Sensitivity Analysis of Biochemical Systems Using Bond Graphs: Additional Material for Pentose Phosphate Pathway

Peter Gawthrop. peter.gawthrop@unimelb.edu.au

January 17, 2023

Contents

1	Introduction	1
2	Set up the code 2.1 Import packages 2.2 Steady-state by simulation 2.3 Stoichiometry 2.4 Linearisation 2.5 Extract subsystem from linear system 2.6 Plotting	1 4 4 4 5
3	Normalisation constants	8
4	Stoichiometric approach	8
5	Import Pentose Phosphate Pathway information from: 5.1 Repeat simulations for sanity check	9 12
6	6.1 Linearise sensitivity system	
7	Sloppy parameter analysis	89

1 Introduction

This notebook generates the figures for the paper: Sensitivity Analysis of Biochemical Systems Using Bond Graphs - Pentose Phosphate Pathway Example

2 Set up the code

2.1 Import packages

```
[1]: ## Some useful imports
import BondGraphTools as bgt
import numpy as np
```

```
import sympy as sp
import matplotlib.pyplot as plt
import control as con
import copy
import importlib
import time
## For reimporting: use imp.reload(module)
import importlib as imp
# ## For path etc: sys.path
import sys
sys.path.append("/home/peterg/WORK/Research/SystemsBiology/lib/python")
## Stoichiometric analysis
import stoich as st
## SVG
import svgBondGraph as sbg
## Stoichiometry to BG
import stoichBondGraph as stbg
## Modularity
import modular as bgm
## Sloppy parameters
import sloppy as slp
## Display (eq disp.SVG(), disp.
import IPython.display as disp
##
quiet = True
Plotting = False
Titles = True
```

Warning - scikit.odes not found. Simulations are disabled.

```
[2]: def tfProps(tf,method='truncate'):
    ## Steady-state gain
    g = con.dcgain(tf)

if not con.issiso(tf):
    g = g[0][0]

## Time constant
    ## Check if direct link
    direct_link = np.any(tf.D)
```

```
## Only set tau=0 if siso
if direct_link and con.issiso(tf):
    ## Instant response
    tau = 0
else:
    ## Reduce to first-order to estimate time constant
    ## Note that method='matchdc' can give a kernel crash - use 'truncate'
    tf1 = con.balred(sys,orders=1,method=method)
    poles = con.poles(tf1)
    realPoles = np.real(poles)
    tau = -1/min(realPoles)
```

```
[3]: ## High frequency gain function (initial response to step)
     def hfgain(sys):
         tf = con.ss2tf(sys)
         num = tf.num[0][0]
         den = tf.den[0][0]
         n = len(den)
         m = len(num)
          print(den)
         print(num)
          print(n,m)
         if n==m:
            g = num[0]/den[0]
         elif n>m:
             g = 0
         elif n<m:
             g = float('inf')
         return g
     \# q = hfqain(sys)
     # print(g)
```

```
[4]: ## Optional plotting
def Savefig(name):
    if Plotting:
        plotname = 'Figs/'+name+'.pdf'
        print('Saving',plotname)
        plt.rcParams.update({'font.size': 20})
        plt.tight_layout()
        plt.savefig(plotname)
        plt.clf()
```

2.2 Steady-state by simulation

2.3 Stoichiometry

```
[6]: def__
      →Stoichiometry(model,chemostats=[],flowstats=[],CommonSpecies=None,sensitivity=False):
         ## Stoichiometry
         s = st.stoich(model,quiet=quiet)
         ## Unify species
         if not (CommonSpecies is None):
             commonSpecies = st.merge(s,CommonSpecies=CommonSpecies)
               print(commonSpecies)
             st.unify(s,commonSpecies=commonSpecies)
         ## Sensitivity
         if sensitivity:
             extra = st.stoichSensitivity(s)
             extra = []
          print(chemostats+extra)
         ## Chemostats and flowstats
         sc = st.statify(s,chemostats=chemostats+extra)
         sf = st.statify(s,flowstats=flowstats)
         return s,sc,sf
```

2.4 Linearisation

2.5 Extract subsystem from linear system

```
[8]: def Index(A,a):
         I = []
         for aa in a:
             i = A.index(aa)
             I.append(i)
         return np.array(I)
     def zapSmall(x,tol=1e-10,quiet=True):
         xx = np.zeros(len(x))
         for i,val in enumerate(x):
             if abs(val)>tol:
                 xx[i] = x[i]
             else:
                 if not quiet:
                     print(f'Setting {i}th coefficient {val:.2} to zero')
         return xx
      -extractSubsystem(SYS,sc,sf,inp,outp,tol=1e-3,order=None,minreal=False,quiet=False):
         Sys = copy.copy(SYS)
         chemostats = sc['chemostats']
         if sf is None:
             flowstats = []
         else:
             flowstats = sf['flowstats']
         species = sc['species']
         reaction = sc['reaction']
```

```
## Index of input and output
if inp[0] in chemostats:
    i_inp = Index(chemostats,inp)
      print('Input:',i_inp,chemostats[i_inp[0]])
else:
    i_inp = Index(flowstats,inp)+len(chemostats)
 print(i_inp)
if outp[0] in chemostats:
      i_outp = Index(chemostats,outp)
    i_outp = Index(species,outp)
elif outp[0] in species:
    i_outp = Index(species,outp)
else:
    if outp[0] in reaction:
        i_outp = Index(reaction,outp)
    else:
        print(f'Output {outp} does not exist')
## Extract tf
n_y = len(i_outp)
n_u = len(i_ip)
nn = Sys.A.shape
n_x = nn[0]
#print(n_x)
sys = con.ss(Sys.A,
             Sys.B[:,i_inp].reshape(n_x,n_u),
             Sys.C[i_outp,:].reshape(n_y,n_x),
             Sys.D[i_outp][:,i_inp].reshape(n_y,n_u))
if minreal:
    sys = con.minreal(sys,tol=tol,verbose=True)
## Reduce order
if not (order is None):
    sys = con.balred(sys,order,method='matchdc')
return sys
```

2.6 Plotting

```
else:
        plt.plot(t,y_step,label=label)
    for reac in reactions:
        i = s['reaction'].index(reac)
        if plotSim:
              label = reac + r' (\$ \land lambda\$ = ' + f' \{ lam-1:0.2f \})'
            pc = int(round(100*(lam-1)))
            if labeling:
                label = f'{reac} ({pc}%)'
            else:
                label = None
            plt.plot(t,(dat['V'][:,i]-V_ss[i])/(lam-1),
                     lw=5,ls='dashed',label=label)
    plt.grid()
    if labeling:
        plt.legend()
    plt.xlabel('$t$')
   plt.ylabel(r'$\tilde{v}/\tilde{\lambda}$')
     plt.ylim(bottom=0)
#
    plt.xlim(left=0)
     plt.tight_layout()
```

```
[10]: def_
       →plotSensitivitydX(dat,species=['A','C'],plotSim=True,name=None,labeling=True,setZero=False
          if plotSim:
              plt.plot(t,y_step,color='black')
          else:
              plt.plot(t,y_step,label=label)
          for spec in species:
              print(spec)
              i_s = s['species'].index(spec)
              if plotSim:
                  pc = int(round(100*(lam-1)))
                  if labeling:
                      label = f'\{spec\} (\{pc\}\%)'
                  else:
                      label=None
                  plt.plot(t,(dat['dX'][:,i_s]-dX_ss[i_s])/(lam-1),
                            lw=5,ls='dashed',label=label)
          plt.grid()
          if labeling:
              plt.legend()
          plt.xlabel('$t$')
          plt.ylabel(r'$\Delta \mathbf{f} /\Delta \lambda$')
```

```
#plt.ylabel(r'$\tilde{\dot{x}}/\tilde{\lambda}$')
if setZero:
    plt.ylim(bottom=0)
    plt.xlim(left=0)
# plt.tight_layout()
```

3 Normalisation constants

```
[12]: T_human = 37  # Human body temperature
K_0 = 273.15
print(f'T_human = {T_human} degC = {T_human+K_0} K')

mu_0 = RT = st.RT(T_cent=T_human)
print(f'mu_0 = {mu_0*1e-3:0.3f} kJ/mol')

F = st.F()  # Faraday's constant
print(f'F = {F*1e-3:0.2f} kC/mol')

V_0 = RT/F
print(f'V_0 = {V_0*1e3:0.2f} mV')

P_0 = 1e-3

v_0 = P_0/mu_0
print(f'v_0 = {v_0*1e6:0.4f} micro mol /s')

i_0 = F*v_0
print(f'i_0 = {i_0*1e3:0.2f} mA')

T_human = 37 degC = 310.15 K
```

mu_0 = 2.579 kJ/mol F = 96.49 kC/mol V_0 = 26.73 mV v_0 = 0.3878 micro mol /s i_0 = 37.42 mA

4 Stoichiometric approach

```
sys = extractSubsystem(Sys,sc,sf,inp=inp,outp=outp,order=order,tol=tol)
## Time
t = np.linspace(0,t_last,500)
## Sensitivity step response
step = con.step_response(sys,T=t)
y_step = step.y[:,0,:].T
## Exact simulation with changed parameter or state
X_s_1 = copy.copy(X_s)
parameter1 = copy.copy(parameter)
inComp = inp[0][1:]
if inComp in chemostats:
    ## Perturb state
    iComp = s['species'].index(inComp)
    X_ss_1[iComp] = lam*X_ss_1[iComp]
else:
    ## Perturb parameter
    if inComp in s['species']:
        parname = 'K_'+inComp
        parname = 'kappa_'+inComp
   print(parname)
    if parname in list(parameter.keys()):
        parameter1[parname] = lam*parameter[parname]
    else:
        parameter1[parname] = lam
dat = st.sim(s,sc=sc,sf=sf,t=t,parameter=parameter1,X0=X_ss_1,quiet=quiet)
return dat, y_step, t, sys
```

5 Import Pentose Phosphate Pathway information from:

Peter J. Gawthrop, Michael Pan, and Edmund J. Crampin. *Modular dynamic biomolecular modelling with bond graphs: the unification of stoichiometry, thermodynamics*, kinetics and data. Journal of The Royal Society Interface, 18(181):20210478, 2021. doi:10.1098/rsif.2021.0478.

```
[14]: ## Load data
import pickle
filename = 'GlyPPP.pic'
datafile = open(filename, 'rb')
GlyPPP = pickle.load(datafile)

s0 = GlyPPP['s']
```

```
parameter = GlyPPP['parameter']
print(parameter)
sc0 = GlyPPP['sc']
chemostats = sc0['chemostats']
print(chemostats)
conc = GlyPPP['conc']
import json
print(json.dumps(parameter,indent=2))
{'kappa_PGI': 154.38997002578546, 'kappa_PFK': 54.85043355391926, 'kappa_FBA':
160.07691981318658, 'kappa_TPI': 353.9323765134668, 'kappa_G6PDH2R':
4.667672687418436, 'kappa_PGL': 291.6414393949941, 'kappa_GND':
1.2695594575280742, 'kappa_RPI': 4206.975085828208, 'kappa_TKT2':
9.170830408091838, 'kappa_TALA': 1.6563170906188465, 'kappa_TKT1':
8.819978413932159, 'kappa_RPE': 96.07053469867635, 'K_6PGC': 6.233520699319548,
'K_6PGL': 1.0397283862123778, 'K_ADP': 5.154605959675528, 'K_ATP':
2.2539233202427162, 'K_CO2': 33.794182951886775, 'K_DHAP': 1.778985981513151,
'K_E4P': 57.93534155565906, 'K_F6P': 1.41401983409913, 'K_FDP':
0.38795235349448437, 'K_G3P': 14.902044199279699, 'K_G6P': 0.8377378837717493,
'K_H': 0.3491739027654742, 'K_H2O': 1.0397283862123778, 'K_NADP':
11747.063287701598, 'K_NADPH': 21.002665768445333, 'K_R5P': 12.241945837528652,
'K_RU5PD': 86.15506601742248, 'K_S7P': 21.751324335743256, 'K_XU5PD':
51.68293204643235}
['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', 'NADPH', 'R5P', 'G3P']
  "kappa_PGI": 154.38997002578546,
  "kappa_PFK": 54.85043355391926,
  "kappa_FBA": 160.07691981318658,
  "kappa_TPI": 353.9323765134668,
  "kappa_G6PDH2R": 4.667672687418436,
  "kappa_PGL": 291.6414393949941,
  "kappa_GND": 1.2695594575280742,
  "kappa_RPI": 4206.975085828208,
  "kappa_TKT2": 9.170830408091838,
  "kappa_TALA": 1.6563170906188465,
  "kappa_TKT1": 8.819978413932159,
  "kappa_RPE": 96.07053469867635,
  "K_6PGC": 6.233520699319548,
  "K_6PGL": 1.0397283862123778,
  "K_ADP": 5.154605959675528,
  "K_ATP": 2.2539233202427162,
  "K_CO2": 33.794182951886775,
  "K_DHAP": 1.778985981513151,
  "K_E4P": 57.93534155565906,
  "K_F6P": 1.41401983409913,
  "K_FDP": 0.38795235349448437,
  "K_G3P": 14.902044199279699,
  "K_G6P": 0.8377378837717493,
```

```
"K_H": 0.3491739027654742,

"K_H2O": 1.0397283862123778,

"K_NADP": 11747.063287701598,

"K_NADPH": 21.002665768445333,

"K_R5P": 12.241945837528652,

"K_RU5PD": 86.15506601742248,

"K_S7P": 21.751324335743256,

"K_XU5PD": 51.68293204643235
```

[15]: ## Reactions

disp.Latex(st.sprintrl(s0,chemformula=True))

[15]:

$$G_6P \stackrel{PGI}{\longleftarrow} F_6P$$
 (1)

$$ATP + F_6P \xrightarrow{PFK} ADP + FDP + H \tag{2}$$

$$FDP \stackrel{FBA}{\longleftarrow} DHAP + G_3P \tag{3}$$

$$DHAP \xrightarrow{TPI} G_3P \tag{4}$$

$$G_6P + NADP \xrightarrow{G_6PDH_2R} {}_{6}PGL + H + NADPH$$
 (5)

$$_{6}\mathrm{PGL} + \mathrm{H}_{2}\mathrm{O} \xrightarrow{\mathrm{PGL}} _{6}\mathrm{PGC} + \mathrm{H}$$
 (6)

$$_{6}$$
PGC + NADP $\stackrel{\text{GND}}{\longleftarrow}$ CO₂ + NADPH + RU₅PD (7)

$$RU_5PD \stackrel{RPI}{\longleftarrow} R_5P$$
 (8)

$$E_4P + XU_5PD \stackrel{TKT_2}{\longleftarrow} F_6P + G_3P$$
 (9)

$$G_3P + S_7P \xrightarrow{TALA} E_4P + F_6P$$
 (10)

$$R_5P + XU_5PD \stackrel{TKT_1}{\longleftarrow} G_3P + S_7P$$
 (11)

$$RU_5PD \stackrel{RPE}{\longleftarrow} XU_5PD$$
 (12)

[16]: ## Pathway reactions
sp0 = st.path(s0,sc0)
disp.Latex(st.sprintrl(sp0,chemformula=True))

[16]:

$$ATP + G_6P \stackrel{pr_1}{\rightleftharpoons} ADP + 2G_3P + H$$
 (13)

$$G_6P + H_2O + 2 \text{ NADP} \xrightarrow{pr_2} CO_2 + 2 H + 2 \text{ NADPH} + R_5P$$
 (14)

$$2 \operatorname{H}_2 \operatorname{O} + 4 \operatorname{NADP} + \operatorname{R}_5 \operatorname{P} \overset{\operatorname{pr}_3}{\longleftarrow} 2 \operatorname{CO}_2 + \operatorname{G}_3 \operatorname{P} + 4 \operatorname{H} + 4 \operatorname{NADPH} \tag{15}$$

5.1 Repeat simulations for sanity check

```
[17]: def setPath(s,path='R5P'):
         print('\n Path =', path)
          if path == 'R5P':
              chemostats = ['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', _
       →'NADPH', 'R5P']
              flowstats =['G6PDH2R']
              dX_G6P = 5
          elif path == 'NADPH':
              chemostats = ['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', 'NADPH']
              flowstats = []
              dX_G6P = 1
          elif path == 'both':
              chemostats = ['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', _
       →'NADPH', 'R5P']
              flowstats = ['PGI', 'TKT2']
              dX_G6P = 1
          elif path == 'all':
              chemostats = ['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', __
       → 'NADPH', 'R5P', 'G3P']
             flowstats = []
              dX_G6P = 10
          elif path == 'all_e4p':
              chemostats = ['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', __

¬'NADPH', 'R5P','G3P','E4P']
             flowstats = []
              dX_G6P = 10
          sc = st.statify(s,chemostats=chemostats)
          sf = st.statify(s,flowstats=flowstats)
          return sc,sf,dX_G6P
[18]: ## Data normalisation
      c_0 = 7.88 \# mM
      t_0 = 7.946218487394957 \# sec
[19]: ## Simulate for each path
      s = copy.copy(s0) ## for compatibility
      spec = s['species']
      approximateFlowstats = True
      Spec = ['G6P','R5P','NADPH','ADP','C02','H','H20']
      #paths = ['all', 'both', 'R5P', 'NADPH']
      #paths = ['R5P']
      #paths = ['all', 'both', 'NADPH', 'R5P']
      paths = ['both','NADPH','R5P','all_e4p','all']
```

```
RATIO = \{\}
for path in paths:
    Ratio = {}
    normalisedRatio = {}
    ## Set up pathway
    sc,sf,dX_G6P_0 = setPath(s,path=path)
    ## Print pathway
    sp = st.path(s,sc)
    print(st.sprintrl(sp))
    ## Set up parameters
    par = copy.copy(parameter)
    if approximateFlowstats:
        small = 1e-3
        par = copy.copy(parameter)
        for fs in sf['flowstats']:
            par['kappa_'+fs] = small
        sf = None
    ## Simulate
    t = np.linspace(0,3*t_0,1000)
      ## Find steady-state with no flowstats
      dat_ss = st.sim(s,sc=sc,sf=sf,t=t,parameter=parameter,X0=conc)
      X_ss = dat_ss['X'][-1,:]
    dat = st.sim(s,sc=sc,sf=sf,t=t,parameter=par,X0=conc)
    #st.plot(s,dat,species=[])
     st.plot(s, dat, reaction=[], species=Spec, dX=True)
    ## Extract some external flows
    DX = dat['dX']
    dX = \{\}
    for Sp in Spec:
        dX[Sp] = DX[:,spec.index(Sp)]
        Ratio[Sp] = -dX[Sp]/dX['G6P']
        normalisedRatio[Sp] = -dX_G6P_0*dX[Sp]/dX['G6P']
    RATIO[path] = normalisedRatio
    ## Print steady-state values
    for Sp in Spec:
        ratio = Ratio[Sp][-1]
        print(f'{Sp}:\t{dX[Sp][0]:3.1f} \t{dX[Sp][-1]:3.}
 \hookrightarrow1f}\t{(dX_G6P_0*ratio):3.1f}\t{100*ratio:3.1f}\%')
```

Path = both

```
\begin{align}
G6P + H2O + 2 NADP &\Leftrightarrow CO2 + 2 H + 2 NADPH + R5P \\
ADP + 4 H2O + 8 NADP + 2 R5P &\Leftrightarrow ATP + 4 CO2 + G6P + 7 H + 8 NADPH
\ensuremath{\mbox{end}\{\mbox{align}\}}
G6P:
        -11.6
                -11.6
                        -1.0
                                -100.0%
                11.5
                        1.0
R5P:
        6.2
                                99.4%
NADPH: 23.2
                23.2
                        2.0
                                200.0%
                        0.0
ADP:
        63.1
                0.0
                                0.4%
CO2:
        11.6
                11.6
                        1.0
                                100.0%
        86.3
                23.2
                        2.0
                                200.4%
Η:
H20:
        -11.6
               -11.6
                        -1.0
                                -100.0%
 Path = NADPH
\begin{align}
ADP + G6P + 6 H2O + 12 NADP &\Leftrightarrow ATP + 6 CO2 + 11 H + 12 NADPH
\end{align}
        -71.1
                -1.9
                        -1.0
                                -100.0%
G6P:
R5P:
        6.2
                0.0
                        0.0
                                0.0%
NADPH: 23.2
                23.1
                        12.0
                                1200.0%
ADP:
        63.1
               -1.9
                        -1.0
                                -100.0%
CO2:
        11.6
               11.6
                        6.0
                                600.0%
Η:
        86.3
                21.2
                        11.0
                                1100.0%
H20:
        -11.6
               -11.6
                        -6.0
                                -600.0%
Path = R5P
\begin{align}
G6P + H2O + 2 NADP &\Leftrightarrow CO2 + 2 H + 2 NADPH + R5P \\
ADP + 4 H2O + 8 NADP + 2 R5P & Leftrightarrow ATP + 4 CO2 + G6P + 7 H + 8 NADPH
\end{align}
G6P:
        -59.5
                -2.4
                        -5.0
                                -100.0%
R5P:
        6.2
                2.9
                        6.0
                                120.0%
NADPH: 11.6
                0.0
                        0.0
                                0.2%
ADP:
        63.1
                0.5
                        1.0
                                20.0%
CO2:
                0.0
                        0.0
        11.6
                                0.1%
Η:
        74.7
                0.5
                        1.0
                                20.2%
H20:
        -11.6
                -0.0
                        -0.0
                                -0.1%
Path = all_e4p
\begin{align}
ATP + G6P &\Leftrightarrow ADP + 2 G3P + H \\
G6P + H2O + 2 NADP &\Leftrightarrow CO2 + 2 H + 2 NADPH + R5P \
G3P + R5P &\Leftrightarrow 2 E4P \\
E4P + H2O + 2 NADP &\Leftrightarrow CO2 + G3P + 2 H + 2 NADPH
\end{align}
G6P:
        -71.1
                -71.1
                        -10.0
                                -100.0%
R5P:
        6.2
                6.2
                        0.9
                                8.7%
```

32.6%

NADPH: 23.2

23.2

3.3

```
ADP:
             63.1
                     63.1
                             8.9
                                     88.8%
     CO2:
             11.6
                     11.6
                             1.6
                                     16.3%
     Н:
             86.3
                     86.3
                             12.1
                                     121.3%
     H20:
             -11.6 -11.6
                             -1.6
                                     -16.3%
      Path = all
     \begin{align}
     ATP + G6P &\Leftrightarrow ADP + 2 G3P + H \\
     G6P + H2O + 2 NADP &\Leftrightarrow CO2 + 2 H + 2 NADPH + R5P \\
     2 H2O + 4 NADP + R5P &\Leftrightarrow 2 CO2 + G3P + 4 H + 4 NADPH
     \end{align}
     G6P:
             -71.1
                     -71.1
                             -10.0
                                     -100.0%
     R5P:
             6.2
                     6.2
                             0.9
                                     8.7%
     NADPH: 23.2
                     23.2
                             3.3
                                     32.6%
     ADP:
             63.1
                     63.1
                             8.9
                                     88.8%
     CO2:
             11.6
                    11.6
                             1.6
                                     16.3%
     H:
             86.3
                     86.3
                             12.1
                                     121.3%
                                     -16.3%
     H20:
             -11.6 -11.6
                             -1.6
[20]: ## Set up chemostats to correspond to above path
      chemostats = sc['chemostats']
      Chemostats = copy.copy(sc['chemostats'])
      print(path)
      print(chemostats)
     all
     ['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', 'NADPH', 'R5P', 'G3P']
[21]: ## Plot ratios
      plottingRatios = False
      if plottingRatios:
          name = ['i','ii','iii']
          for sp in ['R5P','NADPH']:
              BigFont = 24
              plt.rcParams.update({'font.size': BigFont})
              for i,path in enumerate(['both','R5P','NADPH']):
                  Ratio = RATIO[path]
                  label = f'Path {name[i]} ({path})'
                  plt.plot(t/t_0,Ratio[sp],label=label,linewidth=5)
              if sp=='R5P':
                  ylim = 8
              else:
                  ylim=15
              plt.ylim((0,ylim))
              ylabel = r'\$\rho_{i+sp+'}
              plt.ylabel(ylabel)
              plt.xlabel('$t/t_0$')
              plt.legend()
             plt.grid()
                if SaveFig:
```

```
# plt.savefig(f'Figs/{sp}.pdf',bbox_inches='tight')
# plt.show()
```

6 Sensitivity System

```
[22]: ## Create sensitivity system
s = copy.copy(s0)
extra = st.stoichSensitivity(s)
chemostats += extra
sc = st.statify(s,chemostats=chemostats)
```

[23]: ## Reactions
disp.Latex(st.sprintrl(s,all=True,chemformula=True))

[23]:

$$G_6P + sG_6P + sPGI \xrightarrow{PGI} F_6P + sF_6P + sPGI \qquad (16)$$

$$ATP + F_6P + sATP + sF_6P + sPFK \xrightarrow{PFK} ADP + FDP + H + sADP + sFDP + sH + sPFK \qquad (17)$$

$$FDP + sFDP + sFBA \xrightarrow{FBA} DHAP + G_3P + sDHAP + sG_3P + sFBA \tag{18}$$

$$DHAP + sDHAP + sTPI \xrightarrow{TPI} G_3P + sG_3P + sTPI$$
 (19)

$$G_6P + NADP + sG_6P + sNADP + sG_6PDH_2R \xrightarrow{G_6PDH_2R} _6PGL + H + NADPH + s_6PGL + sH + sNADPH + sO(20)$$

$$_6$$
PGL + $_6$ PGL + $_6$ PGL + $_6$ PGC + $_6$ P

$$_{6}$$
PGC + NADP + $_{5}$ PGC + $_{5}$ NADP + $_{5}$ PGC + $_{5}$ NADP + $_{5}$ PGC + $_{5}$ NADPH + $_{5}$ PD + $_{5}$ CO $_{2}$ + $_{5}$ NADPH + $_{5}$ PD + $_{5}$ CO $_{2}$ + $_{5}$ NADPH + $_{5}$ CO $_{2}$ + $_{5}$ CO $_{2}$ CO $_{2}$ + $_{5}$ CO $_{2}$ + $_{5}$ CO $_{2}$ CO $_{2}$ CO $_{2}$ CO $_{2}$ CO $_{2}$ CO $_{2}$ C

$$RU_5PD + sRU_5PD + sRPI \xrightarrow{RPI} R_5P + sR_5P + sRPI$$
 (23)

$$E_4P + XU_5PD + sE_4P + sXU_5PD + sTKT_2 \xrightarrow{TKT_2} F_6P + G_3P + sF_6P + sG_3P + sTKT_2 \tag{24}$$

$$G_3P + S_7P + sG_3P + sS_7P + sTALA \xrightarrow{TALA} E_4P + F_6P + sE_4P + sF_6P + sTALA$$
(25)

$$R_5P + XU_5PD + sR_5P + sXU_5PD + sTKT_1 \xrightarrow{TKT_1} G_3P + S_7P + sG_3P + sS_7P + sTKT_1 \tag{26}$$

$$RU_5PD + sRU_5PD + sRPE \xrightarrow{RPE} XU_5PD + sXU_5PD + sRPE$$
 (27)

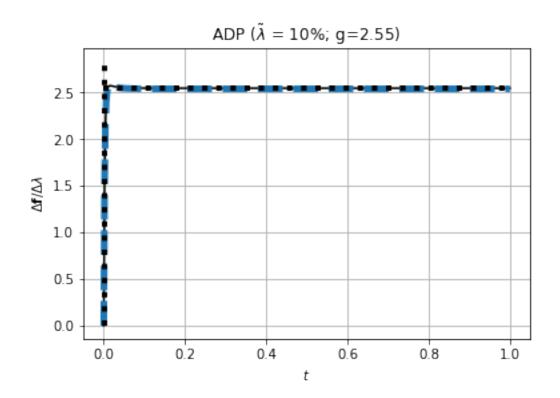
```
[24]: ## Initial conditions
             XO = np.hstack((conc,np.ones(s0['n_X']),np.ones(s0['n_V'])))
              # print(X0)
             print(X0.shape)
            (50,)
[25]: ## Linearise
              \# Sys, X_ss, V_ss, dX_ss = Linear(s, sc, X0=X0)
              \# Sys, X\_ss, V\_ss, dX\_ss = Linear(s, sc, parameter=parameter, X0=X0, invar='X', U=X0, inv
               →outvar = 'V', quiet=False)
[26]:  # species = s['species']
              # reaction = s['reaction']
[27]: | # print(chemostats)
              # print(reaction)
              # print(species)
              # # print(s0['Z'])
[28]: \# sys = extractSubsystem(Sys,sc,sf,inp=['sG6P'],outp=['RPI'],order=3)
            6.1
                        Linearise sensitivity system
[29]: ## Linearise dX as outvar
             Sys, X_ss, V_ss, dX_ss = Linear(s, sc, parameter=parameter, X0=X0, invar='X', __
               →outvar = 'dX', quiet=False)
            Steady-state finder error: 2.10e-14
            V_{ss} = [59.52330023 \ 63.11688312 \ 63.11688312 \ 63.11688312 \ 11.57830405 \ 11.57830405
              11.57830405 7.98472116 1.79679144 1.79679144 1.79679144 3.59358289]
            Setting K_6PGC to 6.233520699319548
            Setting K_6PGL to 1.0397283862123778
            Setting K_ADP to 5.154605959675528
            Setting K_ATP to 2.2539233202427162
            Setting K_CO2 to 33.794182951886775
            Setting K_DHAP to 1.778985981513151
            Setting K_E4P to 57.93534155565906
            Setting K_F6P to 1.41401983409913
            Setting K_FDP to 0.38795235349448437
            Setting K_G3P to 14.902044199279699
            Setting K_G6P to 0.8377378837717493
            Setting K_H to 0.3491739027654742
            Setting K_H2O to 1.0397283862123778
            Setting K_NADP to 11747.063287701598
            Setting K_NADPH to 21.002665768445333
            Setting K_R5P to 12.241945837528652
            Setting K_RU5PD to 86.15506601742248
            Setting K_S7P to 21.751324335743256
            Setting K_XU5PD to 51.68293204643235
            Setting kappa_PGI to 154.38997002578546
```

```
Setting kappa_PFK to 54.85043355391926
Setting kappa_FBA to 160.07691981318658
Setting kappa_TPI to 353.9323765134668
Setting kappa_G6PDH2R to 4.667672687418436
Setting kappa_PGL to 291.6414393949941
Setting kappa_GND to 1.2695594575280742
Setting kappa_RPI to 4206.975085828208
Setting kappa_TKT2 to 9.170830408091838
Setting kappa_TALA to 1.6563170906188465
Setting kappa_TKT1 to 8.819978413932159
Setting kappa_RPE to 96.07053469867635
```

6.2 Compute chemostat sensitivities

```
[30]: ## Chemostat sensitivities
     # Inp = ['sG6P', 'sR5P', 'sNADPH', 'sG3P']
     Outp = ['G6P','R5P','NADPH','G3P']
     Inp = []
     for chemo in sc['chemostats']:
         if not chemo[0] in ['s','H']:
             Inp.append('s'+chemo)
     print(Inp)
     Inp\_chemo = Inp
     # Inp = []
     # for reac in s['reaction']:
     # Inp.append('s'+reac)
     # Inp = ['sG6PDH2R']
     t_last=1
     lam = 1.1 # Perturbation parameter
     dcgain = {}
     Tau = \{\}
     for outp in Outp:
         for inp in Inp:
             print('Doing:',inp)
             sf = None
             order = None
             dat, y_step, t, sys =
      →simSensitivity(s,sc,sf,Sys,X_ss,V_ss,dX_ss,chemostats=Chemostats,parameter=parameter,
      g = con.dcgain(sys)
             g,tau = tfProps(sys)
             print('g =',g)
             name = inp[1:]
             if outp=='G6P':
                 dcgain[name] = -g
```

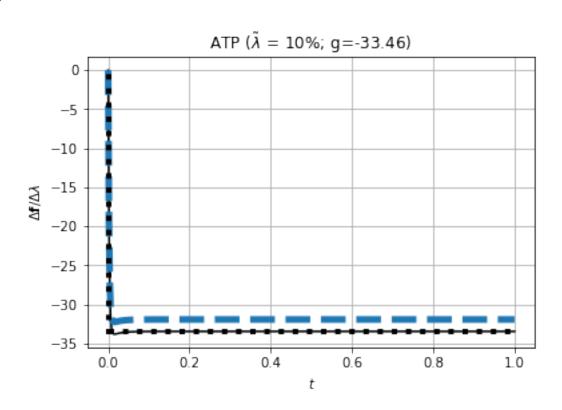
```
ylabel = '$-g_{\infty}
         else:
             dcgain[name] = g
             ylabel = '$g_{\infty}'
         Tau[name] = tau
         dlam = int((lam-1)*100)
        print('dlam =',dlam)
         if Titles:
             plt.title(f'\{name\}'+r' (tilde\abda = ' + f'\{dlam\}'; g=\{g: ...
 \hookrightarrow 2f)')
        plotSensitivitydX(dat,species=[outp],labeling=False)
           plt.hlines(g,min(t),max(t),color='black',ls='dashed')
        plotLines()
         Savefig('PPPdX_'+name+'_'+outp)
        plt.show()
     ## Barcharts
    ## DC gain
    if Titles:
        plt.title(outp)
    plt.tick_params(axis='x', rotation=90)
    plt.grid()
    plt.bar(range(len(dcgain)), dcgain.values(), align='center')
    plt.xticks(range(len(dcgain)), list(dcgain.keys()))
    plt.ylabel(ylabel)
    Savefig(f'PPPdX_{outp}_chem_bar')
    plt.show()
     ## Time constant
    if Titles:
        plt.title(outp)
    plt.tick_params(axis='x', rotation=90)
    plt.grid()
    plt.bar(range(len(Tau)), Tau.values(), align='center')
    plt.xticks(range(len(Tau)), list(Tau.keys()))
    plt.ylabel(r'$\tau$')
    Savefig(f'PPPdX_{outp}_chem_tau_bar')
    plt.show()
['sADP', 'sATP', 'sCO2', 'sG6P', 'sNADP', 'sNADPH', 'sR5P', 'sG3P']
Doing: sADP
g = 2.5484963469688147
dlam = 10
```



Doing: sATP

g = -33.46210603034223

dlam = 10

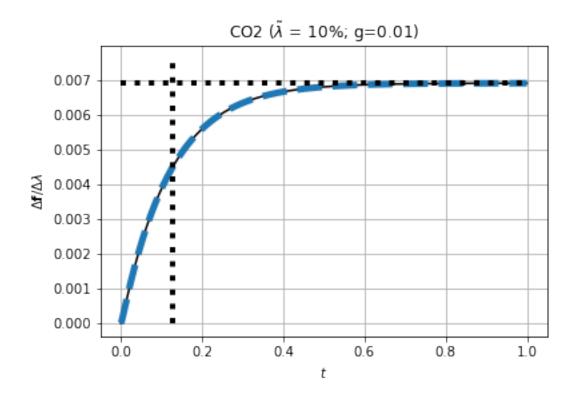


Doing: sCO2

g = 0.006924942429813596

dlam = 10

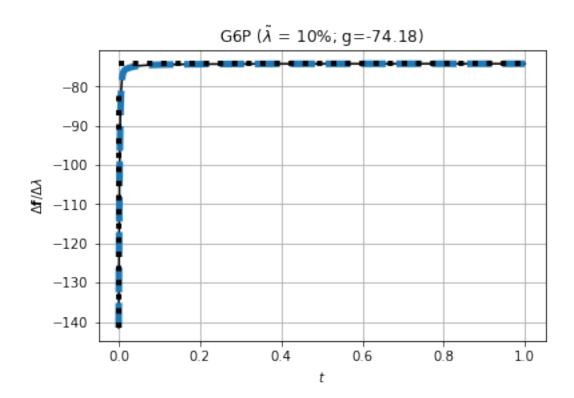
G6P



Doing: sG6P

g = -74.1814457534691

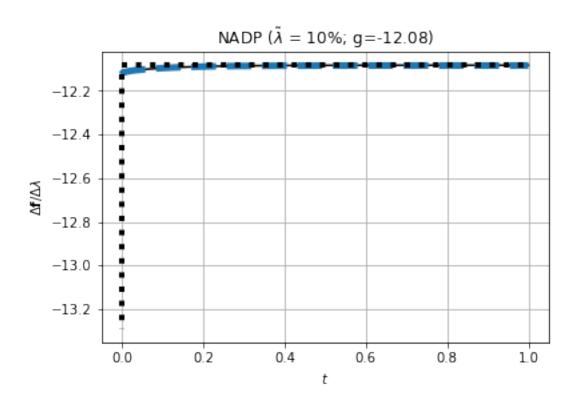
dlam = 10



Doing: sNADP

g = -12.083105188911864

dlam = 10

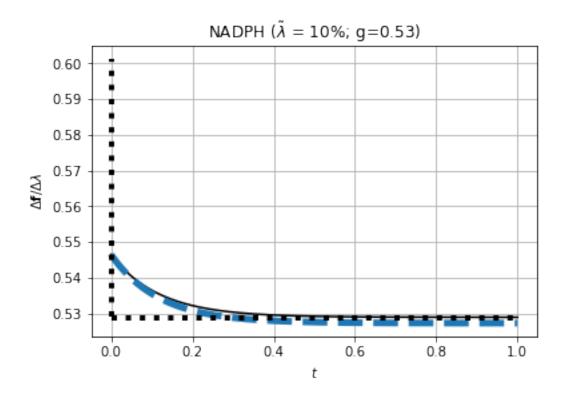


Doing: sNADPH

g = 0.5288892968979394

dlam = 10

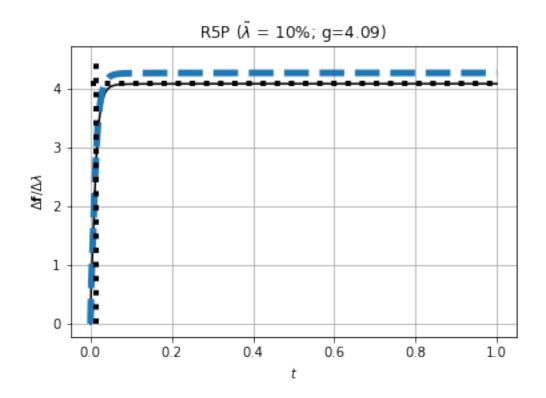
G6P



Doing: sR5P

g = 4.08711120268265

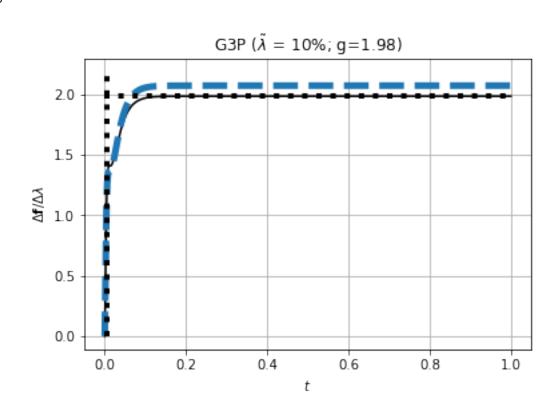
dlam = 10

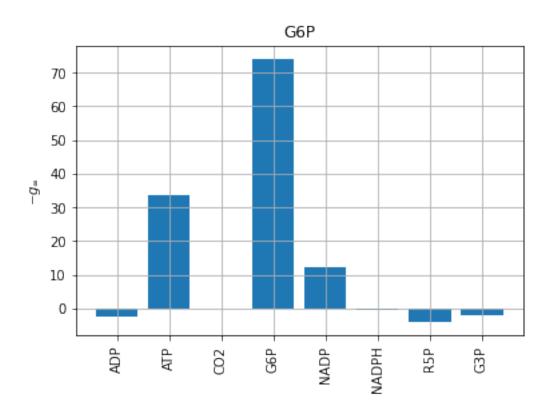


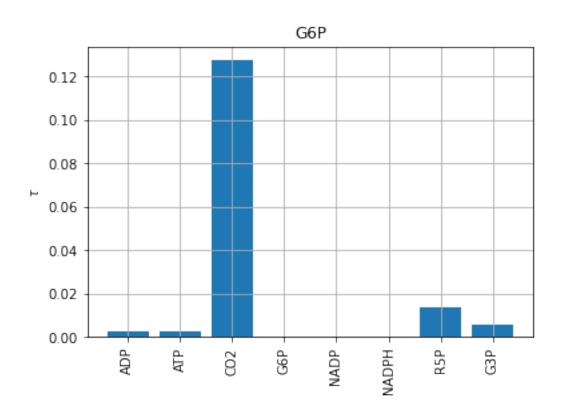
Doing: sG3P

g = 1.984521369029551

dlam = 10

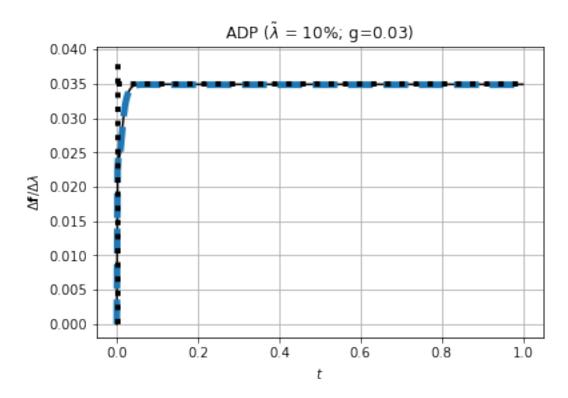






Doing: sADP

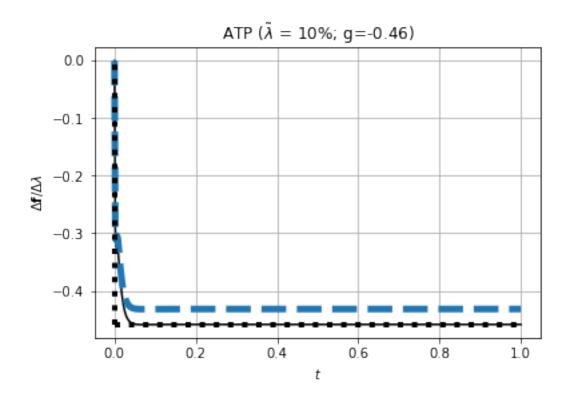
g = 0.0349248023890647dlam = 10 R5P



Doing: sATP

g = -0.4585674380195499

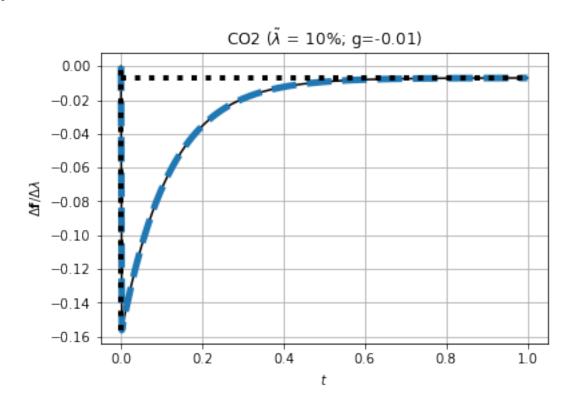
dlam = 10



Doing: sCO2

g = -0.006919387119948707

dlam = 10

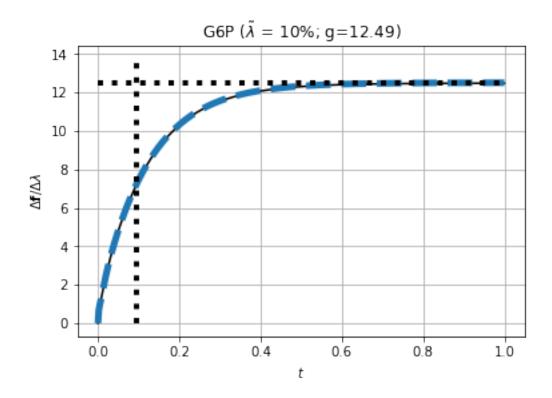


Doing: sG6P

g = 12.485591708850293

dlam = 10

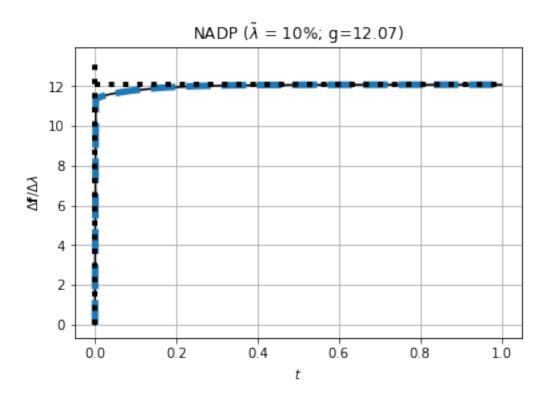
R5P



Doing: sNADP

g = 12.07341191065308

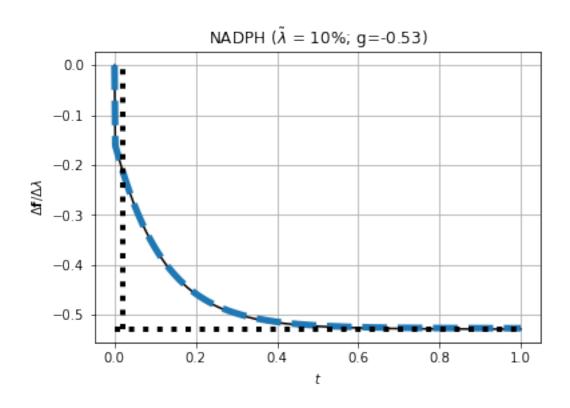
dlam = 10



Doing: sNADPH

g = -0.5284650126561967

dlam = 10

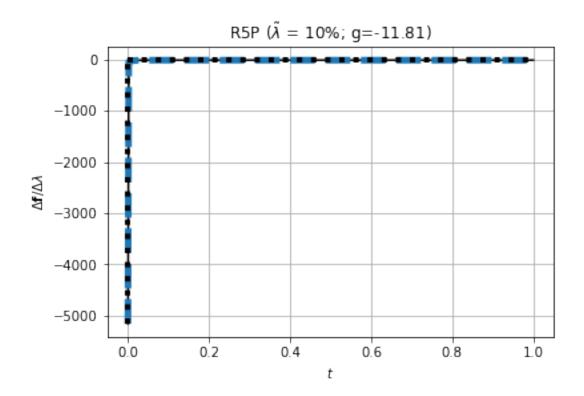


Doing: sR5P

g = -11.81229752462059

dlam = 10

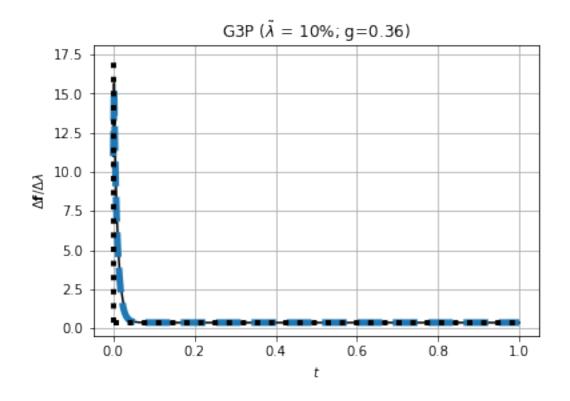
R5P

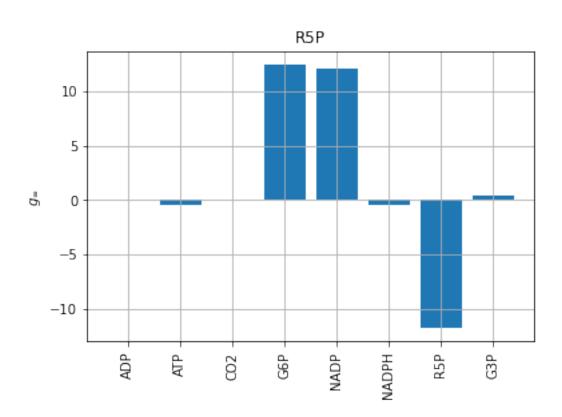


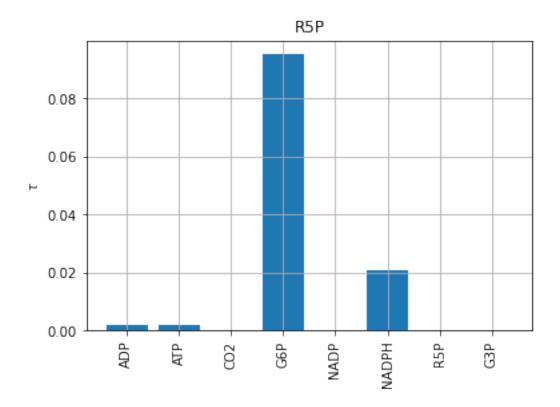
Doing: sG3P

g = 0.3634389561691993

dlam = 10

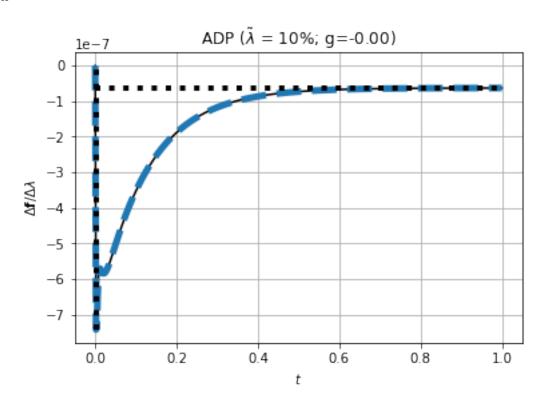






Doing: sADP

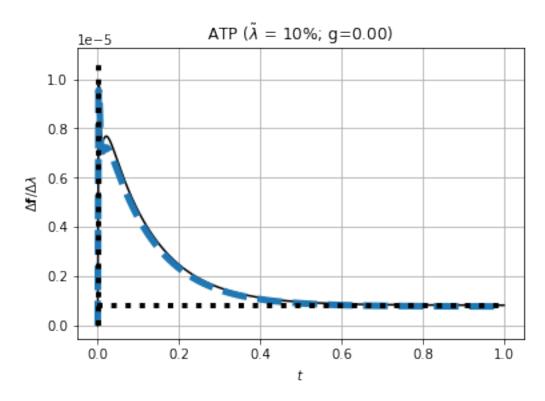
g = -6.262253882820029e-08



Doing: sATP

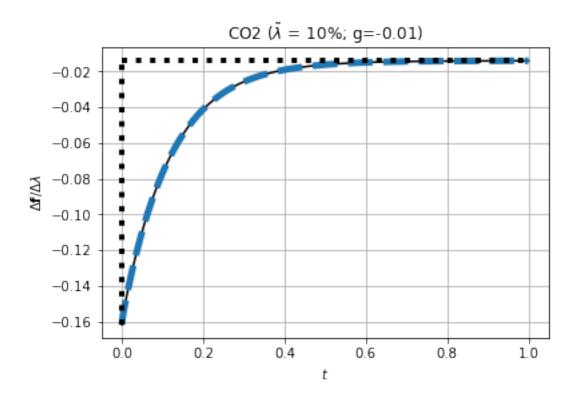
g = 8.22242510439288e-07

dlam = 10 NADPH



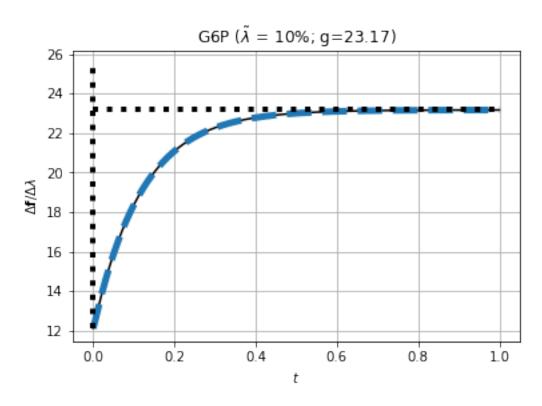
Doing: sCO2

g = -0.013855753145603666



Doing: sG6P

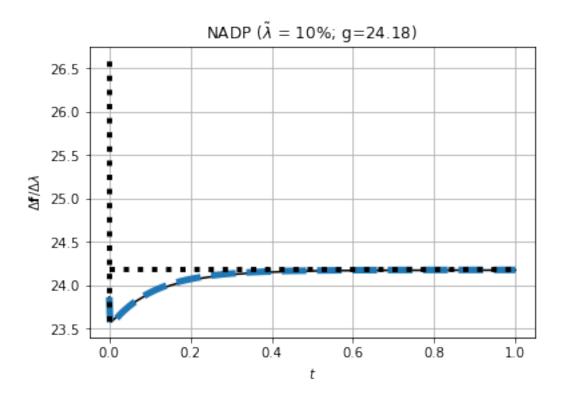
g = 23.170431095483337



Doing: sNADP

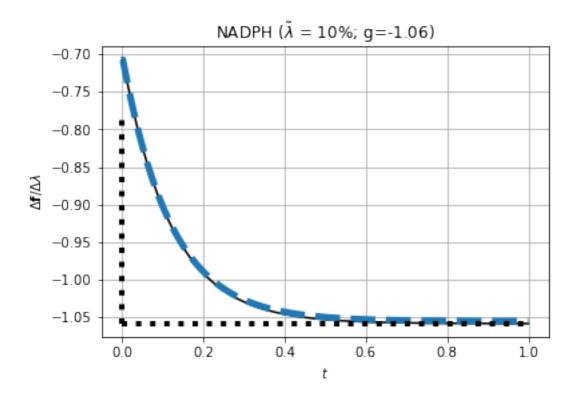
g = 24.176449757783388

dlam = 10 NADPH



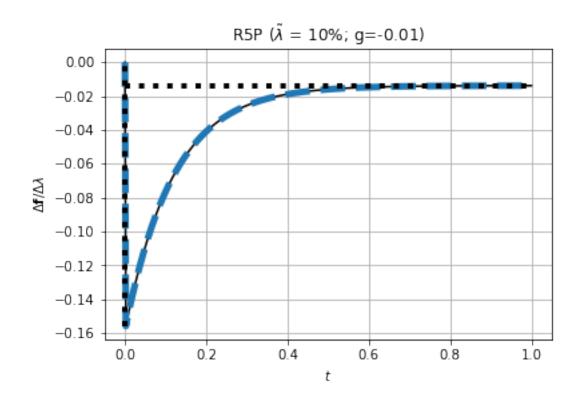
Doing: sNADPH

g = -1.058226781441594



Doing: sR5P

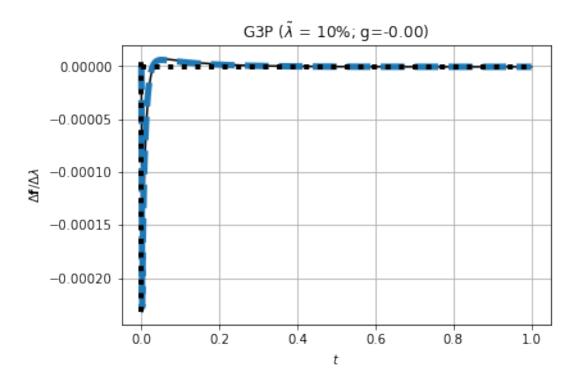
g = -0.013813109583456041

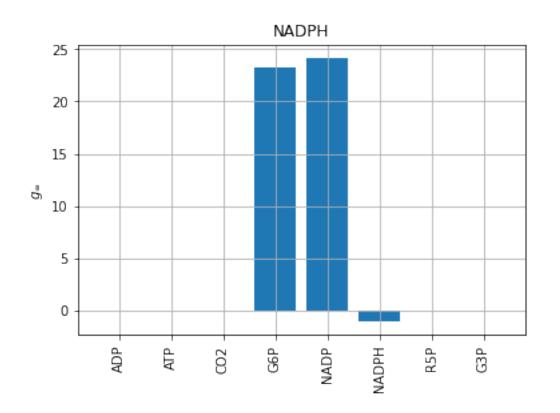


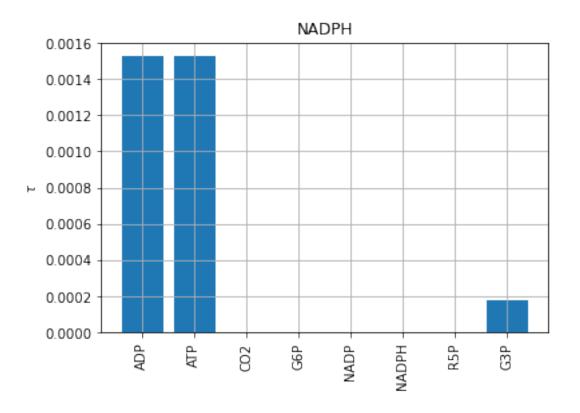
Doing: sG3P

g = -6.516705662261944e-07

dlam = 10 NADPH



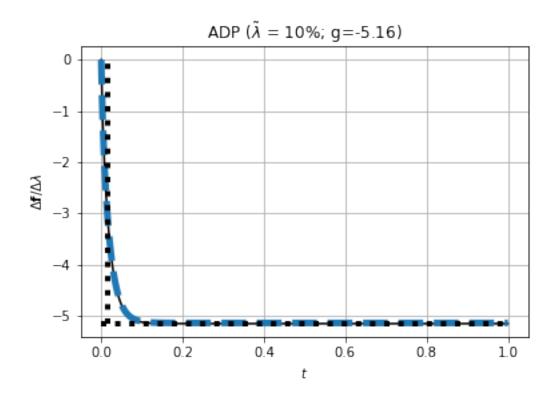




Doing: sADP

g = -5.155200687482189

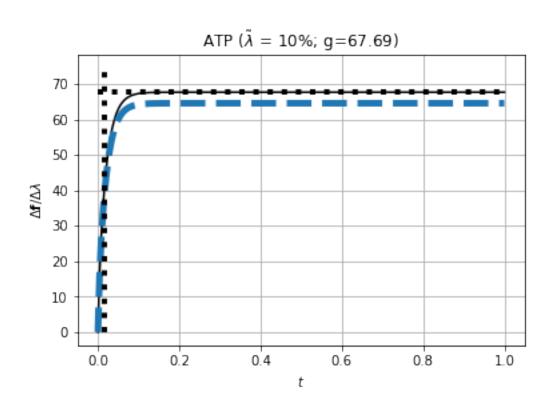
dlam = 10



Doing: sATP

g = 67.68849098700832

dlam = 10

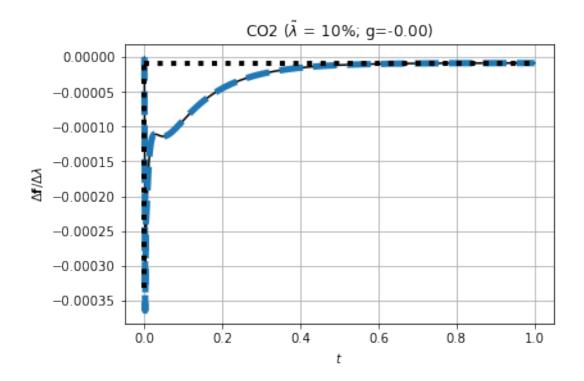


Doing: sCO2

g = -8.280802207064823e-06

dlam = 10

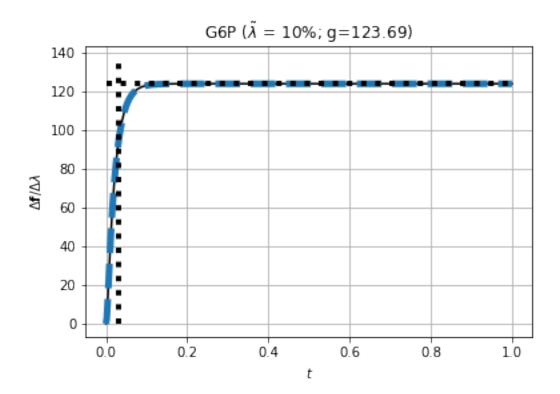
G3P



Doing: sG6P

g = 123.69183347627043

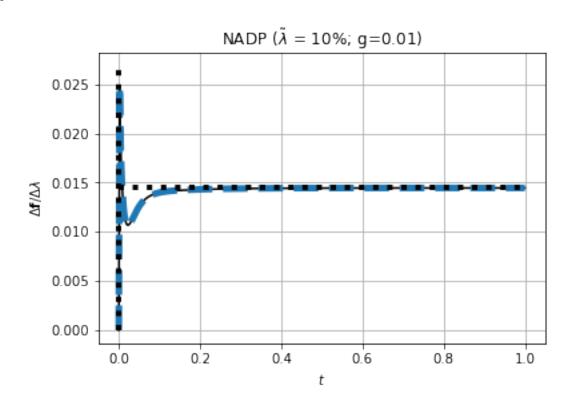
dlam = 10



Doing: sNADP

g = 0.014448900437827383

dlam = 10

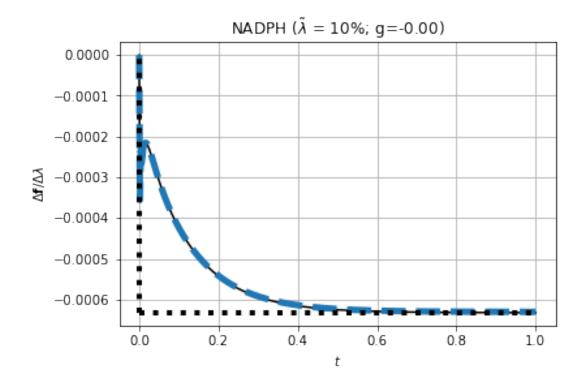


Doing: sNADPH

g = -0.0006324424618591407

dlam = 10

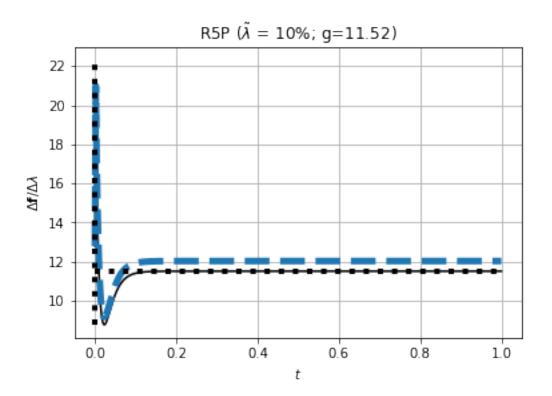
G3P



Doing: sR5P

g = 11.515242320596151

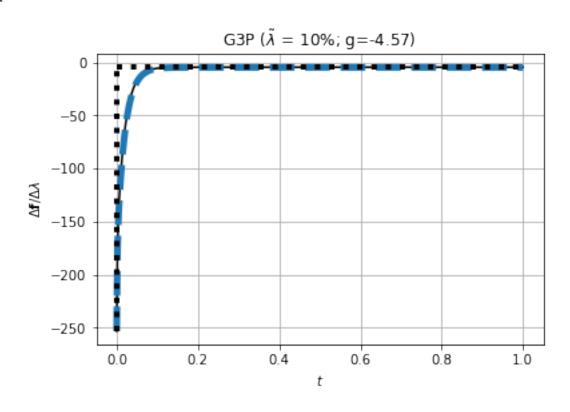
dlam = 10

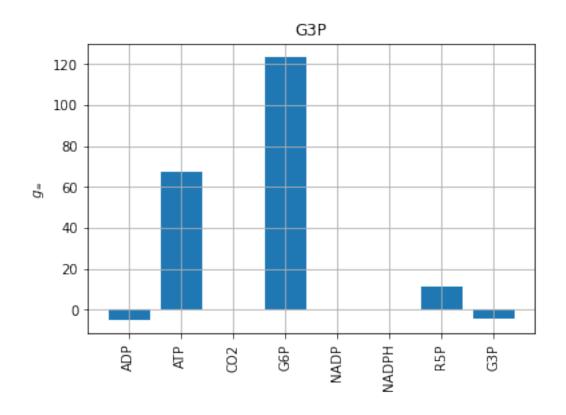


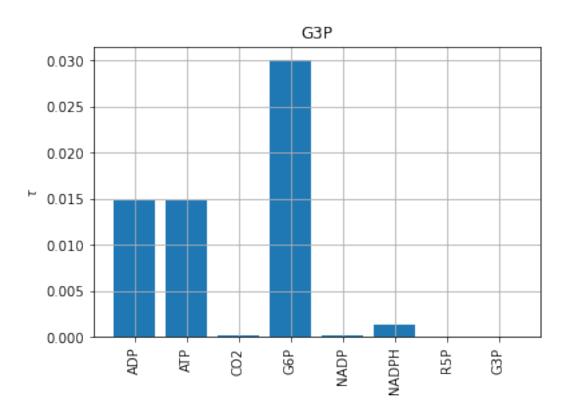
Doing: sG3P

g = -4.574774223072097

dlam = 10





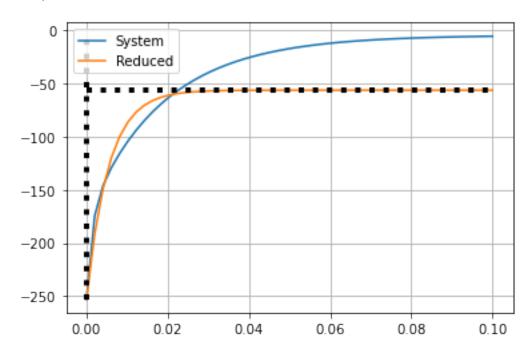


6.3 Test model reduction

```
[31]: ## Test model reduction
      g,tau = tfProps(sys)
      print(f'g = \{g: .2f\}, tau = \{tau: .2f\}')
      sys1 = con.balred(sys,orders=1,method='truncate')
      # print(con.dcgain(sys))
      # print(con.dcgain(sys1))
      con.tf(sys1)
      g,tau = tfProps(sys1)
      print(f'g = \{g:.2f\}, tau = \{tau:.2f\}')
      t = np.linspace(0,0.1)
      step = con.step_response(sys,T=t)
      step1 = con.step_response(sys1,T=t)
      plt.plot(t,step.outputs.T,label='System')
      plt.plot(t,step1.outputs.T,label='Reduced')
      plt.legend()
      plt.grid()
      plotLines()
```

```
g = -4.57, tau = 0.00

g = -56.31, tau = 0.00
```



```
[32]: print(Tau)
```

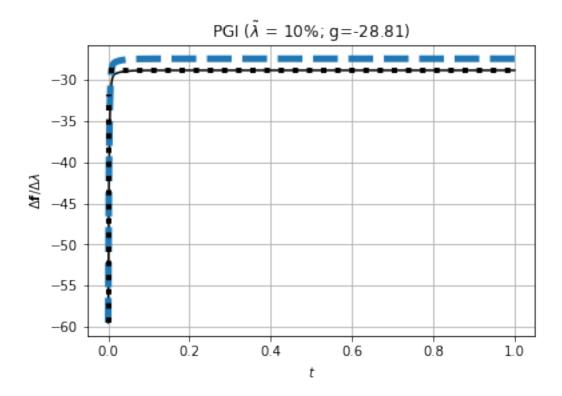
```
{'ADP': 0.014866156765752257, 'ATP': 0.01486615676575225, 'CO2': 0.00017684665624202378, 'G6P': 0.03004139008090118, 'NADP': 0.00017642871957648497, 'NADPH': 0.0013662874711383617, 'R5P': 0, 'G3P': 0}
```

6.4 Compute reaction sensitivities

```
[33]: ## Reaction sensitivities
      Outp = ['G6P', 'R5P', 'NADPH', 'G3P']
      Inp = []
      for reac in s['reaction']:
          Inp.append('s'+reac)
      Inp_reac = Inp
      # Inp = []
      # for reac in s['reaction']:
          Inp.append('s'+reac)
      # Inp = ['sG6PDH2R']
      t_last=1
      lam = 1.1 # Perturbation parameter
      dcgain = {}
      Tau = \{\}
      for outp in Outp:
          for inp in Inp:
              print('Doing:',inp)
              sf = None
              order = None
              dat, y_step, t, sys =__
       ⇒simSensitivity(s,sc,sf,Sys,X_ss,V_ss,dX_ss,chemostats=Chemostats,parameter=parameter,
      →inp=[inp],outp=[outp],lam=lam,t_last=t_last,order=order,tol=None)
                g = con.dcgain(sys)
              g,tau = tfProps(sys)
              print('g =',g, 'tau =',tau)
              poles = con.poles(sys)
                print('Poles:', poles)
                print('Zeros:', con.zeros(sys))
                print(con.ss2tf(sys))
              name = inp[1:]
              if name == 'G6PDH2R':
                  name = 'G6PD.'
              if outp=='G6P':
                  dcgain[name] = -g
                  ylabel = '$-g_{\infty}
              else:
                  dcgain[name] = g
                  ylabel = '$g_{\infty}$'
              Tau[name] = tau
              dlam = int((lam-1)*100)
              print('dlam =',dlam)
              plt.title(f'Order = {poles.size}')
```

```
if Titles:
             plt.title(f'{name}'+r' ($\tilde\lambda$ = ' + f'{dlam}%; g={g:.
 \hookrightarrow 2f)')
         plotSensitivitydX(dat,species=[outp],labeling=False)
           plt.hlines(g,min(t),max(t),color='black',ls='dashed')
         plotLines()
         Savefig('PPPdX_'+name+'_'+outp)
         plt.show()
         #print(con.tf(sys))
     ## Bar chart of gains
    if Titles:
         plt.title(outp)
    plt.tick_params(axis='x', rotation=90)
    plt.grid()
    plt.bar(range(len(dcgain)), dcgain.values(), align='center')
    plt.xticks(range(len(dcgain)), list(dcgain.keys()))
    plt.ylabel(ylabel)
    Savefig(f'PPPdX_{outp}_reac_bar')
    plt.show()
     ## Time constant
    if Titles:
         plt.title(outp)
    plt.tick_params(axis='x', rotation=90)
    plt.grid()
    plt.bar(range(len(Tau)), Tau.values(), align='center')
    plt.xticks(range(len(Tau)), list(Tau.keys()))
    plt.ylabel(r'$\tau$')
    Savefig(f'PPPdX_{outp}_reac_tau_bar')
    plt.show()
Doing: sPGI
```

```
Doing: sPGI
kappa_PGI
g = -28.80991511396469 tau = 0
dlam = 10
G6P
```



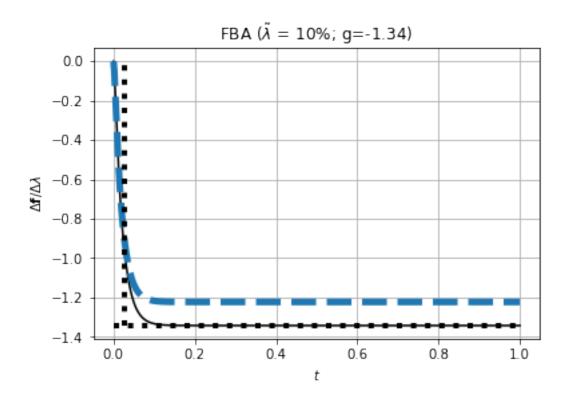
Doing: sPFK kappa_PFK g = -30.913609683373416 tau = 0.002278015825776077 dlam = 10

G6P

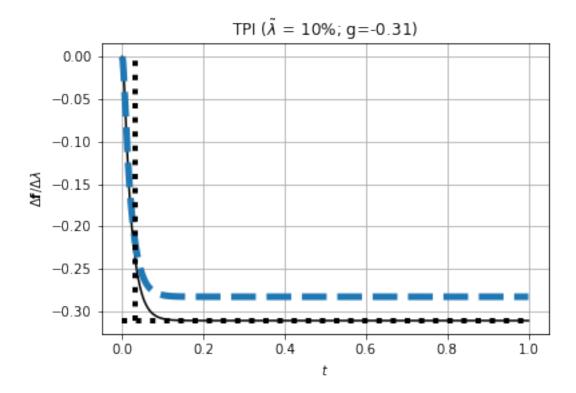
PFK ($\tilde{\lambda} = 10\%$; g=-30.91)

0
-5
-10
-25
-20
-25
-30
0.0
0.2
0.4
0.6
0.8
1.0

Doing: sFBA kappa_FBA g = -1.3427806651016105 tau = 0.026921154867136633 dlam = 10 G6P



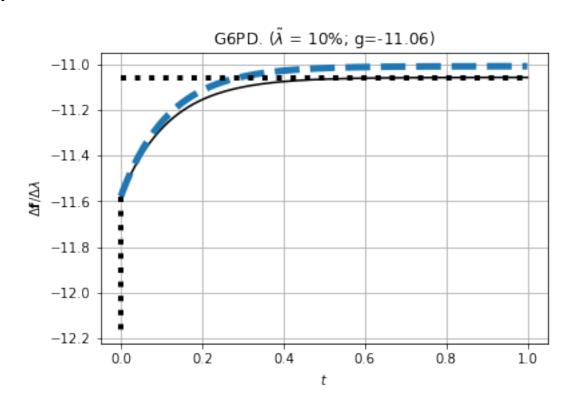
Doing: sTPI
kappa_TPI
g = -0.31124503089893363 tau = 0.03292616976551729
dlam = 10
G6P



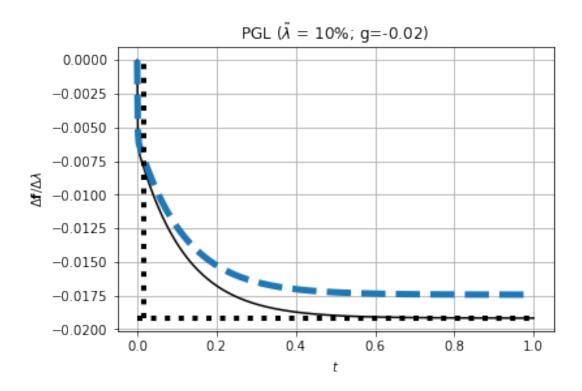
Doing: sG6PDH2R kappa_G6PDH2R

g = -11.05834536093916tau = 0

dlam = 10



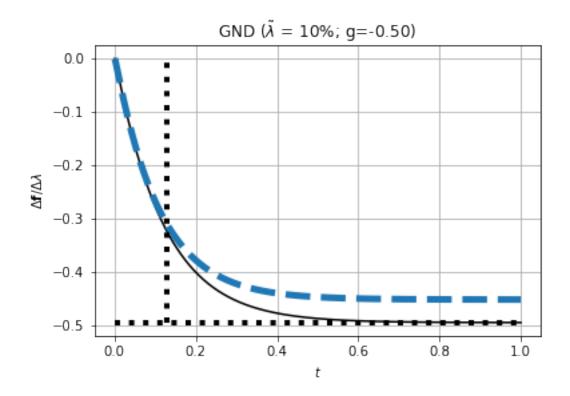
Doing: sPGL
kappa_PGL
g = -0.019168880963538627 tau = 0.016293783108219457
dlam = 10
G6P



Doing: sGND kappa_GND

 $g = -0.4958705310747634 \ \text{tau} = 0.12751609732062125$

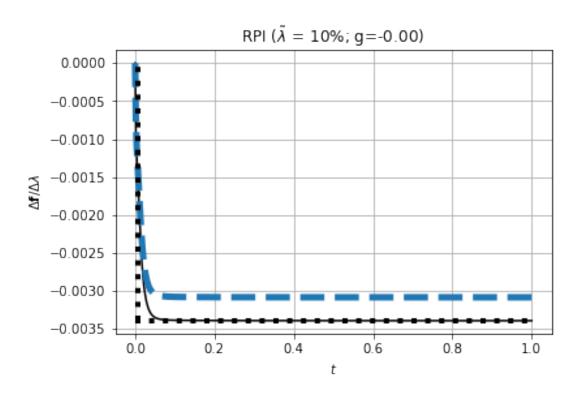
dlam = 10



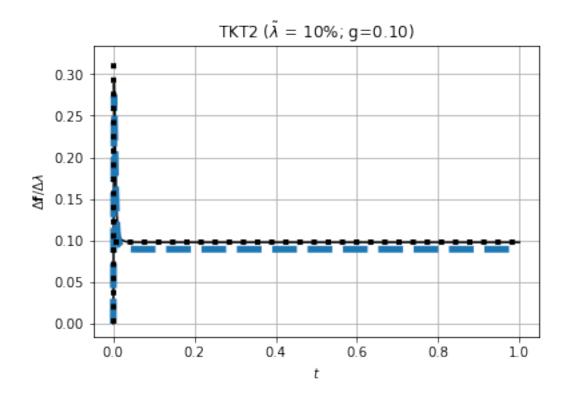
Doing: sRPI kappa_RPI

g = -0.00339246977181743 tau = 0.00824443541241612

dlam = 10



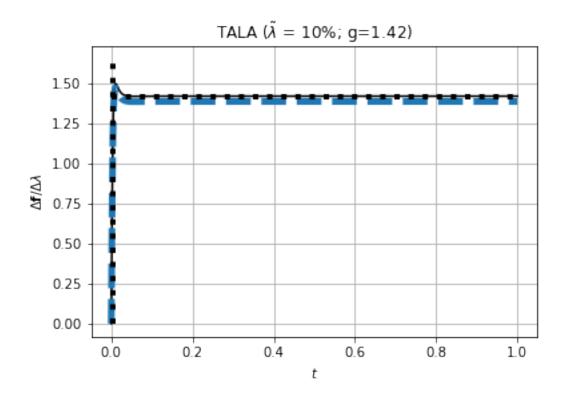
Doing: sTKT2
kappa_TKT2
g = 0.09778496614281532 tau = 0.0009111118563853957
dlam = 10
G6P



Doing: sTALA kappa_TALA

g = 1.4213693479775253 tau = 0.0033987149739030348

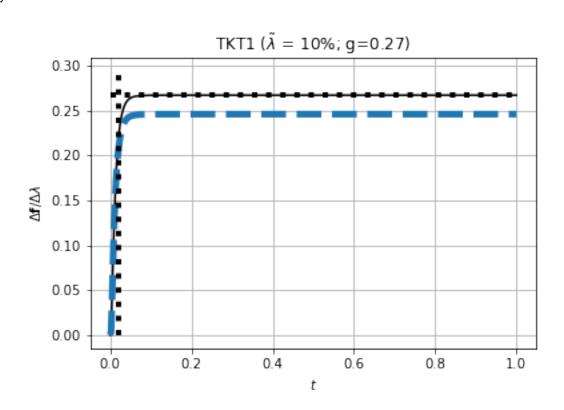
dlam = 10



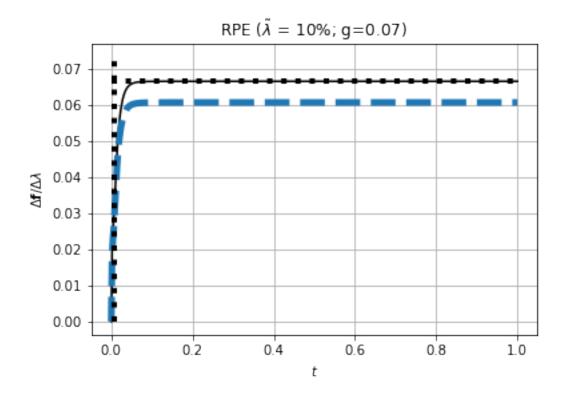
Doing: sTKT1 kappa_TKT1

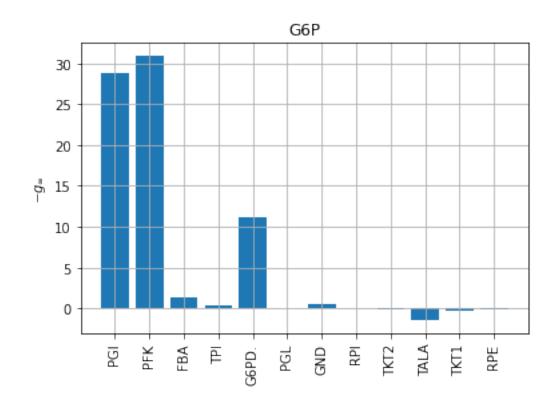
g = 0.2669211003303768 tau = 0.021001174006396915

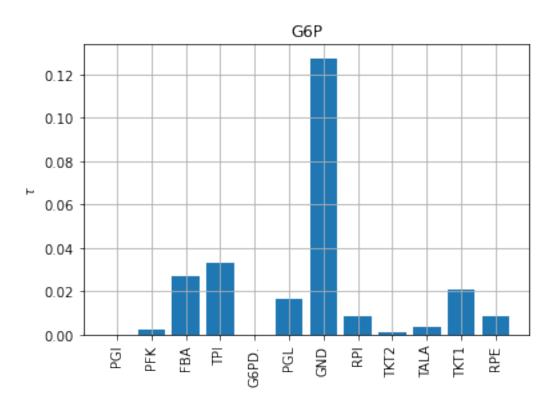
dlam = 10



Doing: sRPE
kappa_RPE
g = 0.06664804356236037 tau = 0.008221010075271595
dlam = 10
G6P



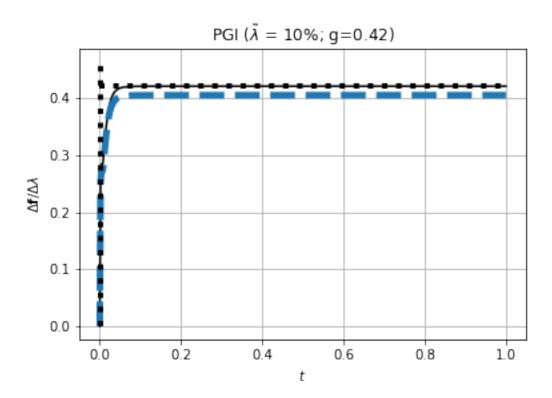




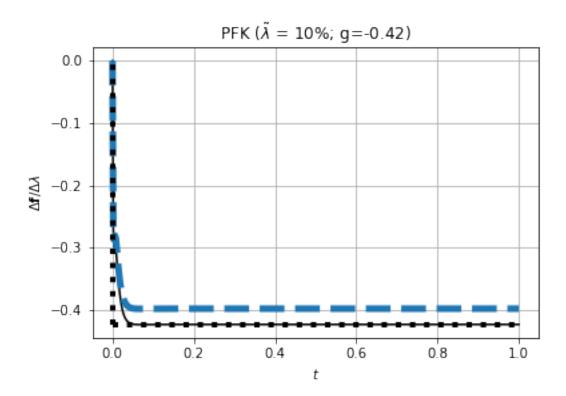
Doing: sPGI kappa_PGI

g = 0.4208987417694914tau = 0.001784019116768216

dlam = 10 R5P



Doing: sPFK kappa_PFK g = -0.42364263563046833 tau = 0.001717964990211834 dlam = 10 R5P

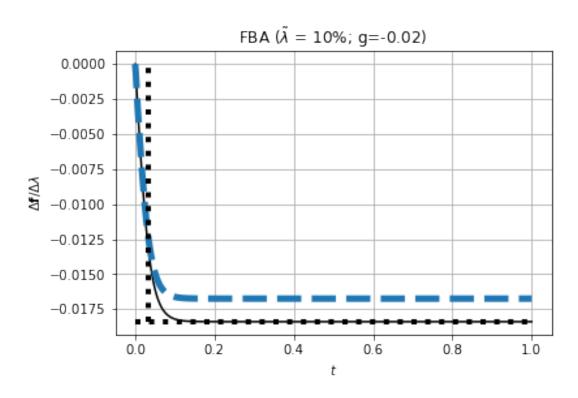


Doing: sFBA kappa_FBA

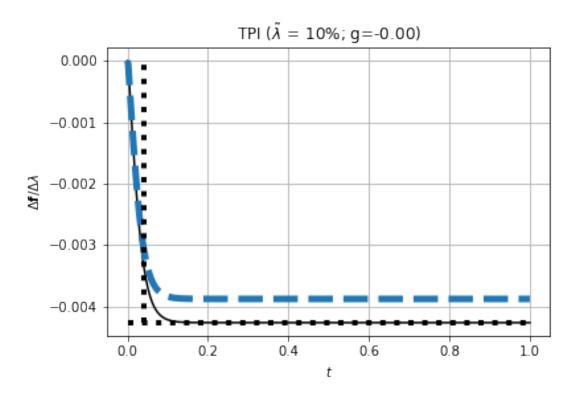
g = -0.018401576065215863 tau = 0.033394801269318826

dlam = 10

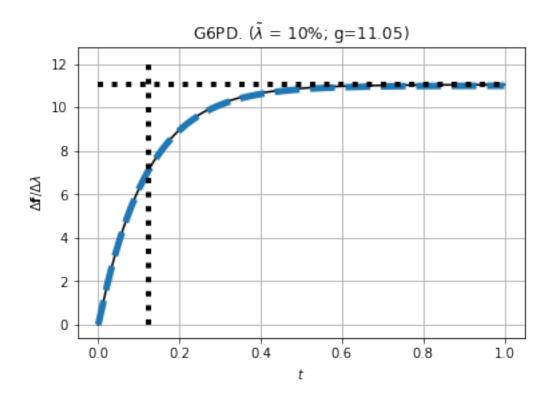
R5P



Doing: sTPI kappa_TPI g = -0.004265327359156856 tau = 0.039469568172793866 dlam = 10 R5P



Doing: sG6PDH2R
kappa_G6PDH2R
g = 11.049474162931 tau = 0.12364579561293719
dlam = 10
R5P

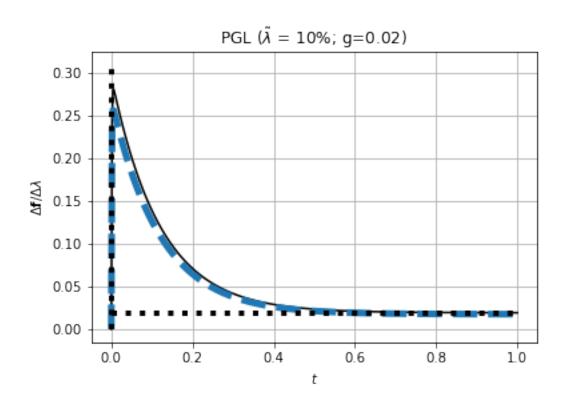


Doing: sPGL kappa_PGL

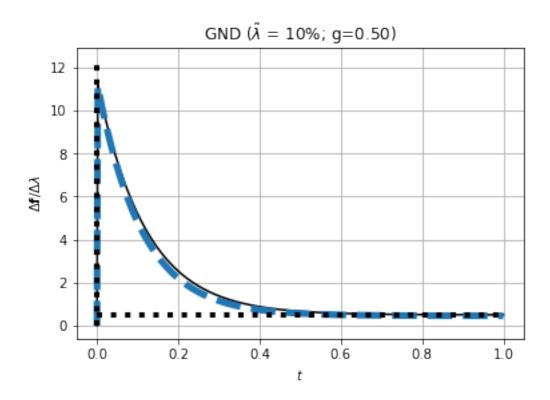
g = 0.01915350335214014 tau = 0.0010470596032330958

dlam = 10

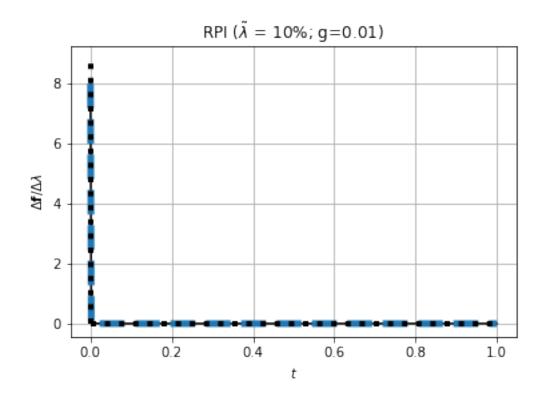
R5P



Doing: sGND kappa_GND g = 0.4954727350638917 tau = 2.6988194236854497e-06 dlam = 10 R5P



Doing: sRPI
kappa_RPI
g = 0.009795231203673538 tau = 0
dlam = 10
R5P

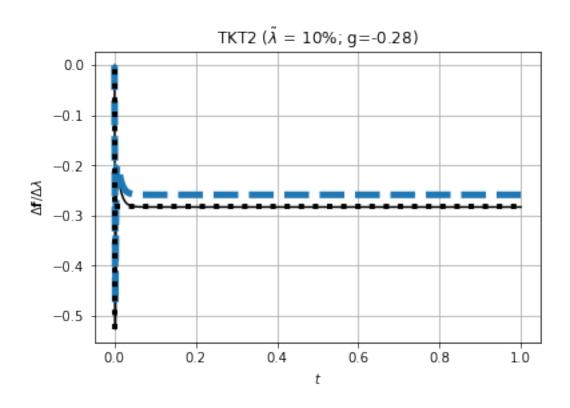


Doing: sTKT2 kappa_TKT2

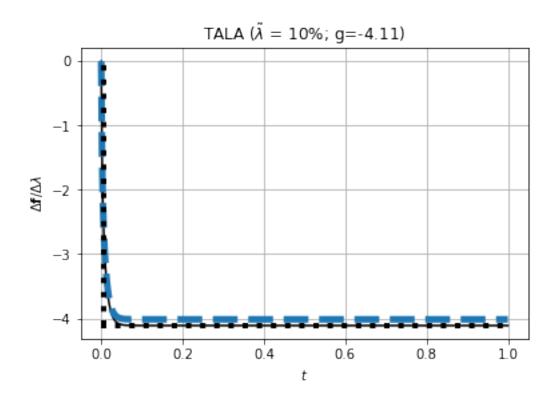
g = -0.282924954852798 tau = 0.00017126550296392315

dlam = 10

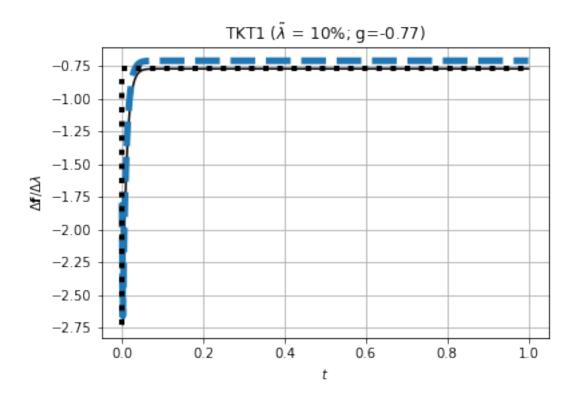
R5P



Doing: sTALA kappa_TALA g = -4.1125018954178 tau = 0.0045160996928807215 dlam = 10 R5P



Doing: sTKT1
kappa_TKT1
g = -0.7722929529876414 tau = 0
dlam = 10
R5P

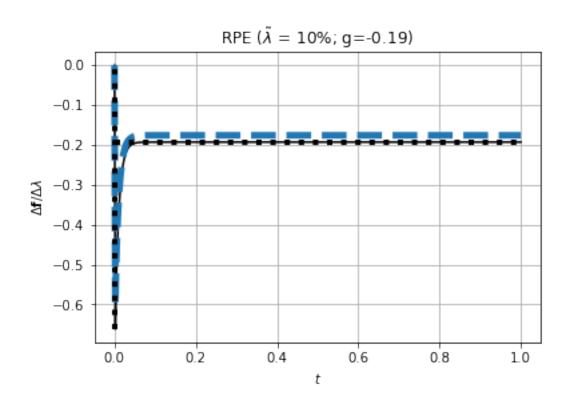


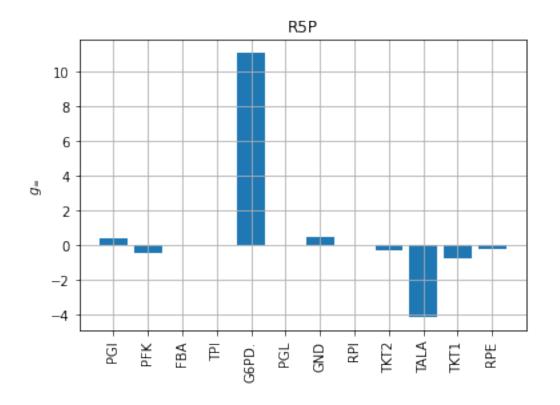
Doing: sRPE kappa_RPE

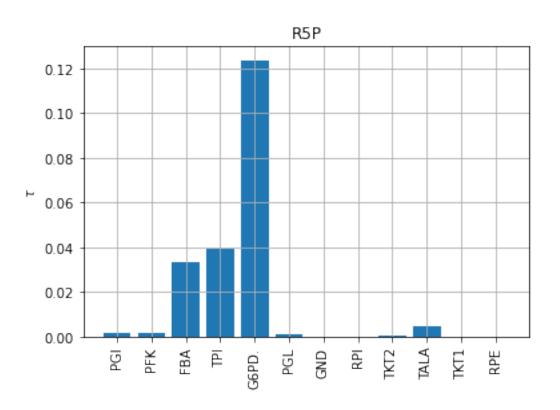
g = -0.19283531466757647 tau = 2.6571744064148927e-06

dlam = 10

R5P



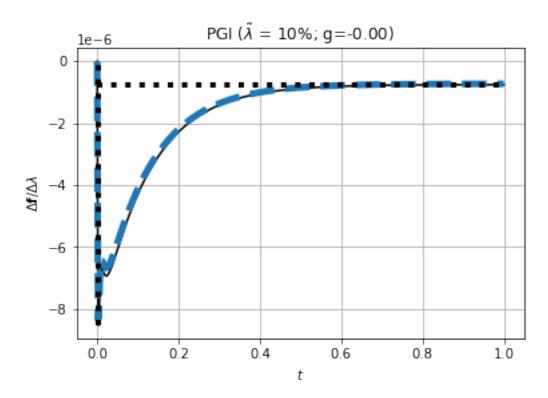




Doing: sPGI kappa_PGI

g = -7.546999477333997e-07 tau = 0.0015271411542531847

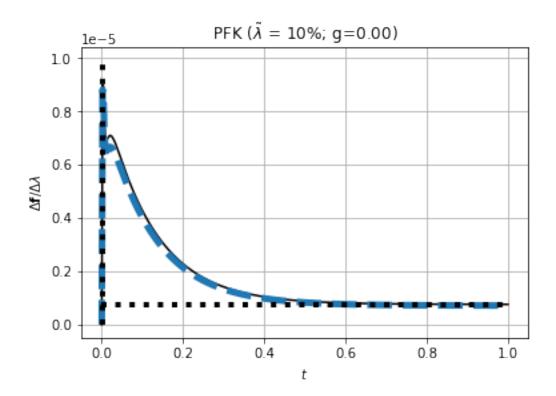
dlam = 10
NADPH



Doing: sPFK kappa_PFK

g = 7.596199715986058e-07 tau = 0.0015261943288164277

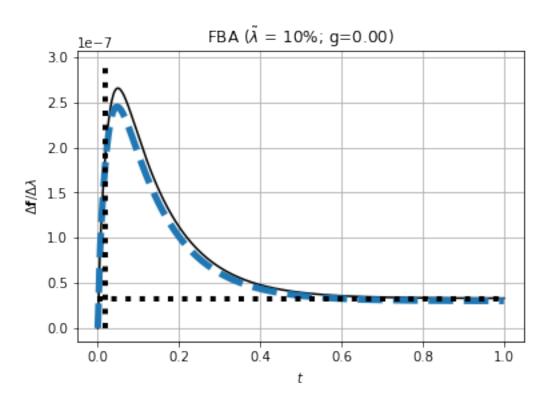
dlam = 10 NADPH



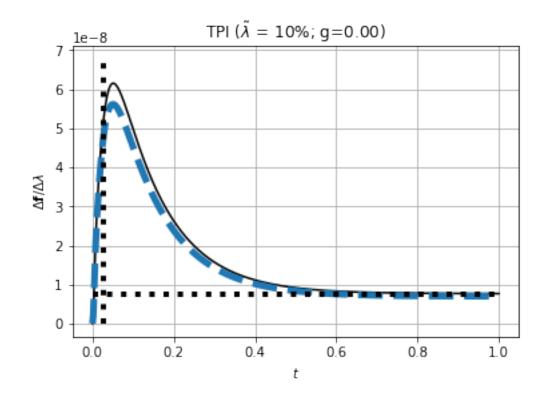
Doing: sFBA kappa_FBA

g = 3.299527434456363e-08 tau = 0.019527507142779787

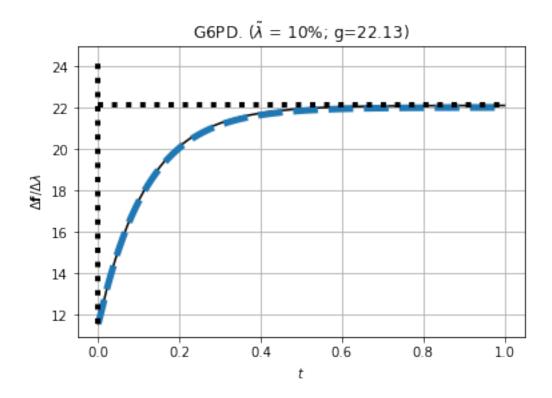
dlam = 10
NADPH



Doing: sTPI kappa_TPI g = 7.647982183374511e-09 tau = 0.02541572981821223 dlam = 10 NADPH



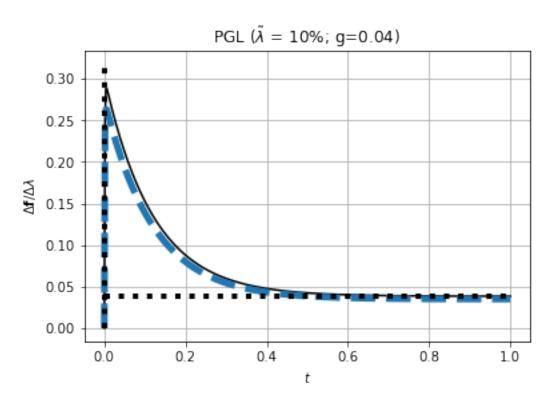
Doing: sG6PDH2R
kappa_G6PDH2R
g = 22.126061707076737 tau = 0
dlam = 10
NADPH



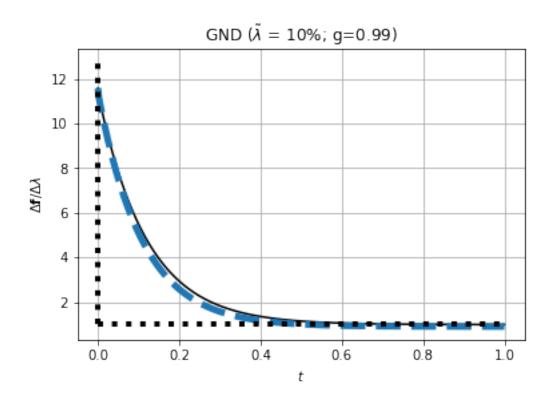
Doing: sPGL kappa_PGL

g = 0.03835400588527368tau = 0.001025533527882495

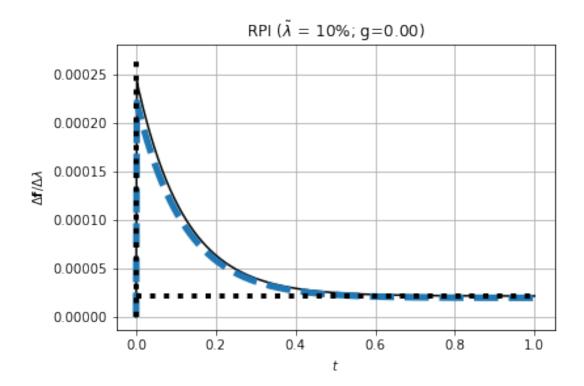
dlam = 10
NADPH



Doing: sGND
kappa_GND
g = 0.9921612692650577 tau = 0
dlam = 10
NADPH



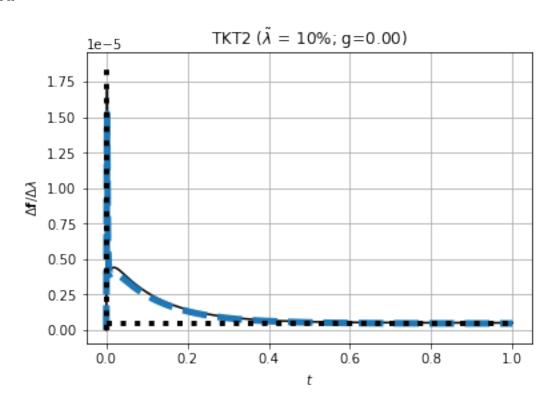
Doing: sRPI
kappa_RPI
g = 2.1458155641420664e-05 tau = 2.699186974424081e-06
dlam = 10
NADPH



Doing: sTKT2 kappa_TKT2

g = 5.073036156607876e-07 tau = 0.00016943022110076314

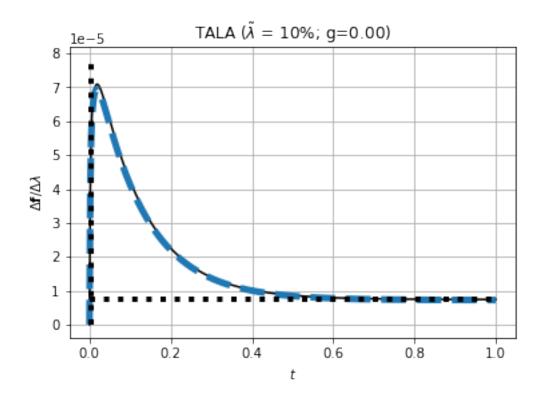
dlam = 10 NADPH



Doing: sTALA kappa_TALA

g = 7.373994570839489e-06 tau = 0.0028552247467929955

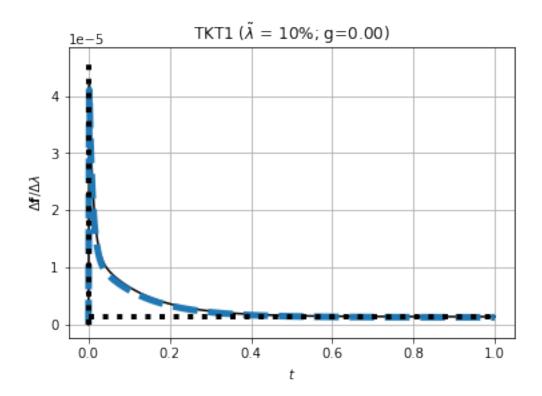
dlam = 10
NADPH



Doing: sTKT1 kappa_TKT1

g = 1.3847735970665945e-06 tau = 0.00018475667332612674

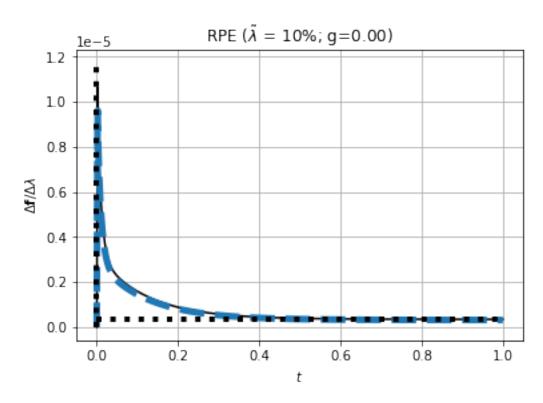
dlam = 10
NADPH

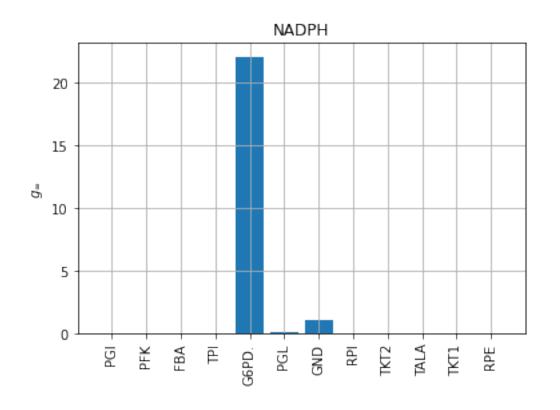


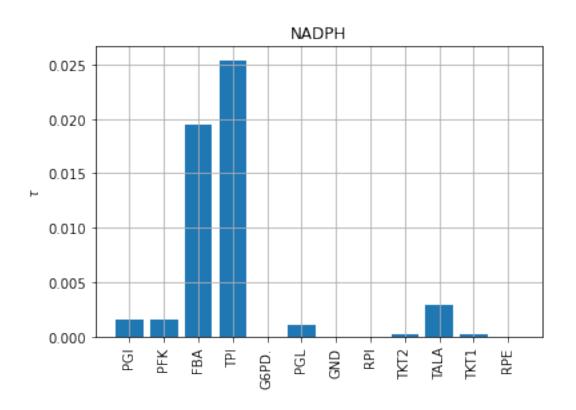
Doing: sRPE kappa_RPE

g = 3.457667861388965e-07 tau = 2.6565945344600007e-06

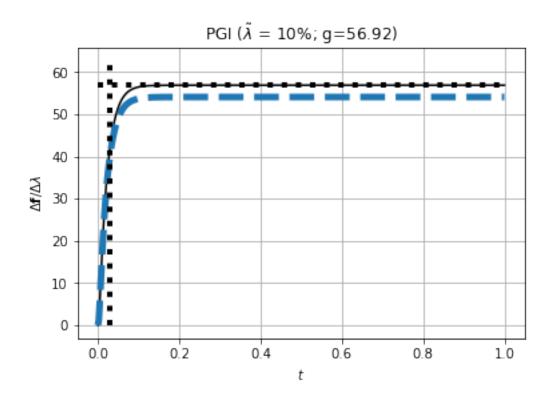
dlam = 10 NADPH



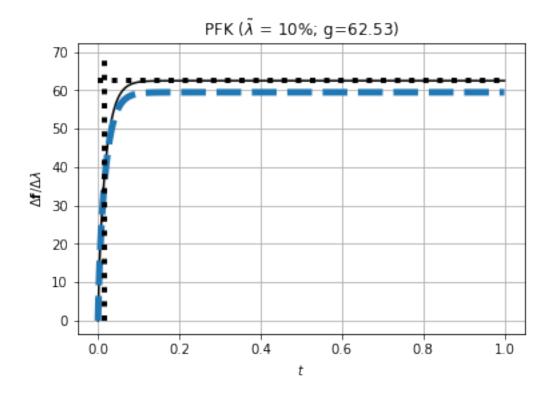




Doing: sPGI
kappa_PGI
g = 56.91833245076208 tau = 0.030041038107083886
dlam = 10
G3P



Doing: sPFK
kappa_PFK
g = 62.533290299526094 tau = 0.014866156765752257
dlam = 10
G3P

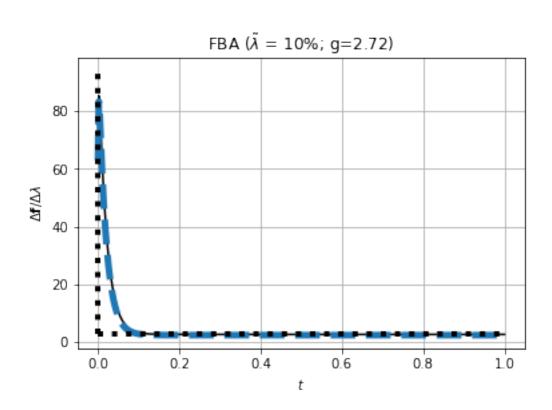


Doing: sFBA kappa_FBA

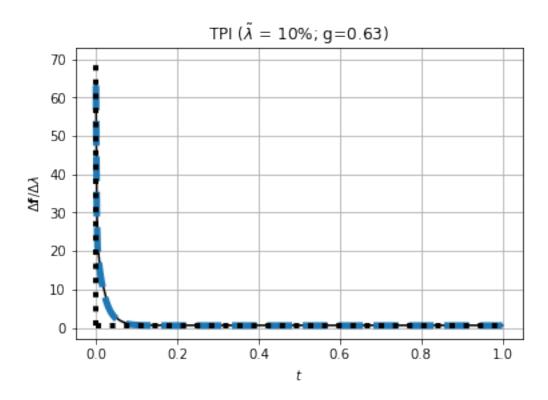
g = 2.7162306181456373 tau = 0

dlam = 10

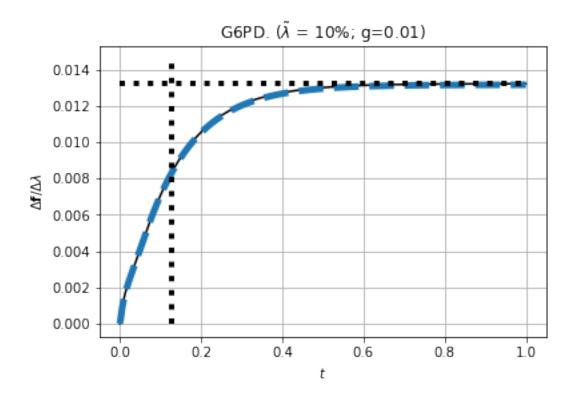
G3P



Doing: sTPI
kappa_TPI
g = 0.6295989394585817 tau = 0
dlam = 10
G3P



Doing: sG6PDH2R
kappa_G6PDH2R
g = 0.01322349914204985 tau = 0.12800864302389323
dlam = 10
G3P

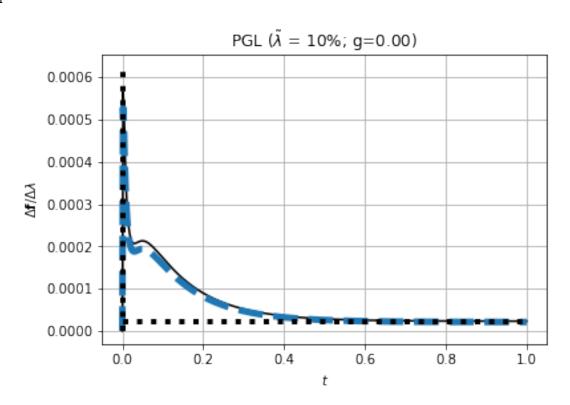


Doing: sPGL kappa_PGL

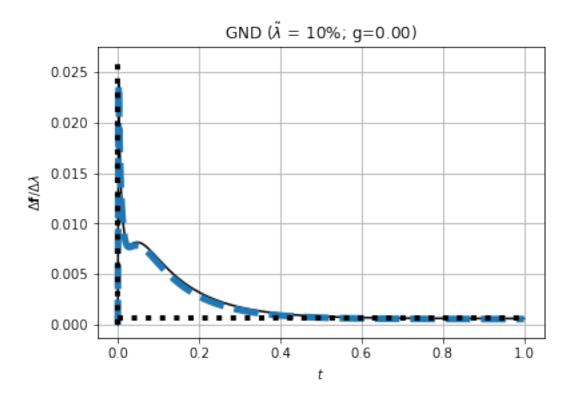
g = 2.2922026197106206e-05 tau = 0.0014045237898896985

dlam = 10

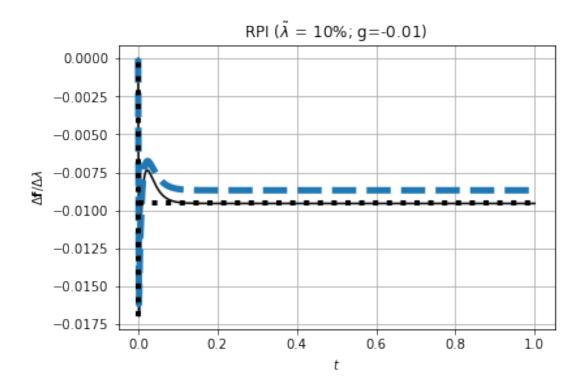
G3P



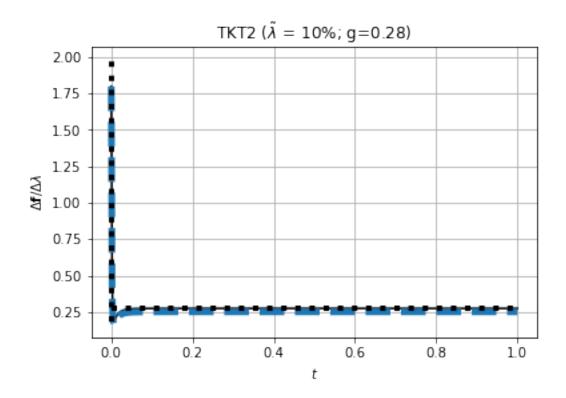
Doing: sGND
kappa_GND
g = 0.0005929588339176166 tau = 0.00017684665624202348
dlam = 10
G3P



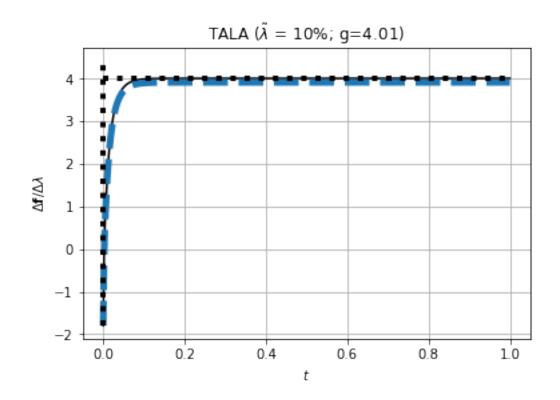
Doing: sRPI
kappa_RPI
g = -0.009544022155095424 tau = 0.00017769464683751047
dlam = 10
G3P



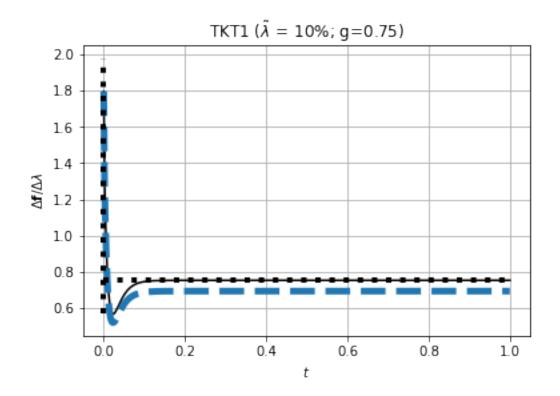
Doing: sTKT2
kappa_TKT2
g = 0.2759715745850926 tau = 0
dlam = 10
G3P



Doing: sTALA
kappa_TALA
g = 4.011429900742118 tau = 0
dlam = 10
G3P



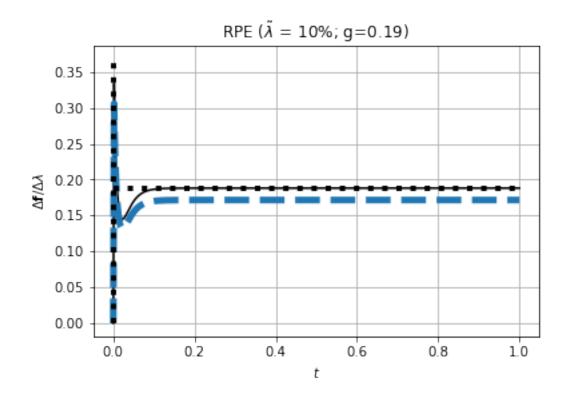
Doing: sTKT1
kappa_TKT1
g = 0.7533124901897006 tau = 0
dlam = 10
G3P

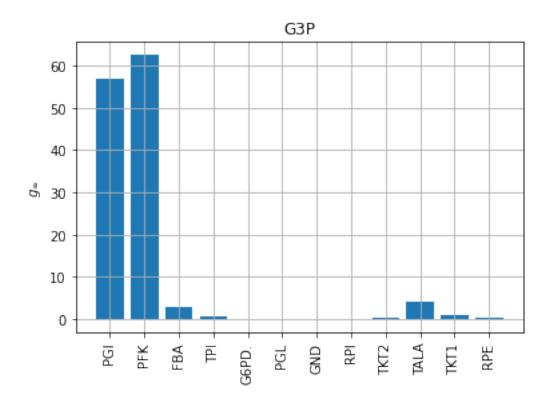


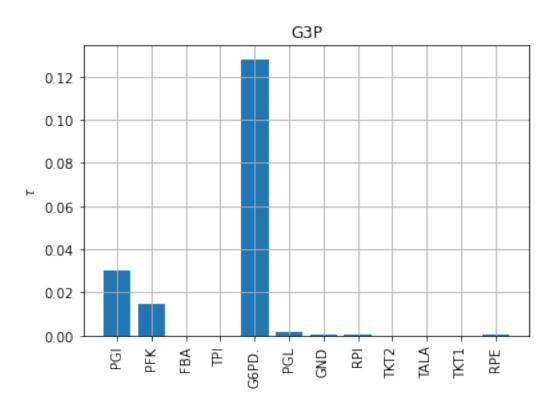
Doing: sRPE kappa_RPE g = 0.1880960

 $g = 0.18809604636010738 \ tau = 0.00016682574714716445$

dlam = 10
G3P







6.5 Vary Lambda

```
[34]:  ## Vary Lambda
      # inp = 'sG6PDH2R'
      # outp = ['R5P']
      Lin = \{\}
      Nlin = \{\}
      Lam = [1.1, 2, 6, 11]
      species = s['species']
      for inp in ['sG6PDH2R','sG6P']:
          for outp in ['R5P','NADPH']:
              lin = []
              nlin = []
              for lam in Lam:
                  dat,y_step,t,sys = simSensitivity(s,sc,sf,Sys,X_ss,V_ss,dX_ss,
       →parameter=parameter,inp=[inp],outp=[outp],lam=lam,t_last=t_last)
                     g = con.dcgain(sys)
                  g,tau = tfProps(sys)
                  g_hf = hfgain(sys)
                  lin.append(g)
                  i = species.index(outp)
                  dX = dat['dX'][-1,i]
                  ng = (dX-dX_ss[i])/(lam-1)
                  nlin.append(ng)
                  name = inp[1:]
                  Name = f'{name}-{outp}'
                  print(f'\{Name\}) (DC gain: \{g:.3f\}, HF gain = \{g_hf:.3f\})
                      plt.title(f'{Name} (DC gain: \{g:.3f\}, HF gain = \{g_hf:.3f\}')
                  if lam>1.5:
                      setZero = outp in ['NADPH']
                      plotSensitivitydX(dat,species=[outp],setZero=setZero)
                      plotLines()
              Lin[Name] = lin
              Nlin[Name] = nlin
              Savefig(f'PPPdX_{name}_{outp}_lambda')
              plt.show()
      # print(Lin)
      # print(Nlin)
```

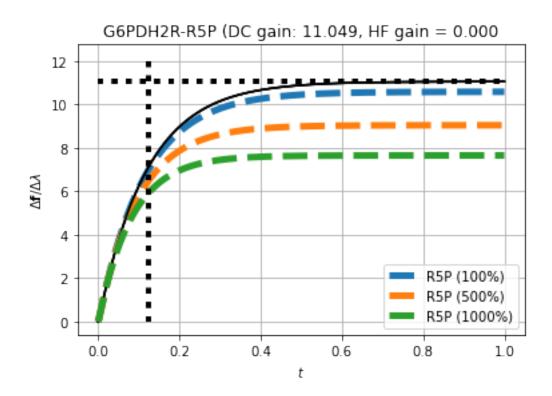
```
kappa_G6PDH2R
G6PDH2R-R5P (DC gain: 11.049, HF gain = 0.000
kappa_G6PDH2R
G6PDH2R-R5P (DC gain: 11.049, HF gain = 0.000
R5P
kappa_G6PDH2R
G6PDH2R-R5P (DC gain: 11.049, HF gain = 0.000
```

R5P

kappa_G6PDH2R

G6PDH2R-R5P (DC gain: 11.049, HF gain = 0.000

R5P



kappa_G6PDH2R

G6PDH2R-NADPH (DC gain: 22.126, HF gain = 11.578

kappa_G6PDH2R

G6PDH2R-NADPH (DC gain: 22.126, HF gain = 11.578

NADPH

 ${\tt kappa_G6PDH2R}$

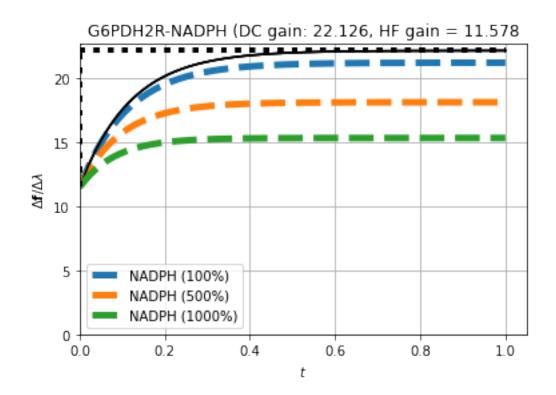
G6PDH2R-NADPH (DC gain: 22.126, HF gain = 11.578

 ${\tt NADPH}$

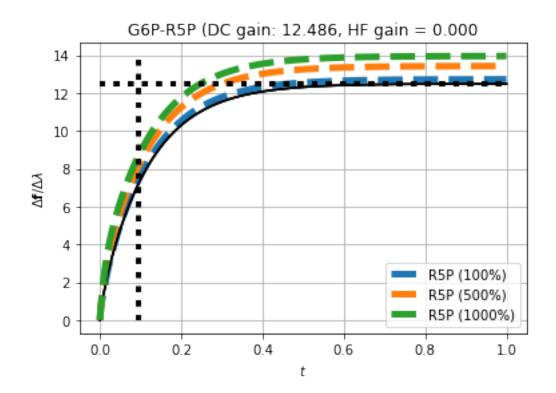
kappa_G6PDH2R

G6PDH2R-NADPH (DC gain: 22.126, HF gain = 11.578

NADPH

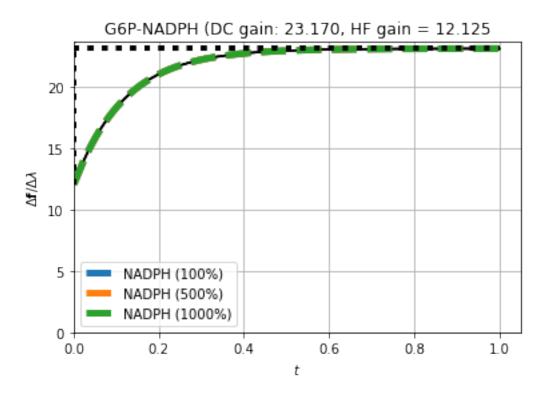


K_G6P
G6P-R5P (DC gain: 12.486, HF gain = 0.000
K_G6P
G6P-R5P (DC gain: 12.486, HF gain = 0.000
R5P
K_G6P
G6P-R5P (DC gain: 12.486, HF gain = 0.000
R5P
K_G6P
G6P-R5P (DC gain: 12.486, HF gain = 0.000
R5P
K_G6P



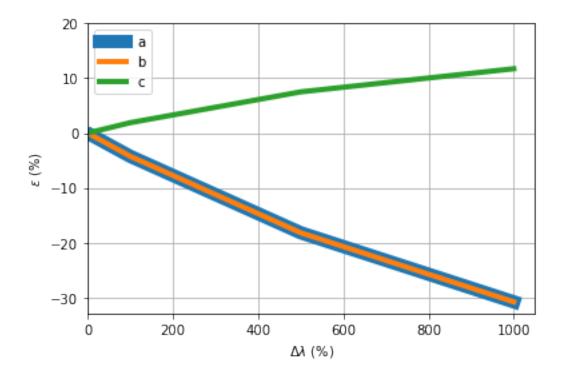
K_G6P
G6P-NADPH (DC gain: 23.170, HF gain = 12.125
K_G6P
G6P-NADPH (DC gain: 23.170, HF gain = 12.125
NADPH
K_G6P
G6P-NADPH (DC gain: 23.170, HF gain = 12.125
NADPH
K_G6P
G6P-NADPH (DC gain: 23.170, HF gain = 12.125

NADPH



```
[35]: ## Plot error curves
      name = ['a','b','c','d']
      lw = [10,4,4,4]
      for i,Name in enumerate(Nlin):
          print(Name)
          G = np.array(Lin[Name])
          NG = np.array(Nlin[Name])
          Err = (NG-G)/G
          ErrPC = 100*Err
          LamPC = (Lam-np.ones(len(Lam)))*100
          ## Just plot the first 3
          if i<3:
              plt.plot(LamPC,ErrPC,label=name[i],lw=lw[i])
      plt.legend()
      plt.grid()
      plt.ylim(top=20)
      plt.xlim(left=0)
      plt.xlabel(r'$\Delta \lambda$ (%)')
      # plt.ylabel(r'$\tilde{\dot{x}}/\tilde{\lambda}-g$ (%)')
      plt.ylabel(r'$\epsilon$ (%)')
      Savefig(f'PPPdX_lambda')
      plt.show()
```

G6PDH2R-R5P G6PDH2R-NADPH G6P-R5P G6P-NADPH



7 Sloppy parameter analysis

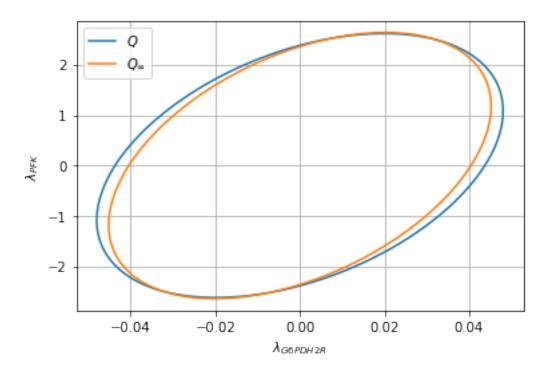
```
[36]: imp.reload(slp)
     def sloppy(Sys,inp,outp,t=None):
          sys = extractSubsystem(Sys,sc,sf,inp,outp)
          #print(sys)
         H,eig,eigv,t = slp.Sloppy(sys,t=t)
         cond = np.linalg.cond(H)
         print(f'Log10 condition number: {np.log10(cond):.1f}')
         slp.SloppyPrint(eig,eigv,inp)
         slp.SloppyPlot(eig,eigv,inp,square=False)
          #slp.SloppyPlotData(t,y,inp,outp)
     def sloppyBoth(Sys,inp,outp,t=None):
          sys = extractSubsystem(Sys,sc,sf,inp,outp)
          #SysName = sc['name']
         blurb = '\n********************
         print(blurb,SysName,blurb)
         if t is None:
             name = f'{SysName}_sloppy_{inp[0]}'
          else:
```

```
name = f'{SysName}_sloppy_{inp[0]}_long'
    #print(sys)
    H,eig,eigv,t = slp.Sloppy(sys,t=t,GainOnly=False)
    print('H:\n',H)
    slp.SloppyPrint(eig,eigv,inp,min_eig=0)
    H,Eig,Eigv,t = slp.Sloppy(sys,t=t,GainOnly=True)
    print('H_ss:\n',H)
    slp.SloppyPrint(Eig,Eigv,inp,min_eig=0)
    slp.SloppyPlot(eig,eigv,inp,Eig=Eig,Eigv=Eigv)
    Savefig(name)
    plt.show()
for t_last in [0,1e1]:
    if t_last==0:
       t = None
    else:
        t = np.linspace(0,t_last,100)
    SysName = 'GlyPPP_PFK_2'
    inp = ['sG6PDH2R','sPFK']
    outp = ['NADPH','R5P']
    sloppyBoth(Sys,inp,outp,t=t)
    SysName = 'GlyPPP_PFK_all'
    Outp = []
    for chemo in sc['chemostats']:
        if not chemo[0] in ['s']:
            Outp.append(chemo)
    print(Outp)
    sloppyBoth(Sys,inp,Outp,t=t)
    SysName = 'GlyPPP_PGI_all'
    inp = ['sG6PDH2R','sPGI']
    sloppyBoth(Sys,inp,Outp,t=t)
```

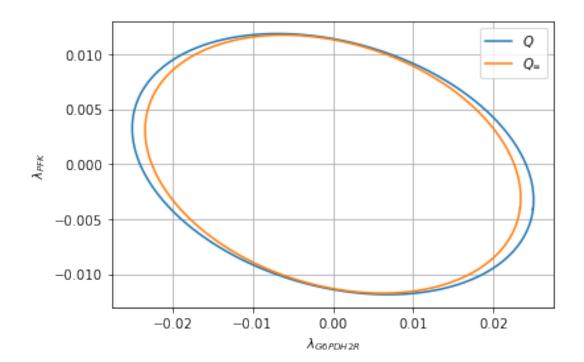
```
************************************

H:
    [[ 5.23558279e+02 -3.97898795e+00]
    [-3.97898795e+00    1.76090966e-01]]
    \sqrt\sigma_1 &= 23 & V_1\Lambda &= + 1.00 \lambda_{G6PDH2R} - 0.01 \lambda_{FFK}
    \sqrt\sigma_2 &= 0.38 & V_2\Lambda &= + 1.00 \lambda_{FFK} + 0.01 \lambda_{G6PDH2R}
H_ss:
```

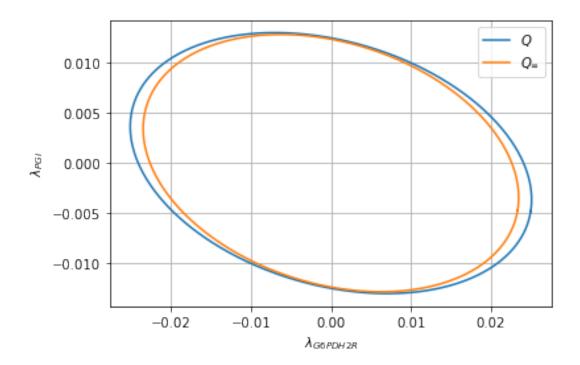
```
[[ 6.12e+02 -4.68e+00]
[-4.68e+00 1.79e-01]]
\sqrt\sigma_1 &= 25 & V_1\Lambda &= + 1.00 \lambda_{G6PDH2R} - 0.01 \lambda_{PFK}
\sqrt\sigma_2 &= 0.38 & V_2\Lambda &= + 1.00 \lambda_{PFK} + 0.01 \lambda_{G6PDH2R}
```



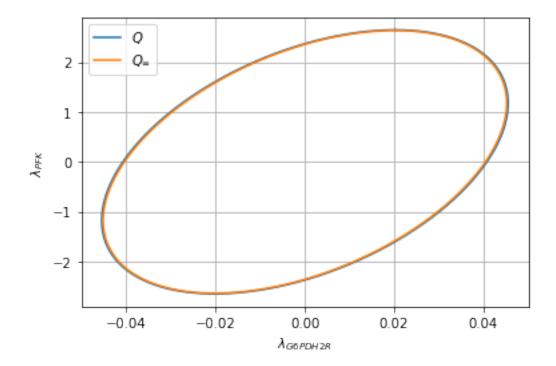
```
['ADP', 'ATP', 'CO2', 'G6P', 'H', 'H2O', 'NADP', 'NADPH', 'R5P', 'G3P']
*********
GlyPPP_PFK_all
**********
Η:
[[1726.16 1001.12]
[1001.12 7677.69]]
\lambda_{G6PDH2R}
\cdotsqrt\sigma_2 \&= 40 \& V_2\Lambda \&= + 0.99 \lambda_{G6PDH2R} - 0.16
\lambda_{PFK}
H_ss:
[[1958.04 1028.65]
[1028.65 7785.82]]
\sqrt 2 = 89 \& V_1\Delta \& = 0.99 \
\lambda_{G6PDH2R}
\sqrt\sigma_2 &= 42 & V_2\Lambda &= + 0.99 \lambda_{G6PDH2R} - 0.17
\lambda_{PFK}
```



```
*********
GlyPPP_PGI_all
*********
H:
[[1726.16 928.88]
[ 928.88 6383.28]]
\sqrt 2 = 81 \& V_1\Delta \& = 0.98 \Lambda_{PGI} + 0.19
\lambda_{G6PDH2R}
\cdots = 39 \& V_2\Lambda \&= + 0.98 \lambda_{G6PDH2R} - 0.19
\lambda_{PGI}
H_ss:
[[1958.04 955.61]
[ 955.61 6511.65]]
\sqrt\sigma_1 &= 82 & V_1\Lambda &= + 0.98 \lambda_{PGI} + 0.20
\lambda_{G6PDH2R}
\cdotsqrt\sigma_2 \&= 42 \& V_2\Lambda \&= + 0.98 \lambda_{G6PDH2R} - 0.20
\lambda_{PGI}
```



```
*********
GlyPPP_PFK_2
**********
Η:
[[ 6.02e+02 -4.60e+00]
[-4.60e+00 1.78e-01]]
\sqrt\sigma_1 &= 25 & V_1\Lambda &= + 1.00 \lambda_{G6PDH2R} - 0.01
\lambda_{PFK}
\sqrt\sigma_2 &= 0.38 & V_2\Lambda &= + 1.00 \lambda_{PFK} + 0.01
\lambda_{G6PDH2R}
H_ss:
[[ 6.12e+02 -4.68e+00]
[-4.68e+00 1.79e-01]]
\sqrt\sigma_1 &= 25 & V_1\Lambda &= + 1.00 \lambda_{G6PDH2R} - 0.01
\lambda_{PFK}
\sqrt\sigma_2 &= 0.38 & V_2\Lambda &= + 1.00 \lambda_{PFK} + 0.01
\lambda_{G6PDH2R}
```



Η:

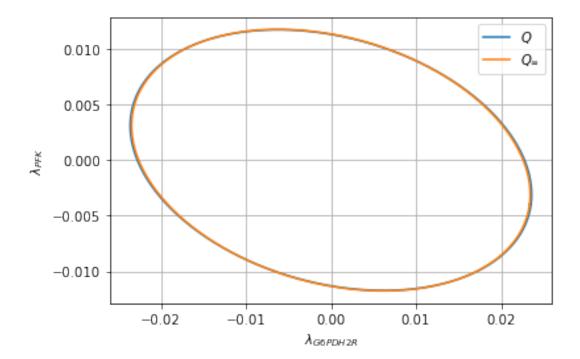
[[1930.92 1024.22] [1024.22 7827.07]]

H_ss:

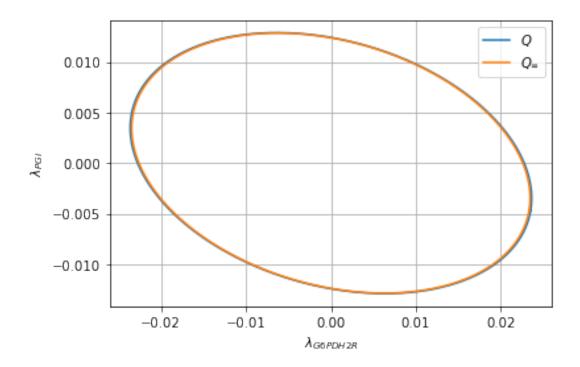
[[1958.04 1028.65]

[1028.65 7785.82]]

 $\cdots = 42 \& V_2\Lambda \&= + 0.99 \lambda_{G6PDH2R} - 0.17 \\ \abel{eq:c6PDH2R}$



```
*********
 GlyPPP_PGI_all
*********
Η:
 [[1930.92 951.55]
 [ 951.55 6481.53]]
\sqrt \frac{1}{\Delta_{0}} \& V_1\Delta \& + 0.98 \Lambda_{PGI} + 0.20
\lambda_{G6PDH2R}
\cdotsqrt\sigma_2 \&= 42 \& V_2\Lambda \&= + 0.98 \lambda_{G6PDH2R} - 0.20
\lambda_{PGI}
H_ss:
 [[1958.04 955.61]
 [ 955.61 6511.65]]
\sqrt\sigma_1 &= 82 & V_1\Lambda &= + 0.98 \lambda_{PGI} + 0.20
\lambda_{G6PDH2R}
\cdotsqrt\sigma_2 \&= 42 \& V_2\Lambda \&= + 0.98 \lambda_{G6PDH2R} - 0.20
\lambda_{PGI}
```



```
[37]: imp.reload(slp)
      Outp = ['R5P','NADPH','G3P']
      # Outp = ['R5P']
      def PrintSloppy(Inp,Outp,GainOnly=True):
          blurb = '\n******\n'
          for outp in Outp:
              print(blurb,outp,blurb)
              sys = extractSubsystem(Sys,sc,sf,Inp,[outp])
                print(con.dcgain(sys))
              gain = con.dcgain(sys)[0]
              norm = np.sum(gain*gain)
              ngain = gain/np.sqrt(norm)
                print(norm, gain/np.sqrt(norm))
              H,eig,eigv,t = slp.Sloppy(sys,GainOnly=GainOnly,small=1e-10)
              slp.SloppyPrint(eig,eigv,Inp,min_eig=0.01,min_eigv=0.1,max_eigs=5)
              if GainOnly:
                  print('Direct')
                  slp.SloppyPrint([norm],np.array([ngain]).T,Inp,min_eig=0.
       \hookrightarrow01,min_eigv=0.1,max_eigs=5)
                print(eigv[:,0])
                print(ngain)
      ## Reactions
      for GainOnly in [True,False]:
          print('\nGainOnly =',GainOnly)
          PrintSloppy(Inp_reac,Outp,GainOnly=GainOnly)
```

GainOnly = True

```
******
R5P
*****
\lambda_{TALA}
Direct
\sqrt\sigma_1 &= 12 & V_1\Lambda &= + 0.93 \lambda_{G6PDH2R} - 0.35
\lambda_{TALA}
*****
NADPH
******
\sqrt\sigma_1 &= 22 & V_1\Lambda &= + 1.00 \lambda_{G6PDH2R}
Direct
\sqrt \frac{1}{\Delta e} 22 \& V_1\Delta \& = + 1.00 \ada_{G6PDH2R}
*****
G3P
*****
\sqrt 2 = 85 \& V_1\Delta \& = 0.74 \
Direct
\sqrt\sigma_1 &= 85 & V_1\Lambda &= + 0.74 \lambda_{PFK} + 0.67 \lambda_{PGI}
GainOnly = False
*****
R5P
*****
\sqrt\sigma_1 &= 11 & V_1\Lambda &= + 0.91 \lambda_{G6PDH2R} - 0.38
\lambda_{TALA} + 0.15 \lambda_{GND}
- 0.21 \lambda_{TALA}
******
NADPH
*****
\sqrt\sigma_1 &= 21 & V_1\Lambda &= + 0.99 \lambda_{G6PDH2R} + 0.11
\lambda_{GND}
\lambda_{G6PDH2R}
******
G3P
```

```
\sqrt\sigma_1 &= 76 & V_1\Lambda &= + 0.74 \lambda_{PFK} + 0.65 \lambda_{PGI} +
    0.16 \lambda_{FBA}
    \sqrt\sigma_2 &= 26 & V_2\Lambda &= + 0.94 \lambda_{FBA} + 0.29 \lambda_{TPI} -
    0.17 \lambda_{PGI}
[38]: ## Chemostats
     for GainOnly in [True,False]:
        print('\nGainOnly =',GainOnly)
        PrintSloppy(Inp_chemo,Outp,GainOnly=GainOnly)
    GainOnly = True
    *****
     R5P
    ******
    0.56 \lambda_{R5P}
    Direct
    \sqrt\sigma_1 &= 21 & V_1\Lambda &= + 0.59 \lambda_{G6P} + 0.57 \lambda_{NADP} -
    0.56 \lambda_{R5P}
    *****
     NADPH
    *****
    \sqrt\sigma_1 &= 34 & V_1\Lambda &= + 0.72 \lambda_{NADP} + 0.69 \lambda_{G6P}
    Direct
    \sqrt 2 = 34 \& V_1\Delta \& = 0.72 \Lambda_{NADP} + 0.69 \Lambda_{G6P}
    ******
     G3P
    *****
    \sqrt \frac{1}{4e+02 \& V_1\Delta \& e} + 0.87 \
    \lambda_{ATP}
    Direct
    \sqrt\sigma_1 &= 1.4e+02 & V_1\Lambda &= + 0.87 \lambda_{G6P} + 0.48
    \lambda_{ATP}
    GainOnly = False
    ******
     R5P
    ******
    \sqrt\sigma_1 &= 1.5e+03 & V_1\Lambda &= + 1.00 \lambda_{R5P}
    *****
     NADPH
```

```
\sqrt\sigma_1 &= 32 & V_1\Lambda &= + 0.74 \lambda_{NADP} + 0.67 \lambda_{G6P}
     ******
      G3P
     *****
     \sqrt\sigma_1 &= 1.3e+02 & V_1\Lambda &= + 0.86 \lambda_{G6P} + 0.48
     \lambda_{ATP} - 0.16 \lambda_{G3P}
     \sqrt\sigma_2 &= 48 & V_2\Lambda &= + 0.98 \lambda_{G3P} + 0.17 \lambda_{G6P}
[39]: ## Three outputs
      Outp = ['R5P','NADPH','G3P']
      sys = extractSubsystem(Sys,sc,sf,Inp,Outp)
      for GainOnly in [True,False]:
          print('\nGainOnly =',GainOnly)
          H,eig,eigv,t = slp.Sloppy(sys,GainOnly=GainOnly,small=1e-10)
          slp.SloppyPrint(eig,eigv,Inp,min_eig=1e-3,min_eigv=0.1,max_eigs=4)
     GainOnly = True
     \sqrt\sigma_1 &= 85 & V_1\Lambda &= + 0.74 \lambda_{PFK} + 0.67 \lambda_{PGI}
     \sqrt\sigma_2 \&= 25 \& V_2\Lambda \&= + 1.00 \lambda_{G6PDH2R}
     \sqrt\sigma_3 &= 3.8 & V_3\Lambda &= + 0.97 \lambda_{TALA} + 0.18 \lambda_{TKT1}
     - 0.14 \lambda_{PGI}
     GainOnly = False
     \sqrt\sigma_1 &= 83 & V_1\Lambda &= + 0.74 \lambda_{PFK} + 0.67 \lambda_{PGI}
     \sqrt\sigma_2 &= 23 & V_2\Lambda &= + 0.99 \lambda_{G6PDH2R} + 0.12
     \lambda_{GND}
     \sqrt Sqrt\sigma_3 \&= 11 \& V_3\Lambda \&= + 0.95 \Lambda_{FBA} + 0.29 \Lambda_{TPI}
     \sqrt\sigma_4 &= 4.3 & V_4\Lambda &= + 0.73 \lambda_{GND} - 0.64 \lambda_{TALA}
     - 0.14 \label{lambda_{G6PDH2R} - 0.14 \lambda_{TKT1}}
[40]: ## All outputs
      for chemo in sc['chemostats']:
          if not chemo[0] in ['s']:
              Outp.append(chemo)
      print(Outp)
      sys = extractSubsystem(Sys,sc,sf,Inp,Outp)
      for GainOnly in [True, False]:
          print('\nGainOnly =',GainOnly)
          H,eig,eigv,t = slp.Sloppy(sys,GainOnly=GainOnly,small=1e-10)
          slp.SloppyPrint(eig,eigv,Inp,min_eig=0.001,min_eigv=0.1,max_eigs=4)
     ['R5P', 'NADPH', 'G3P', 'ADP', 'ATP', 'C02', 'G6P', 'H', 'H2O', 'NADP', 'NADPH',
     'R5P', 'G3P']
     GainOnly = True
     \sqrt\sigma_1 &= 1.5e+02 & V_1\Lambda &= + 0.74 \lambda_{PFK} + 0.67
```

```
\lambda_{PGI}
\sqrt\sigma_2 &= 50 & V_2\Lambda &= + 0.99 \lambda_{G6PDH2R}
\sqrt\sigma_3 &= 6.3 & V_3\Lambda &= + 0.97 \lambda_{TALA} + 0.18 \lambda_{TKT1}
- 0.13 \lambda_{PGI}

GainOnly = False
\sqrt\sigma_1 &= 1.5e+02 & V_1\Lambda &= + 0.74 \lambda_{PFK} + 0.67
\lambda_{PGI}
\sqrt\sigma_2 &= 47 & V_2\Lambda &= + 0.99 \lambda_{G6PDH2R} + 0.11
\lambda_{GND}
\sqrt\sigma_3 &= 16 & V_3\Lambda &= + 0.95 \lambda_{FBA} + 0.29 \lambda_{TPI}
\sqrt\sigma_4 &= 7.4 & V_4\Lambda &= + 0.92 \lambda_{GND} - 0.35 \lambda_{TALA}
- 0.12 \lambda_{G6PDH2R}
```