999e-01
                                                              1.549999e-01
    _{
m L}_{
m L} 1
              -8.935864e+18 -9.289725e+18
                                           ... -3.369339e+44 -3.536458e+44
    _K_HD__1 5.210492e+18 5.416827e+18
                                           ... 2.327275e+44 2.442708e+44
    _I_h__1
                            2.063355e+17
                                           ... 1.099779e+43 1.154328e+43
               1.984758e+17
    _ph__1
               1.000000e+00
                             1.000000e+00
                                                1.000000e+00 1.000000e+00
                       1243
                                     1244
                                                   1245
                                                                 1246 \
    C
               8.149381e+42 8.553590e+42
                                           8.977848e+42 9.423150e+42
               7.408102e+42 7.775544e+42 8.161211e+42 8.566007e+42
    C_w
    C k
               7.412791e+41
                            7.780465e+41
                                          8.166376e+41 8.571429e+41
    FD
               2.470930e+42 2.593488e+42
                                           2.722125e+42 2.857143e+42
    Fn
               2.470930e+43 2.593488e+43
                                           2.722125e+43
                                                        2.857143e+43
                                           1.549999e-01
     _h__1
               1.549999e-01
                            1.549999e-01
                                                         1.549999e-01
     _L__1
              -3.711866e+44 -3.895975e+44 -4.089215e+44 -4.292040e+44
    _K_HD__1 2.563866e+44 2.691034e+44
                                           2.824509e+44 2.964604e+44
    _I_h__1
               1.211583e+43 1.271677e+43
                                           1.334753e+43
                                                        1.400956e+43
    _ph__1
               1.000000e+00 1.000000e+00
                                           1.000000e+00 1.000000e+00
                       1247
                                                   1249
                                                                 1250
                                     1248
    C
               9.890538e+42 1.038111e+43
                                          1.089601e+43 1.143645e+43
    C_w
               8.990881e+42
                            9.436828e+42
                                           9.904895e+42
                                                        1.039618e+43
    C k
               8.996572e+41 9.442802e+41
                                           9.911164e+41
                                                        1.040276e+42
                                           3.303721e+42
                                                        3.467586e+42
    FD
               2.998857e+42 3.147601e+42
               2.998857e+43 3.147601e+43 3.303721e+43 3.467586e+43
    Fn
     _h__1
               1.549999e-01 1.549999e-01
                                          1.549999e-01
                                                        1.549999e-01
    _{
m L}_{
m L} 1
              -4.504925e+44 -4.728369e+44 -4.962897e+44 -5.209056e+44
    _K_HD__1 3.111649e+44 3.265987e+44
                                                        3.598007e+44
                                           3.427980e+44
    _{
m I} _{
m h} _{
m I}
               1.470444e+43 1.543378e+43
                                           1.619929e+43
                                                        1.700278e+43
    _ph__1
               1.000000e+00 1.000000e+00
                                           1.000000e+00
                                                         1.000000e+00
     [67 rows x 1251 columns]
[28]: summary = SummaryShock(shock)
     summary
[28]:
                          0
                                                      1
                                    Shock
                                                                     2
     С
               6.527354e+17
                            6.846897e+17
                                          7.182326e+17
                                                         7.534437e+17
    C_w
               5.844302e+17 6.132802e+17
                                           6.435693e+17
                                                         6.753698e+17
    C k
               6.830513e+16 7.140947e+16
                                           7.466328e+16
                                                        7.807393e+16
    FD
               2.276838e+17 2.380316e+17
                                           2.488776e+17
                                                         2.602464e+17
               2.276838e+18 2.380316e+18
                                           2.488776e+18
                                                         2.602464e+18
    Fn
     _h__1
               1.237525e-01 1.237524e-01
                                          1.237617e-01 1.237800e-01
```

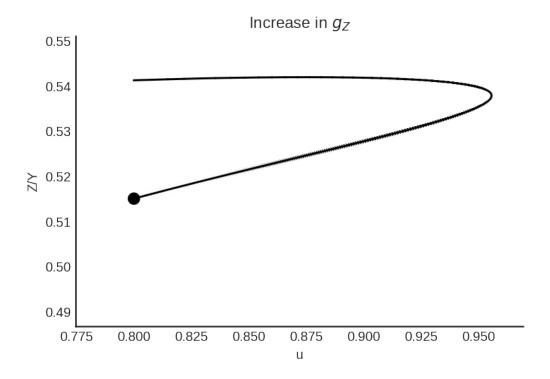
```
-4.565836e+19 -4.746643e+19 -4.935571e+19 -5.133008e+19
    _K_HD__1 2.662331e+19 2.767760e+19
                                           2.878417e+19 2.994563e+19
    _I_h__1
               1.014124e+18 1.054283e+18
                                           1.106576e+18 1.161462e+18
    _ph__1
               1.000000e+00
                             1.000000e+00
                                           1.000000e+00
                                                        1.000000e+00
                                                           difference
                          3
                                      t.-1
                                                      t
     С
                                           1.143645e+43 5.404422e+41
               7.904064e+17
                             1.089601e+43
    C_w
               7.087572e+17
                             9.904895e+42
                                           1.039618e+43 4.912828e+41
     C_k
               8.164915e+16 9.911164e+41
                                           1.040276e+42 4.915938e+40
    FD
               2.721638e+17
                             3.303721e+42
                                           3.467586e+42 1.638646e+41
    Fn
               2.721638e+18 3.303721e+43
                                           3.467586e+43
                                                        1.638646e+42
     . . .
     _h__1
               1.238071e-01 1.549999e-01 1.549999e-01 8.355082e-10
    _L__1
              -5.339358e+19 -4.962897e+44 -5.209056e+44 -2.461597e+43
     _K_HD__1 3.116470e+19 3.427980e+44
                                           3.598007e+44 1.700278e+43
    _I_h__1
               1.219070e+18 1.619929e+43
                                           1.700278e+43 8.034850e+41
               1.000000e+00 1.000000e+00
                                           1.000000e+00 0.000000e+00
    _ph__1
     [67 rows x 8 columns]
[29]: | shock1 = shock.round(decimals = 5).tail(1).transpose().loc['alpha':,:]
     shock1.columns = ['$\Delta \phi_0$']
     shock1
[29]:
                $\Delta \phi_0$
     alpha
                   1.000000e+00
     alpha 2
                   3.000000e-01
     gamma_F
                   9.000000e-01
    gamma_u
                   1.000000e-02
     omega
                   3.000000e-01
                   2.000000e-02
    rm
     spread_1
                   0.000000e+00
     spread_mo
                   0.000000e+00
    un
                   8.00000e-01
     v
                   2.500000e+00
    phi_0
                   5.000000e-02
    phi_1
                   2.000000e-02
                   1.000000e+00
    phparam
     infla
                   0.000000e+00
     gZn
                   3.922000e-02
     _K_f__1
                   1.082934e+44
     _M__1
                  -1.611049e+44
    _MO__1
                   3.598007e+44
     _Lf__1
                  -5.209056e+44
    _h__1
                   1.550000e-01
    _L__1
                  -5.209056e+44
    _K_HD__1
                   3.598007e+44
    _I_h__1
                   1.700278e+43
```

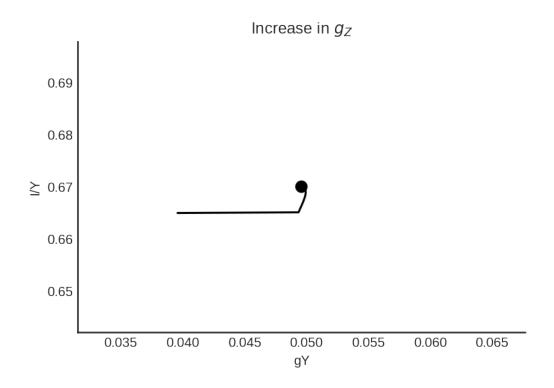
#### \_ph\_\_1 1.000000e+00

```
[30]: shock["TIME"] = [i+1 for i in range(len(shock.index))]
    shock["Ih/Y"] = shock["I_h"]/shock["Y"]
    shock["I/Y"] = shock["I_t"]/shock["Y"]
    shock["Z/Y"] = shock["Z"]/shock["Y"]
    shock["gY"] = shock["Y"].pct_change()
    sns.scatterplot(y = 'Ih/Y', x='u', data=shock, size="TIME", sizes = (1,100), u

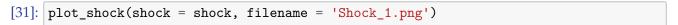
→color = 'black', legend=False)
    sns.lineplot(y = 'Ih/Y', x='u', data=shock, sort=False, color = 'black')
    plt.title('Increase in $g_Z$')
    sns.despine()
    plt.show()
    sns.scatterplot(y = 'Z/Y', x='u', data=shock, size="TIME", sizes = (1,100), u
     sns.lineplot(y = 'Z/Y', x='u', data=shock, sort=False, color = 'black')
    plt.title('Increase in $g Z$')
    sns.despine()
    plt.show()
    sns.scatterplot(y = 'I/Y', x='gY', data=shock, size="TIME", sizes = (1,100), __
    sns.lineplot(y = 'I/Y', x='gY', data=shock, sort=False, color = 'black')
    plt.title('Increase in $g_Z$')
    sns.despine()
    plt.show()
```

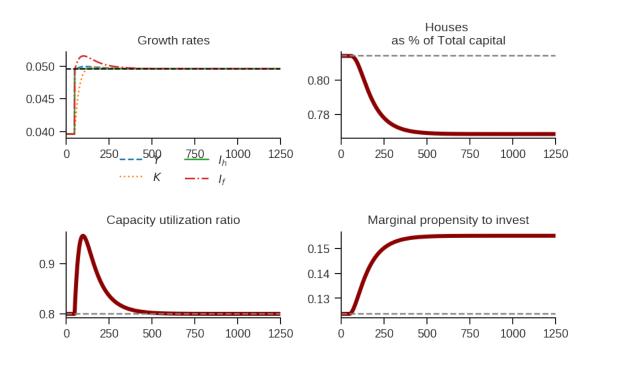






#### 9.1.1 Plots



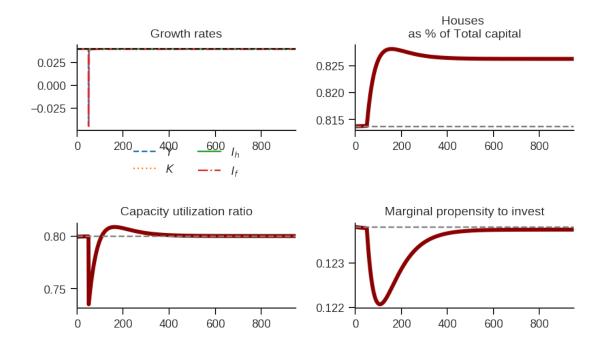


#### 9.2 Decrease in wage-share ( $\omega$ )

```
[32]: base = model()
     df = SolveSFC(base, time=700)
     shock = ShockModel(base_model=base, create_function=model(), variable='omega',_
      \rightarrowincrease=-0.05, time = 900)
     shock.transpose()
                                                    2
[32]:
                        0
                                      1
                                                                  3
     С
               3.964712e+13 4.121709e+13
                                           4.284923e+13 4.454601e+13
                                                         3.988472e+13
    C_w
               3.549846e+13 3.690415e+13
                                           3.836550e+13
    C_k
               4.148657e+12 4.312940e+12
                                           4.483730e+12 4.661283e+12
    FD
               1.382886e+13 1.437647e+13
                                           1.494577e+13 1.553761e+13
    Fn
               1.382886e+14
                             1.437647e+14
                                           1.494577e+14
                                                         1.553761e+14
                             1.238098e-01
                                           1.238090e-01
                                                         1.238083e-01
     _h__1
               1.238098e-01
     _L__1
              -2.882750e+15 -2.882750e+15 -2.996908e+15 -3.115587e+15
     _K_HD__1 1.680973e+15 1.680973e+15
                                           1.747539e+15
                                                         1.816742e+15
     _I_h__1
               6.403090e+13 6.403090e+13
                                           6.656652e+13
                                                        6.920255e+13
    _ph__1
               1.000000e+00 1.000000e+00
                                           1.000000e+00 1.000000e+00
                                      5
                                                    6
                                                                  7
     С
               4.630997e+13 4.814379e+13
                                           5.005022e+13 5.203215e+13
     C_{w}
               4.146410e+13 4.310603e+13
                                           4.481297e+13 4.658750e+13
    C_k
               4.845866e+12 5.037759e+12
                                           5.237251e+12 5.444643e+12
                             1.679253e+13
                                           1.745750e+13
    FD
               1.615289e+13
                                                         1.814881e+13
    Fn
               1.615289e+14
                             1.679253e+14
                                           1.745750e+14
                                                         1.814881e+14
     . . .
     _h__1
               1.238075e-01 1.238068e-01
                                          1.238061e-01
                                                        1.238054e-01
     _L__1
              -3.238965e+15 -3.367229e+15 -3.500573e+15 -3.639197e+15
    _K_HD__1
              1.888685e+15 1.963477e+15
                                           2.041230e+15
                                                        2.122063e+15
     _I_h__1
               7.194297e+13 7.479192e+13
                                           7.775368e+13
                                                         8.083272e+13
               1.000000e+00
                            1.000000e+00
                                           1.000000e+00
                                                        1.000000e+00
     _ph__1
                        8
                                      9
                                                         941
                                                                       942
    С
               5.409255e+13 5.623455e+13
                                                2.327875e+29 2.420059e+29
     C_w
               4.843231e+13 5.035016e+13
                                                2.022041e+29 2.102114e+29
     C_k
               5.660248e+12
                            5.884390e+12
                                                3.058338e+28 3.179448e+28
    FD
               1.886749e+13
                             1.961463e+13
                                           ... 1.019446e+29 1.059816e+29
     Fn
               1.886749e+14
                             1.961463e+14
                                           . . .
                                                1.019446e+30
                                                              1.059816e+30
     . . .
                                           . . .
               1.238046e-01 1.238040e-01
                                           ... 1.237499e-01
                                                             1.237499e-01
     _h__1
                                           ... -2.064168e+31 -2.145909e+31
    _L__1
              -3.783310e+15 -3.933131e+15
     _K_HD__1 2.206097e+15 2.293458e+15
                                           ... 1.201864e+31 1.249458e+31
```

```
_{\rm I}_{\rm h}_{\rm 1} 8.403370e+13 8.736143e+13 ... 4.578088e+29 4.759381e+29
_ph__1
         1.000000e+00
                      1.000000e+00
                                    ... 1.000000e+00 1.000000e+00
                  943
                                944
                                              945
                                                           946 \
С
         2.515893e+29 2.615523e+29 2.719097e+29 2.826774e+29
C_w
         2.185358e+29 2.271898e+29
                                    2.361865e+29 2.455395e+29
C_k
         3.305354e+28 3.436246e+28 3.572321e+28 3.713785e+28
FD
         1.101785e+29 1.145415e+29 1.190774e+29 1.237928e+29
          1.101785e+30 1.145415e+30 1.190774e+30 1.237928e+30
Fn
. . .
_h__1
         1.237499e-01 1.237499e-01 1.237499e-01 1.237499e-01
        -2.230887e+31 -2.319230e+31 -2.411071e+31 -2.506550e+31
_L__1
_K_HD__1 1.298936e+31 1.350374e+31 1.403849e+31 1.459441e+31
_I_h__1
         4.947852e+29 5.143787e+29 5.347481e+29 5.559241e+29
         1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
_ph__1
                                             949
                                                           950
                  947
                                948
С
         2.938714e+29 3.055087e+29
                                    3.176068e+29 3.301841e+29
C_w
         2.552629e+29 2.653713e+29
                                     2.758800e+29 2.868048e+29
         3.860851e+28 4.013741e+28 4.172685e+28 4.337924e+28
C_k
FD
         1.286950e+29 1.337914e+29 1.390895e+29 1.445975e+29
         1.286950e+30 1.337914e+30 1.390895e+30 1.445975e+30
Fn
. . .
_h__1
        1.237499e-01 1.237499e-01 1.237499e-01
                                                 1.237499e-01
_L__1
        -2.605809e+31 -2.708999e+31 -2.816275e+31 -2.927800e+31
_K_HD__1 1.517235e+31 1.577318e+31 1.639779e+31 1.704715e+31
_I_h__1
         5.779387e+29 6.008251e+29 6.246178e+29 6.493526e+29
         1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
_ph__1
[67 rows x 951 columns]
```

[33]: plot\_shock(shock = shock, filename = 'Shock\_2.png')



```
[34]: shock2 = shock.round(decimals = 3).tail(1).transpose().loc['alpha':,:] shock2.columns = ['$\Delta \omega$'] shock2
```

```
[34]:
                $\Delta \omega$
                    1.000000e+00
     alpha
     alpha_2
                    3.000000e-01
     gamma_F
                    9.00000e-01
     gamma_u
                    1.000000e-02
                    2.500000e-01
     omega
                    2.000000e-02
     rm
     spread_1
                    0.000000e+00
     spread_mo
                    0.000000e+00
     un
                    8.000000e-01
                    2.500000e+00
     phi_0
                    4.000000e-02
     phi_1
                    2.000000e-02
     phparam
                    1.000000e+00
     infla
                    0.000000e+00
     gZn
                    3.900000e-02
     _K_f__1
                    3.585057e+30
     _M__1
                   -1.223085e+31
     _MO__1
                    1.704715e+31
     _Lf__1
                  -2.927800e+31
     _h__1
                    1.240000e-01
     _L__1
                   -2.927800e+31
     _K_HD__1
                    1.704715e+31
```

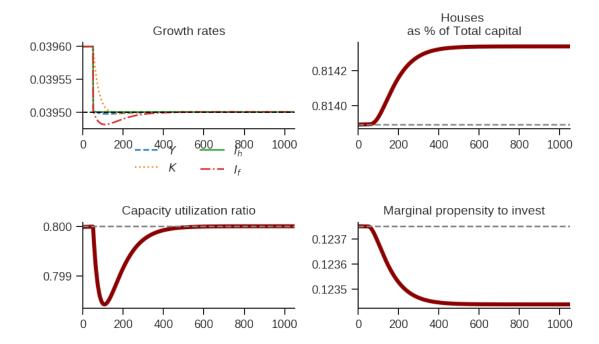
```
_I_h__1 6.493526e+29
_ph__1 1.000000e+00
```

#### 9.3 Increase in interest rates

```
[35]: base = model()
    df = SolveSFC(base, time=1000)
    shock = ShockModel(base_model=base, create_function=model(),__
     →variable='spread_mo', increase=0.005, time = 1000)
    shock.transpose()
[35]:
                      0
                                                   2
                                                                 3
    С
              4.550403e+18 4.730599e+18 4.917930e+18 5.112680e+18
    C_w
              4.074228e+18 4.235567e+18 4.403296e+18 4.577666e+18
    C k
              4.761748e+17
                            4.950313e+17
                                          5.146346e+17
                                                        5.350141e+17
    FD
               1.587249e+18 1.650104e+18
                                          1.715449e+18
                                                        1.783380e+18
    Fn
               1.587249e+19
                            1.650104e+19
                                          1.715449e+19
                                                        1.783380e+19
     . . .
              1.237513e-01 1.237513e-01
                                         1.237513e-01
                                                       1.237512e-01
    _h__1
             -3.309027e+20 -3.309027e+20 -3.440065e+20 -3.576292e+20
    _L__1
    _K_HD__1 1.929487e+20 1.929487e+20 2.005895e+20 2.085328e+20
    _I_h__1
              7.349721e+18 7.349721e+18 7.640770e+18 7.943344e+18
    _ph__1
              1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
                      4
                                     5
                                                                7
                                                                      \
                                                   6
    С
               5.315142e+18 5.525621e+18
                                          5.744436e+18 5.971915e+18
              4.758941e+18 4.947395e+18
    C_w
                                          5.143312e+18 5.346987e+18
                            5.782262e+17
                                          6.011239e+17
                                                        6.249284e+17
    C_k
               5.562007e+17
    FD
               1.854002e+18
                            1.927421e+18
                                          2.003746e+18 2.083095e+18
               1.854002e+19
                            1.927421e+19
                                          2.003746e+19 2.083095e+19
    Fn
    _h__1
               1.237512e-01 1.237512e-01
                                         1.237512e-01
                                                       1.237512e-01
    L 1
             -3.717913e+20 -3.865142e+20 -4.018202e+20 -4.177323e+20
    _K_HD__1 2.167907e+20 2.253757e+20
                                          2.343005e+20 2.435788e+20
    _I_h__1
              8.257900e+18 8.584913e+18 8.924876e+18 9.278301e+18
    _ph__1
               1.000000e+00 1.000000e+00
                                          1.000000e+00
                                                        1.000000e+00
                      8
                                     9
                                                       1041
                                                                      1042 \
    С
               6.208403e+18 6.454256e+18
                                               1.493965e+36 1.552977e+36
    C_w
               5.558728e+18
                                           ... 1.337334e+36
                                                            1.390159e+36
                            5.778853e+18
    C_k
               6.496756e+17
                            6.754027e+17
                                               1.566313e+35
                                                             1.628183e+35
    FD
               2.165585e+18
                            2.251342e+18
                                               5.221045e+35 5.427276e+35
               2.165585e+19
                            2.251342e+19
                                               5.221045e+36 5.427276e+36
    Fn
                                           . . .
     . . .
              1.237512e-01 1.237511e-01
                                           ... 1.234375e-01
                                                             1.234375e-01
    _h__1
                                           ... -1.050300e+38 -1.091786e+38
             -4.342745e+20 -4.514717e+20
                                               6.110271e+37 6.351627e+37
    _K_HD__1 2.532246e+20 2.632522e+20
```

```
9.645722e+18 1.002769e+19 ... 2.321844e+36 2.413557e+36
_I_h__1
_ph__1
         1.000000e+00
                      1.000000e+00
                                    ... 1.000000e+00 1.000000e+00
                 1043
                               1044
                                            1045
                                                          1046 \
С
         1.614319e+36 1.678085e+36 1.744369e+36 1.813272e+36
C_w
         1.445070e+36 1.502150e+36
                                    1.561485e+36 1.623164e+36
C_k
         1.692496e+35 1.759350e+35
                                    1.828844e+35 1.901083e+35
FD
         5.641653e+35 5.864499e+35
                                    6.096146e+35 6.336944e+35
         5.641653e+36 5.864499e+36
                                    6.096146e+36 6.336944e+36
Fn
_h__1
         1.234375e-01 1.234375e-01 1.234375e-01 1.234375e-01
        -1.134912e+38 -1.179741e+38 -1.226341e+38 -1.274781e+38
_L__1
_K_HD__1 6.602516e+37 6.863316e+37 7.134417e+37 7.416226e+37
         2.508893e+36 2.607994e+36 2.711010e+36 2.818095e+36
_I_h__1
         1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
_ph__1
                                                          1050
                 1047
                               1048
                                            1049
С
         1.884896e+36 1.959350e+36
                                    2.036744e+36 2.117195e+36
C_w
         1.687279e+36 1.753926e+36
                                   1.823206e+36
                                                 1.895223e+36
C_k
         1.976176e+35 2.054235e+35
                                    2.135377e+35 2.219725e+35
FD
         6.587254e+35 6.847450e+35 7.117924e+35 7.399082e+35
         6.587254e+36 6.847450e+36 7.117924e+36 7.399082e+36
Fn
. . .
_h__1
         1.234375e-01 1.234375e-01
                                   1.234375e-01
                                                 1.234375e-01
_L__1
        -1.325135e+38 -1.377478e+38 -1.431888e+38 -1.488448e+38
_K_HD__1 7.709167e+37 8.013679e+37 8.330220e+37 8.659263e+37
_I_h__1
         2.929409e+36 3.045121e+36 3.165403e+36 3.290437e+36
         1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
_ph__1
[67 rows x 1051 columns]
```

[36]: plot\_shock(shock = shock, filename = 'Shock\_3.png')



```
[37]: shock3 = shock.round(decimals = 3).tail(1).transpose().loc['alpha':,:] shock3.columns = ['$\Delta rm$'] shock3

[37]: $\Delta rm$
```

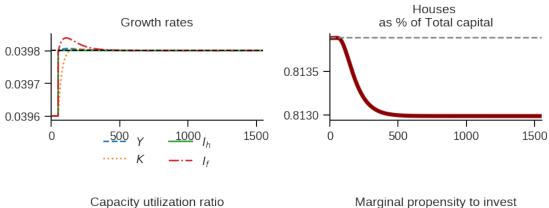
```
[37]:
     alpha
                1.000000e+00
     alpha_2
                3.000000e-01
                9.00000e-01
     gamma_F
     gamma_u
                1.000000e-02
     omega
                3.000000e-01
                2.000000e-02
     rm
     spread_1
                0.000000e+00
     spread_mo
                5.000000e-03
     un
                8.000000e-01
     v
                2.500000e+00
                4.000000e-02
    phi_0
    phi_1
                2.000000e-02
    phparam
                1.000000e+00
     infla
                0.000000e+00
     gZn
                3.900000e-02
     _K_f__1
                1.974191e+37
               -6.225215e+37
     _M__1
    _MO__1
                8.659263e+37
     _Lf__1
               -1.488448e+38
     _h__1
                1.230000e-01
               -1.488448e+38
     _L__1
```

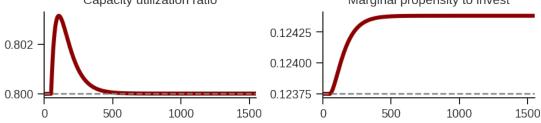
```
_K_HD__1 8.659263e+37
_I_h__1 3.290437e+36
_ph__1 1.000000e+00
```

#### 9.4 Increase in inflation

```
[38]: base = model()
df = SolveSFC(base, time=1000)
shock = ShockModel(base_model=base, create_function=model(), variable='infla',__
increase=0.01, time = 1500)

plot_shock(shock = shock, filename = 'Shock_4.png')
```





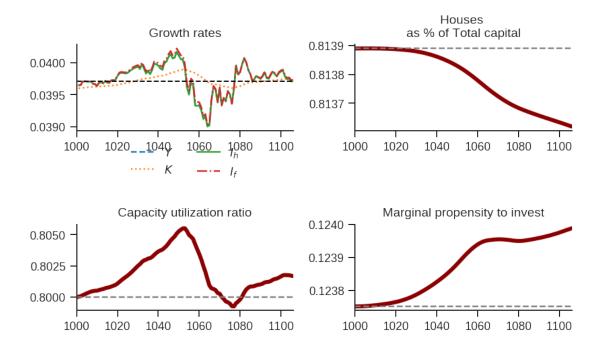
```
[39]: shock4 = shock.round(decimals = 3).tail(1).transpose().loc['alpha':,:] shock4.columns = ['$\Delta $ Infla'] shock4
```

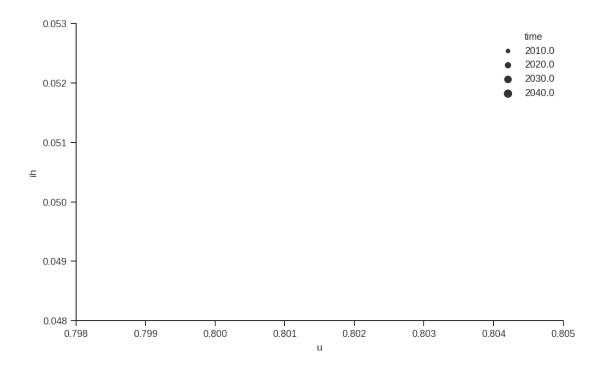
```
[39]:
                 $\Delta $ Infla
     alpha
                    1.000000e+00
                    3.000000e-01
     alpha_2
     gamma_F
                    9.000000e-01
                    1.000000e-02
     gamma_u
                    3.000000e-01
     omega
                    2.000000e-02
     rm
                    0.000000e+00
     spread_1
     spread_mo
                    0.000000e+00
```

```
8.00000e-01
un
               2.500000e+00
phi_0
              4.000000e-02
phi_1
               2.000000e-02
phparam
               1.000000e+00
infla
               1.000000e-02
gZn
              3.900000e-02
_K_f__1
              7.896508e+45
_M__1
             -2.427467e+46
_MO__1
              3.432732e+46
_Lf__1
             -5.860199e+46
_h__1
             1.240000e-01
_L__1
             -5.860199e+46
_K_HD__1
              3.432732e+46
_{
m I}_{
m h}_{
m 1}
              1.313996e+45
_ph__1
              3.004272e+06
```

#### 9.5 Plugging real data

```
[40]: base=model()
     initial=1000
     shock_duration=1
     SolveSFC(base, time=initial, table=False)
     for i in data.index:
         lagged = [key for key in base.solutions[-1].keys()]
         lagged = [i for i in lagged if "__" in i]
         for j in lagged:
             del base.solutions[-1][j]
         base.set_values(base.solutions[-1])
         base.set_values({
                 'own':data['Taxa Própria'][i],
                 'infla':data['Inflação'][i],
             })
         try:
             SolveSFC(base, time=shock_duration, table=False)
         except Exception as e:
             print(f'For own interest rate = {i}, {e}')
             pass
     shock = SFCTable(base)[initial:]
     plot_shock(shock = shock, filename = 'Shock_5.png')
```





### 10 Merging tables

```
[42]: | df = df.round(decimals = 3).tail(1).transpose().loc['alpha':,:]
     df.columns = ['Base scenario']
     df
     table = pd.merge(left = df, right = shock1, left_index = True, right_index = __
      →True)
     table = pd.merge(left = table, right = shock2, left_index = True, right_index = __
     table = pd.merge(left = table, right = shock3, left_index = True, right_index = __
      →True)
     table = pd.merge(left = table, right = shock4, left_index = True, right_index = __
     table.to_latex("table.tex", bold_rows = True, column_format = 'c', decimal = __
      →',', escape=False,)
     table
[42]:
                Base scenario
                                $\Delta \phi_0$
                                                 $\Delta \omega$
                                                                    $\Delta rm$
                                   1.000000e+00
                                                    1.000000e+00
     alpha
                 1.000000e+00
                                                                   1.000000e+00
                 3.00000e-01
                                   3.000000e-01
                                                    3.000000e-01
                                                                   3.000000e-01
     alpha_2
     gamma_F
                 9.00000e-01
                                   9.00000e-01
                                                    9.000000e-01
                                                                   9.00000e-01
                 1.000000e-02
                                   1.000000e-02
                                                    1.000000e-02
                                                                   1.000000e-02
     gamma_u
                 3.000000e-01
                                   3.000000e-01
                                                    2.500000e-01
                                                                   3.000000e-01
     omega
```

```
2.000000e-02
                              2.000000e-02
                                                2.000000e-02
                                                              2.000000e-02
rm
                              0.000000e+00
                                                              0.000000e+00
spread_1
            0.000000e+00
                                                0.000000e+00
spread_mo
            0.000000e+00
                              0.000000e+00
                                                0.000000e+00
                                                              5.000000e-03
un
            8.00000e-01
                              8.00000e-01
                                                8.00000e-01
                                                              8.00000e-01
            2.500000e+00
                              2.500000e+00
                                                2.500000e+00
                                                              2.500000e+00
v
phi_0
            4.000000e-02
                              5.000000e-02
                                                4.000000e-02
                                                              4.00000e-02
            2.000000e-02
                              2.000000e-02
                                                2.000000e-02
                                                              2.000000e-02
phi_1
                              1.000000e+00
phparam
            1.000000e+00
                                                1.000000e+00
                                                              1.000000e+00
infla
            0.000000e+00
                              0.000000e+00
                                                0.000000e+00
                                                              0.000000e+00
            3.900000e-02
                              3.922000e-02
                                                3.900000e-02
                                                              3.900000e-02
gZn
_K_f__1
            4.244058e+19
                              1.082934e+44
                                                3.585057e+30
                                                              1.974191e+37
_{M}_{1}
           -1.326991e+20
                             -1.611049e+44
                                               -1.223085e+31 -6.225215e+37
_MO__1
            1.855990e+20
                              3.598007e+44
                                                1.704715e+31
                                                              8.659263e+37
_Lf__1
           -3.182981e+20
                             -5.209056e+44
                                               -2.927800e+31 -1.488448e+38
                              1.550000e-01
                                                              1.230000e-01
_h__1
            1.240000e-01
                                                1.240000e-01
_L__1
           -3.182981e+20
                             -5.209056e+44
                                               -2.927800e+31 -1.488448e+38
K_HD_1
            1.855990e+20
                              3.598007e+44
                                                1.704715e+31
                                                              8.659263e+37
            7.069758e+18
                              1.700278e+43
                                                6.493526e+29
                                                              3.290437e+36
_I_h__1
_ph__1
            1.000000e+00
                              1.000000e+00
                                                1.000000e+00
                                                              1.000000e+00
```

#### 1.000000e+00 alpha alpha\_2 3.000000e-01 gamma F 9.000000e-01 gamma\_u 1.000000e-02 omega 3.000000e-01 2.000000e-02 rmspread\_1 0.000000e+00 spread\_mo 0.000000e+00 8.000000e-01 un 2.500000e+00 v phi\_0 4.00000e-02 2.000000e-02 phi\_1 phparam 1.000000e+00 infla 1.000000e-02 gZn 3.900000e-02 \_K\_f\_\_1 7.896508e+45 $_{M}_{1}$ -2.427467e+46 \_MO\_\_1 3.432732e+46 \_Lf\_\_1 -5.860199e+46 \_h\_\_1 1.240000e-01 \_L\_\_1 -5.860199e+46 \_K\_HD\_\_1 3.432732e+46

\_I\_h\_\_1

\_ph\_\_1

\$\Delta \$ Infla

1.313996e+45

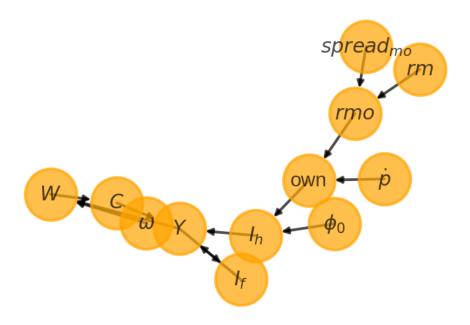
3.004272e+06

## 11 Dag

```
[43]: df = pd.DataFrame(
         {
             'Determinada' : [
                 '$Y$', # 1
                 #'$Y$', # 2
                 '$C$', # 3
                 '$W$', # 5
                 '$Y$', # 7
                 '$Y$', # 8
                 '$I_h$', # 9
                 'own', # 10
                 'own', # 11
                 '$I_f$', # 12
                 '$I_h$', # 13
                 '$rmo$', # 14
                 '$rmo$', # 15,
                 '$W$', # 16
             ],
             'Determina' : [
                 '$C$', # 1
                 #'$It$', # 2
                 '$W$', # 3
                 '$\omega$', # 5
                 '$I_f$', # 7
                 '$I_h$', # 8
                 'own', # 9
                 '$\dot p$', # 10
                 '$rmo$', # 11
                 '$Y$', # 12,
                 '$\phi_0$', # 13
                 '$rm$', # 14
                 '$spread_{mo}$', # 15
                 '$Y$', # 16
             ]
     }
     )
     # Build your graph
     G=nx.from_pandas_edgelist(
         df,
         'Determina',
         'Determinada',
         create_using=nx.DiGraph() # To use with arrows
     )
```

```
fig, ax = plt.subplots()
# Plot it
nx.draw(
    G,
    with_labels=True,
    node_color='orange',
    node_size=1500,
    edge_color='black',
    linewidths=2,
    width=2,
    arrows=True,
    font_size=15,
    alpha=.7,
    ax = ax,
    #pos=nx.spring_layout(G),
    pos=nx.fruchterman_reingold_layout(G),
    #pos=nx.shell_layout(G),
plt.show()
fig.savefig('Dag.png', dpi = 300)
```

```
/home/gpetrini/.local/lib/python3.6/site-
packages/networkx/drawing/nx_pylab.py:579: MatplotlibDeprecationWarning:
The iterable function was deprecated in Matplotlib 3.1 and will be removed in
3.3. Use np.iterable instead.
  if not cb.iterable(width):
/home/gpetrini/.local/lib/python3.6/site-
packages/networkx/drawing/nx_pylab.py:676: MatplotlibDeprecationWarning:
The iterable function was deprecated in Matplotlib 3.1 and will be removed in
3.3. Use np.iterable instead.
  if cb.iterable(node_size): # many node sizes
```



## 12 Analytical solution

```
[44]: base_eq = model()
    SolveSFC(base_eq, time=1, table = False)
    t = sp.Symbol('t')
    initials = {
        key: base_eq.evaluate(key) for key in base_eq.parameters
    }
    initials.update({key: base_eq.evaluate(key) for key in base_eq.variables})

for i in base_eq.variables:
    globals()["_" + i] = sp.Function(i)

for i in base_eq.parameters:
    globals()[i] = sp.symbols(i, positive=True)
    globals()['infla'] = sp.symbols('infla')
```

#### 12.1 General equations

```
[45]: Y = _C(t) + _I_t(t)
    pprint(sp.Eq(_Y(t), Y))
    I = _I_f(t) + _I_h(t)
    pprint(sp.Eq(_I_t(t), I))
    Yk = _K_f(t)/v
    pprint(sp.Eq(_Yk(t), Yk))
```

```
u = _Y(t)/_Yk(t)
pprint(sp.Eq(_u(t), u))
Z = _I_h(t)
pprint(sp.Eq(_Z(t), Z))
W = omega*_Y(t)
pprint(sp.Eq(_W(t), W))
K = _K_HD(t) + _K_f(t)
pprint(sp.Eq(_K(t), K))
```

#### 12.2 Households

C(t) = W(t)

```
[46]: C = alpha*_W(t)
    pprint(sp.Eq(_C(t), C))
    YD = _W(t) + _FD(t) + rm*_M(t-1) - _rmo(t)*_MO(t-1)
    pprint(sp.Eq(_YD(t), YD))
    S_h = _YD(t) - _C(t)
    pprint(sp.Eq(_S_h(t), S_h))
    dMO = _I_h(t)
    pprint(sp.Eq(_MO(t) - _MO(t-1), dMO))
    V_h = _M(t) + _K_HD(t) - _MO(t)
    pprint(sp.Eq(_V_h(t), V_h))
    NFW_h = _S_h(t) - _I_h(t)
    pprint(sp.Eq(_NFW_h(t), NFW_h))
```

```
⊔
⊶------
```

NameError Traceback (most recent call

→last)

<invthor-input-46-8afb2da33411> in <module>

<ipython-input-46-8afb2da33411> in <module>
2 pprint(sp.Eq(\_C(t), C))

```
3 YD = _W(t) + _FD(t) + rm*_M(t-1) - _rmo(t)*_MO(t-1)
---> 4 pprint(sp.Eq(_YD(t), YD))
5 S_h = _YD(t) - _C(t)
6 pprint(sp.Eq(_S_h(t), S_h))
```

NameError: name '\_YD' is not defined

#### **12.3** Firms

```
[]: I_f = _h(t)*_Y(t)
   pprint(sp.Eq(_I_f(t), I_f))
   dK_f = I_f(t)
   pprint(sp.Eq(_K_f(t) - _K_f(t-1), dK_f))
   Lf = _I_f(t) - _FU(t) + _L(t-1)
   pprint(sp.Eq(_Lf(t), Lf))
   FT = _FU(t) + _FD(t)
   pprint(sp.Eq(_FT(t), FT))
   FU = gamma_F*(_FT(t) - _rl(t)*_L(t-1))
   pprint(sp.Eq(_FU(t), FU))
   FD = (1 - gamma_F)*(_FT(t) - _rl(t)*_L(t-1))
   pprint(sp.Eq(_FD(t), FD))
   h = h(t-1)*gamma_u*(u(t)-un) + h(t-1)
   pprint(sp.Eq(_h(t), h))
   NFW_f = _FU(t) - _I_f(t)
   pprint(sp.Eq(_NFW_f(t), NFW_f))
   V_f = K_f(t) - L(t)
   pprint(sp.Eq(_V_f(t), V_f))
```

#### 12.4 Banks

```
[]: L = _Lf(t)
    pprint(sp.Eq(_L(t), L))
    M = (_L(t) - _L(t-1)) + (_MO(t) - _MO(t-1)) + _M(t-1)
    pprint(sp.Eq(_M(t), M))
    rmo = rm + spread_mo
    pprint(sp.Eq(_rmo(t), rmo))
    rl = rm + spread_l
    pprint(sp.Eq(_rl(t), rl))
    V_b = _L(t) + _MO(t) - _M(t)
    pprint(sp.Eq(_V_b(t), V_b))
    NFW_b = _rl(t)*_L(t-1) + _rmo(t)*_MO(t-1) - rm*_M(t-1)
    pprint(sp.Eq(_NFW_b(t), NFW_b))
```

#### 12.5 Residential Investment

```
[]: own = sp.Function('own')
   K_HS = K_HD(t)
   pprint(sp.Eq(_K_HS(t), K_HS))
   Is = Ih(t)
   pprint(sp.Eq(_Is(t), Is))
   dK HD = I h(t)
   pprint(sp.Eq(_K_HD(t) - _K_HD(t-1), dK_HD))
   I_h = (1+g_Z(t))*I_h(t-1)
   pprint(sp.Eq(_I_h(t), I_h))
   K_k = K_HD(t)/(K(t))
   pprint(sp.Eq(_K_k(t), K_k))
   ph = (1+infla)*_ph(t-1)
   pprint(sp.Eq(_ph(t), ph))
   own = ((1+_rmo(t))/(1+infla))-1
   pprint(sp.Eq(_own(t), own))
   g_Z = phi_0 - phi_1*_own(t)
   pprint(sp.Eq(_g_Z(t), g_Z))
```

#### 12.5.1 Stability condition

```
[]: g = sp.Function('g')
   gK = sp.Function('g_K')
   def replacer(express):
       #print("\nReplacing the initial values....")
       df = SolveSFC(model(), time=1)
       df = df.iloc[1, :]
       express = express.subs(alpha, df['alpha']).subs(
           omega, df['omega'])
       express = express.subs(un, df['un']).subs(
           gamma_u, df['gamma_u'])
       express = express.subs(
           infla, df['infla'])
       express = express.subs(phi_0, df['phi_0']).subs(
           phi_1,
           df['phi_1']).subs(rm, df['rm']).subs(
               spread_mo, df['spread_mo'])
       express = express.subs(rm, df['rm']).subs(
               spread_mo, df['spread_mo']).subs(v, df['v'])
       return express
```

```
[]: EqY = Y - Y(t)
   EqY = EqY.subs(_C(t), C).subs(_I_t(t), I)
   EqY = EqY.subs(If(t),If)
   EqY = EqY.subs(W(t), W)
   EqY = EqY.subs(_I_h(t), _Z(t))
   EqY = sp.solve(EqY, _Y(t))[0].collect(alpha).collect(omega)
   solY = EqY
   pprint(cse(solY, optimizations='basic')[1], use_unicode=True)
   print('dY/d alpha = ', EqY.diff(alpha))
   print('dY/d omega = ', EqY.diff(omega))
   print("\nGowth rate....")
   gY, h_{,} gz_{,} = sp.symbols('gY h gZ')
   gY_{-} = alpha * omega * gY + _h(t) * gY + _h(t) - _h(
       t - 1) + (Z(t) / Y(t)) * gZ(t) - gY
   gY_{-} = gY_{-}subs(g_{Z}(t), gz_{-}).subs(Y(t), solY).subs(g_{Z}(t), gz_{-})
   gY_{-} = gY_{-}.subs(h(t) - h(t - 1), h - h(t - 1))
   gY_{=} = sp.solve(gY_{,} gY)[0].collect(gz_{)}
   pprint(sp.Eq(g(t), sp.simplify(gY_)))
   print('\nd gY/ d alpha\n')
   pprint(gY_.diff(alpha))
   print('\nd gY/ d omega\n')
   pprint(gY_.diff(omega))
```

#### 12.5.2 Stability condition (I)

```
[]: own_ = sp.Symbol('own')
g_LR = gY_.subs(_u(t), un)
pprint(sp.Eq(g(t), g_LR))

Equ = _u(t)*(g(t) - gK(t)) + _u(t-1)
pprint(sp.Eq(_u(t), Equ))
g_K = (_h(t)*_u(t))/v
pprint(sp.Eq(gK(t), g_K))
Equ = _u(t)*(g(t) - g_K)
Equ = Equ.subs(g(t), gz_)
pprint(sp.Eq(_u(t), Equ))
print(sp.Eq(_u(t), Equ))
print(sp.latex(sp.Eq(_u(t), Equ)))

Eqh = _h(t)*gamma_u*(_u(t) - un)
pprint(sp.Eq(_h(t), Eqh))
print(sp.Latex(sp.Eq(_h(t), Eqh)))

print('\nBuilding Jacobian matrix and evaluating at u = un\n')
```

```
J = sp.Matrix([
             Eqh.diff(h(t)).subs(u(t), un).subs(h(t-1), h(t)).subs(h(t),
 \rightarrow g_LR*v/un),
             Eqh.diff(\underline{u}(t)).subs(\underline{u}(t), un).subs(\underline{h}(t-1), \underline{h}(t)).subs(\underline{h}(t),
 \rightarrowg LR*v/un)
         ],
         Γ
             Equ.diff(_h(t)).subs(_u(t), un).subs(_h(t-1), _h(t)).subs(_h(t),
 \rightarrow g_LR*v/un),
             Equ.diff(\underline{u}(t)).subs(\underline{u}(t), un).subs(\underline{h}(t-1), \underline{h}(t)).subs(\underline{h}(t),
 \rightarrowg LR*v/un)
         ]])
pprint(J)
print(sp.latex(J))
print('\nDeterminant:\n')
pprint(J.det()>0)
print(sp.latex(J.det()>0))
estability = sp.solve(J.det().subs(gz_, g_Z).subs(_own(t), own_)>0, own_)
print("\nStability condition\n")
pprint(estability)
estability = replacer(estability)
pprint(estability)
pprint(sp.Eq(own_, initials['own']))
estability = estability.subs(own_, own)
pprint(estability)
print('Rewriting')
estability = -(1+infla) + (1+rmo) < 2*(1+infla)
pprint(estability)
estability = sp.solve(estability, infla)
pprint(estability)
estability = estability.subs(spread_mo, initials['spread_mo']).subs(rm,__
 →initials['rm'])
pprint(estability)
print('\nTrace:\n')
pprint(J.trace()<0)</pre>
print(sp.latex(J.trace()<0))</pre>
print("\nStability condition\n")
estability = sp.solve(J.trace()<0).subs(gz_, g_Z)</pre>
pprint(estability)
estability = estability.subs(_own(t), own).subs(_rmo(t), rmo)
pprint(estability)
estability = estability.subs(spread_mo, initials['spread_mo']).subs(rm,__
 →initials['rm'])
```

```
estability = estability.subs(phi_0, initials['phi_0']).subs(phi_1,⊔
→initials['phi_1'])
pprint(estability)
```

#### 12.5.3 Stability condition (II)

### 12.6 Capacity utilization on the long-run

Consider *k* as the fraction between real housing and total capital (including households' capital):

$$k = \frac{K_h}{K}$$

The capacity utilization ration can be definede as:

$$u = \frac{Y \cdot v}{K \cdot (1 - k)}$$

So, dividing Y by houseolds' capital is the same as:

$$\frac{Y}{k \cdot K}$$

Multiplying by *v*:

$$\frac{Y}{k \cdot K} \cdot v = \frac{Y \cdot v}{K} \cdot \left(\frac{1}{k}\right)$$

Multiplying and dividing by 1 - k:

$$\frac{Y \cdot v}{K \cdot (1-k)} \cdot \left(\frac{1-k}{k}\right) = u \cdot \left(\frac{1-k}{k}\right)$$

Therefore,

$$Y \frac{v}{K_h} = u \cdot \left(\frac{1-k}{k}\right)$$
$$u = Y \frac{v}{K_h} \cdot \left(\frac{k}{1-k}\right)$$

```
[]: k = sp.Symbol('K_k')
   rel = solY*(v/_K_HD(t))*(k/(1-k))
   rel = rel.subs(_Z(t), Z)
   rel = rel.subs(_I_h(t)/_K_HD(t), g_Z)
   rel = rel.subs(_h(t), h)
   pprint(rel)
   print('\nFor the long run...\n')
   rel = rel.subs(u(t), un).subs(h(t-1), h(t))
   rel = rel.subs(_h(t), g_Z*v/un)
   rel = rel.subs(_own(t), own)
   rel = rel.subs(_rmo(t), rmo)
   pprint(sp.Eq(_u(t), rel))
   print(sp.latex(sp.Eq(_u(t), rel)))
   def collector(express):
       express = express.simplify().collect(phi_0).collect(phi_1).collect(v)
       express = express.collect(omega).collect(alpha).collect(un).collect(infla+1)
       express = express.simplify()
       return express
   rel = rel - un
   rel = sp.solve(rel,k/(1-k))[0]
   pprint(sp.Eq(k/(1-k), collector(rel)))
   print(sp.latex(sp.Eq(k/(1-k), collector(rel))))
   pprint(sp.Eq(k/(1-k), replacer(rel)))
   print(f"Error = {replacer(rel) - base.evaluate('K_k/(1-K_k)')}")
   print('\n' + '='*70)
   print('\nChange in income distribution:\n')
   result = collector(rel.diff(omega))
   pprint(result)
   print('\nReplacing ...\n')
   result = replacer(result)
   print(result)
   print(f'd (k/1-k)/ d omega > 0? \{result > 0\}')
```

```
print('\n' + '='*70)
print('\nChange in spread:\n')
result = collector(rel.diff(spread_mo))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd (k/1-k)/ d spread > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in inflation:\n')
result = collector(rel.diff(infla))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd (k/1-k)/ d inflation > 0? \{result > 0\}')
print('\n' + '='*70)
print('\nChange in autonomous component:\n')
result = collector(rel.diff(phi_0))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd (k/1-k)/ d phi_0 > 0? \{result > 0\}')
print('\n' + '='*70)
print('\nHousing as % of total Capital\n')
rel = rel*(1-k) - k
rel = sp.solve(rel, k)[0]
pprint(sp.Eq(k, collector(rel)))
print(sp.latex(sp.Eq(k, collector(rel))))
print(f"Error = {replacer(rel) - base.evaluate('K_k')}")
print(f"Error = {replacer(1-rel) - base.evaluate('(g_Z*v/un)/
→(1-alpha*omega)')}")
print('\n' + '='*70)
print('\nChange in income distribution:\n')
result = collector(rel.diff(omega))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d omega > 0? {result > 0}')
print('\n' + '='*70)
```

```
print('\nChange in spread:\n')
result = collector(rel.diff(spread_mo))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d spread > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in inflation:\n')
result = collector(rel.diff(infla))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d inflation > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in autonomous component:\n')
result = collector(rel.diff(phi_0))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d phi_0 > 0? {result > 0}')
print('\n' + '='*70)
print('\nFirms capital as % of total Capital:\n')
rel = (g_Z*v/un)/(1-omega*alpha)
rel = rel.subs(_own(t), own).subs(_rmo(t), rmo)
pprint(sp.Eq(1-k, rel))
print('\n' + '='*70)
print('\nChange in income distribution:\n')
result = collector(rel.diff(omega))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d omega > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in spread:\n')
result = collector(rel.diff(spread_mo))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
```

```
print(result)
print(f'd k/ d spread > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in inflation:\n')
result = collector(rel.diff(infla))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d inflation > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in autonomous component:\n')
result = collector(rel.diff(phi_0))
pprint(result)
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d phi_0 > 0? {result > 0}')
print('\n' + '='*70)
print('\nRevisiting Houses as % of total Capital:\n')
rel = 1 - rel
pprint(sp.Eq(k, rel))
print(sp.latex(sp.Eq(k, rel)))
print('\n' + '='*70)
print('\nChange in income distribution:\n')
result = collector(rel.diff(omega))
pprint(result)
print(sp.latex(result))
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d omega > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in spread:\n')
result = collector(rel.diff(spread_mo))
pprint(result)
print(sp.latex(result))
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d spread > 0? {result > 0}')
```

```
print('\n' + '='*70)
print('\nChange in inflation:\n')
result = collector(rel.diff(infla))
pprint(result)
print(sp.latex(result))
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d inflation > 0? {result > 0}')
print('\n' + '='*70)
print('\nChange in autonomous component:\n')
result = collector(rel.diff(phi 0))
pprint(result)
print(sp.latex(result))
print('\nReplacing ...\n')
result = replacer(result)
print(result)
print(f'd k/ d phi_0 > 0? {result > 0}')
```

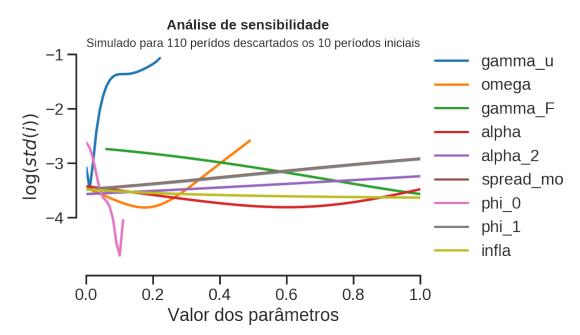
### 13 Steady State

```
[]: \#pprint(sp.Eq(Y / \_K\_f(t), I\_f / K\_HS)) \\ \#pprint(sp.Eq(K\_HS / \_K\_f(t), I\_f / Y)) \\ \#pprint(sp.Eq(rel, I\_f / Y)) \\ \#flow = I\_f / Y \\ \#flow = flow.subs(\_Y(t), Y).subs(\_h(t), (g\_Z.subs(\_own(t), own).subs(\_rmo(t), \_) \\ \rightarrow rmo))*v/un) \\ \#pprint(sp.Eq(rel, flow)) \\ \#ss = sp.solve(sp.Eq(rel, flow), omega)[0] \\ \#pprint(sp.Eq(omega, ss)) \\ \#pprint(sp.Eq(omega, replacer(ss)))
```

#### 14 Sobol

```
'phi_1',
    'infla'.
]
def sobol(
   bound = np.linspace(0,1,101),
   time = 10,
   skip = 10,
   parameters = parameters
):
   t2 = datetime.now()
   bound = bound
   df = pd.DataFrame()
   empty_list = [i for i in range(len(bound))]
   for param in parameters:
       for i in range(len(bound)):
           base = model()
           base.set_values({param:bound[i]})
           try:
               empty_list[i] = np.log(SolveSFC(base,time=time+skip)["u"][skip:
\rightarrow].std())
           except Exception as e:
               empty_list[i] = np.infty
               pass
       df[param] = empty_list ############## Replace here
df.index = bound
   sns.set_context('talk')
   fig, ax = plt.subplots()
   df.plot(
       ax = ax,
       lw = 2.5
   )
   ax.ticklabel_format(useOffset=False)
   ax.set_ylabel("$\log(std(i))$")
   ax.set_xlabel("Valor dos parâmetros")
   ax.legend(loc='center left', bbox_to_anchor=(1, 0.5))
   fig.suptitle("Análise de sensibilidade", fontsize = 14, weight="bold")
   ax.set_title("Simulado para {} perídos descartados os {} períodos iniciais".
 →format(time+skip,skip), fontsize = 12, y = .98)
   sns.despine(offset=10, trim=True);
```

```
ylim = ax.get_ylim()
         plt.show()
         print("Total running time: ", datetime.now()-t2)
         return df
[48]: sobol(time=100)
```



Total running time: 0:11:40.250208

```
[48]:
            gamma_u
                        omega
                                 gamma F
                                             alpha
                                                     alpha_2 spread_mo
                                                                             phi_0 \
     0.00 -3.081586 -3.453268
                                     inf -3.422897 -3.565038
                                                               -3.476689 -2.617742
     0.01 -3.476689 -3.478576
                                     inf -3.430510 -3.562300
                                                              -3.472092 -2.716890
     0.02 -2.996520 -3.504268
                                     inf -3.438168 -3.559546
                                                              -3.467438 -2.915440
     0.03 -2.420178 -3.530262
                                     inf -3.445868 -3.556776
                                                               -3.462728 -3.198037
     0.04 -2.031450 -3.556476
                                     inf -3.453611 -3.553990
                                                              -3.457961 -3.476689
     0.96
                inf
                          inf -3.531529 -3.531065 -3.252454
                                                               -2.935542
                                                                                inf
     0.97
                inf
                          inf -3.540162 -3.517688 -3.248877
                                                               -2.930463
                                                                                inf
     0.98
                inf
                          inf -3.548632 -3.504162 -3.245296
                                                               -2.925420
                                                                                inf
     0.99
                inf
                           inf -3.556927 -3.490487 -3.241711
                                                               -2.920412
                                                                                inf
                                                              -2.915440
     1.00
                           inf -3.565038 -3.476689 -3.238122
                inf
                                                                                inf
              phi_1
                        infla
```

```
0.01 -3.481228 -3.481272
0.02 -3.476689 -3.485708
0.03 -3.472092 -3.490000
0.04 -3.467438 -3.494154
... ...
0.96 -2.945803 -3.628871
0.97 -2.940656 -3.629317
0.98 -2.935542 -3.629757
0.99 -2.930463 -3.630190
1.00 -2.925420 -3.630616

[101 rows x 9 columns]
```

# 15 Finishing

```
[]: print("Total running time: ", datetime.now() - t1)
print("Simulation running time: ", t2 - t1)
print("Sobol running time: ", datetime.now() - t2)
```

# 16 Shock yourself