Supplementary Data

March 15, 2019

1 S.0 Evaluation of computational fluid dynamics and lumpedparameter modelling approaches for planktonic phototrophic growth: Supplementary data

This case file compares the solution of the standard PAM solver as published by Puyol et al, and that of the recently developed CFD solution, namely the combination of the photoBio libraries and pamFoam solver. There will be three different comparisons with the complete Eulerian solution: 1. Fixed radiative field equal to that of incident irradiance (in this case, we have imposed 30 W/m2 @ 850 nm 2. Fixed radiative field at 8.7 W/m2, the radiative field is constant at the value of the half-saturation constant. 2. Dynamic radiative field based on U.water field streamlines in the radiative field at an average fixed

There is an hypothesis that flat plate PBRs can be approximated to CSTRs if one does some appropriate investigation and understands the coupling between the radiative field and the biokinetics. This only applies to the the flat plate reactor

```
In [1]: from IPython.display import HTML
In [2]: HTML('''<script>
        code_show=true;
        function code_toggle() {
         if (code_show){
         $('div.input').hide();
         } else {
         $('div.input').show();
         code_show = !code_show
        $( document ).ready(code_toggle);
        </script>
        The raw code for this IPython notebook is by default hidden for easier reading.
        To toggle on/off the raw code, click <a href="javascript:code_toggle()">here</a>.''')
Out[2]: <IPython.core.display.HTML object>
In [3]: import numpy as np
        import pandas as pd
```

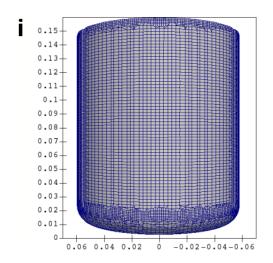
import matplotlib.pyplot as plt

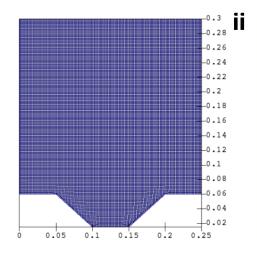
```
from scipy.integrate import odeint
%matplotlib inline
plt.style.use('seaborn-whitegrid')
```

2 S.1 Mesh Generation

Below is are the meshes for both the stirred vessel reactor (CSTR) and the flat plate reactor (FPR). Both meshes were done using cfMesh, and they both consist of 300k cells.

Out[4]:





3 S.2 Initial Conditions for the simulations

| State | Initial Condition | Units | Meaning |
|-------|-------------------|--------|---|
| SS | 0.2 | gCOD/L | soluble organic substrate excluding acetic acid |
| S AC | 0.5 | gCOD/L | acetic acid substrate |
| SI | 0.15 | gCOD/L | inert soluble matter |
| SIN | 0.045 | gN/L | ammonium concentration as nitrogen |
| S IP | 0.00815 | gP/L | phosphorus concentration as phosphate |
| X PB | 0.5 | gCOD/L | PPB biomass concentration as COD |
| ΧS | 0.3 | gCOD/L | biodegradable organic particulates |
| ΧI | 0.2 | gCOD/L | inert particulate matter |

```
In [5]: # Define the PAM solver for the ODE systems
       def monod(I, Ks):
            return I / (Ks + I)
        def pam_batch(y, t, u):
            """ Definition of the system of differential equations to integrate."""
            # Unpack the states
            # Assign variables for convenience of notation
            SS = y[0]
            SAC = y[1]
            SIC = y[2]
           SH2 = y[3]
           SIN = y[4]
           SIP = y[5]
           SI = y[6]
           XPB = y[7]
           XS = y[8]
           XI = y[9]
            \#SCAT = 0.001 + y[2] * 0.3
            # Define parameters of the model (basis days)
            fSSXS = 1.6382408312061000E-01
            fSAXS = 1.166839250294608E-01
            fICXS = 1.3039707398869100E-06
            fH2XS = 8.4424680871970500E-02
            fINXS = 0.011622
            fIPXS = 0.002835
            fSIXS = 0.1518209
            fXIXS = 0.4330922
           fICPHAC = 6.44841269841271e-6
                                            # molHCD3-C/mgCOD
            fICPHSS = -1.242761443579e-6
                                             # molHCO3-C/mgCOD
                                 # mqCOD/mqCOD
            fACCH = 0.6691
            fH2CH = 1.0 - fACCH # mgCOD/mgCOD
                                 # molHCO3-C/molHCO3-C
           fICAU = 1.0
           fH2AU = 40320.0
                               # mgCOD/molHCO3-C
           fNB = 0.086
                                 # mgNH3-N/mgCOD
            fPB = 0.015
                                # mgPO4-P/mgCOD
                               # d-1
           kHYD = 7.09e-2
           kDEC = 9.00e-2
                               # d-1
           kMAC = 2.375
                                # d-1
```

```
# d-1
kMPH = 1.435
kMCH = 7.386e-2
                   # d-1
kMIC = 6.0e-6
                   # mol HCO3-C/mqCOD/d
KSS = 0.524235146324262
                          # mg COD/L
KSAC = 20.222562
                          # mg COD/L
                        # mol HCO3-C/L
KSIC = 4.2e-4
KSIN = 0.02
                        # mgNH3-N/L
KSIP = 0.081
                        # mgPO4-P/L
                        # mg NH4-N/L
KIFA = 7850.0
KSH2 = 1.0
                          # mq COD/L
                          # mgCOD/mgCOD
YPBPH = 1.0
YPBCH = 0.680705638006362 \# mqCOD/mqCOD
YPBAU = 40320.00
                          # mgCOD/molHCO3-C
KSE = 8.76
                          # W/m2
SE = u
                          # W/m2
fICDEC = -1.98412698412703E-07 # mol C-HCO3 / mgCOD
fINDEC = 0.058
                              # mgNH3-N / mgCOD
                              # mgPO4-P / mgCOD
fIPDEC = 0.01
# Constitutive equations
rHYD = kHYD * XS
rDEC = kDEC * XPB
IIN = SIN/(KSIN + SIN)
IIP = SIP/(KSIP + SIP)
IFA = KIFA/(KIFA + SIN)
IE = monod(SE, 8.76)
ICS = SAC / (SS + SAC) # ACT inhibition due to SS
ICAC = SS / (SS + SAC) # PHT inhibition due to SAC
rACT = kMAC * XPB * IFA * IIN * IIP * IE * (SAC/(KSAC + SAC)) * ICS
rPHT = kMPH * XPB * IFA * IIN * IIP * IE * (SS/(KSAC + SAC)) * ICAC
rCHE = kMCH * XPB * IFA * IIN * IIP * SS/(KSS + SS)
rAUT = kMIC * XPB * IFA * IIN * IIP * IE * SIC/(SIC + KSIC)*SH2/(SH2 + KSH2);
n = len(y)
               # 10: number of states
dydt = np.empty((n))
dydt[0] = fSSXS * rHYD \
       - rPHT \
       - rCHE
dydt[1] = fSAXS * rHYD \
```

```
- rACT \
                   + (1 - YPBCH)*fACCH*rCHE
           dydt[2] = fICXS * rHYD \
                  + fICPHAC * rACT \
                  + fICPHSS * rPHT \
                   - fICAU * rAUT \
                   + fICDEC * rDEC
           dydt[3] = fH2XS * rHYD \setminus
                   + (1 - YPBCH) * fH2CH * rCHE \
                  - fH2AU * rAUT
           dydt[4] = fINXS*rHYD \
                   - fNB*YPBPH*rACT \
                   - fNB*YPBPH*rPHT \
                  + fINDEC*rDEC \
                   - fNB*YPBCH*rCHE \
                   - fNB*YPBAU*rAUT
           dydt[5] = fIPXS*rHYD \
                  - fPB*YPBPH*rACT \
                   - fPB*YPBPH*rPHT \
                  + fIPDEC*rDEC \
                   - fPB*YPBCH*rCHE \
                   - fPB*YPBAU*rAUT
           dydt[6] = fSIXS*rHYD
           dydt[7] = YPBPH*rACT \
                  + YPBPH*rPHT \
                   - rDEC \
                  + YPBCH*rCHE \
                   + YPBAU*rAUT
           dydt[8] = rDEC - rHYD
           dydt[9] = fXIXS * rHYD
           return dydt
In [6]: # 30 W/m2 uniform irradiance
       # USER INPUT
       num_steps = 800
```

```
days = 120604/3600/24
SS0 = 200.0
SACO = 500.0
SICO = 7.0E-6
SH20 = 0.0
SINO = 45 # 45
SIP0 = 8.15
            # 8.15
SIO = 150.0
XPBO = 500.0
XS0 = 300.0
XIO = 100.0
y0 = [SSO, SACO, SICO, SH2O, SINO, SIPO, SIO, XPBO, XSO, XIO,]
time_ode = np.linspace(0,days,num_steps)
#u_t = interp1d(time_input.squeeze(), full_G850.squeeze())
# Store our state outputs here
y_{out} = [y0]
# Loop over each point in the dataset
for step in range(num_steps - 1):
   t = [time_ode[step], time_ode[step+1]]
    # Get G850 for this step
    G850 = 30
    # Solve the ODEs
    soln = odeint(pam_batch, y0, t, args=(G850,))
   y_inter = np.float64(soln[1])
   y_out.append(y_inter)
   y0 = y_inter.tolist()
y_out = np.asarray(y_out)
df_30 = pd.DataFrame(y_out)
df_30['SSout'] = y_out[:,0]
df_30['SACout'] = y_out[:,1]
df_30['SICout'] = y_out[:,2]
df_30['SH2out'] = y_out[:,3]
df_30['SINout'] = y_out[:,4]
df_30['SIPout'] = y_out[:,5]
```

```
df_30['SIout'] = y_out[:,6]
       df_30['XPBout'] = y_out[:,7]
       df_30['XSout'] = y_out[:,8]
       df_30['XIout'] = y_out[:,9]
       df_30['time'] = time_ode
In [7]: # Beer Lambert irradiance (ODE2)
       # USER INPUT
       num_steps = 800
       days = 120604/3600/24
       SS0 = 200.0
       SACO = 500.0
       SICO = 7.0E-6
       SH20 = 0.0
       SINO = 45 # 45
       SIPO = 8.15 # 8.15
       SIO = 150.0
       XPBO = 500.0
       XSO = 300.0
       XIO = 100.0
       def beer lambert(GO, AplusS, L):
              return G0*np.exp(-AplusS*L)
       G850 = beer_lambert(30, 53+15, 0.0075)
       y0 = [SSO, SACO, SICO, SH2O, SINO, SIPO, SIO, XPBO, XSO, XIO,]
       time_ode = np.linspace(0,days,num_steps)
       #u_t = interp1d(time_input.squeeze(), full_G850.squeeze())
       # Store our state outputs here
       y_out = [y0]
       # Loop over each point in the dataset
       for step in range(num_steps - 1):
          t = [time_ode[step], time_ode[step+1]]
          # Get G850 for this step
          G850 = beer_lambert(30, 53, 0.015)
          # Solve the ODEs
          soln = odeint(pam_batch, y0, t, args=(G850,))
```

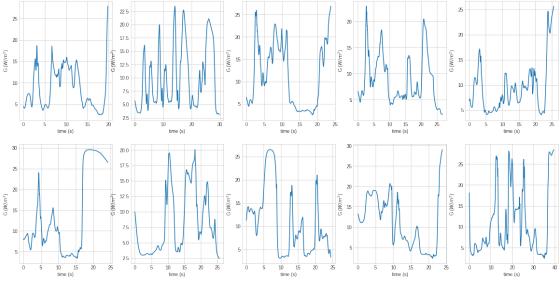
```
y_inter = np.float64(soln[1])
    y_out.append(y_inter)
    y0 = y_inter.tolist()
y_out = np.asarray(y_out)
df_10 = pd.DataFrame(y_out)
df_10['SSout'] = y_out[:,0]
df_10['SACout'] = y_out[:,1]
df_10['SICout'] = y_out[:,2]
df_10['SH2out'] = y_out[:,3]
df_10['SINout'] = y_out[:,4]
df_10['SIPout'] = y_out[:,5]
df_10['SIout'] = y_out[:,6]
df_10['XPBout'] = y_out[:,7]
df_10['XSout'] = y_out[:,8]
df_10['XIout'] = y_out[:,9]
df_10['time'] = time_ode
```

4 S.3 Solve transient radiative field problem using streamlines on U.water

We will first do the process for the flat plate multiphase bioreactor, then we can look at the stirred vessel.

```
In [8]: stream_flat = pd.read_csv("./data/streamlines1.csv")["G850"]
        streamtime_flat = pd.read_csv("./data/streamlines1.csv")["IntegrationTime"]
        # Check where time goes back to zero,
        # If the size of the resulting array is greater than a certain value, M,
        # then return that to a list of indices
        # This is for the flat plate reactor
        lst_flat = []
        M = 500
        j = 0
        starter = 30
        j = starter
        while starter < len(streamtime_flat)-2:</pre>
            while streamtime_flat[j] < streamtime_flat[j+1]:</pre>
                j = j + 1
            if j - starter > M:
                lst_flat.append([starter, j])
```

As we can see from the results above, there are 10 intervals which give us more than 500 datapoints each. These intervals will be used as a basis of potential candidates for the dynamic particles. Below, we can see the graphic of the streamlines as they experience the radiative field at 850 nm.

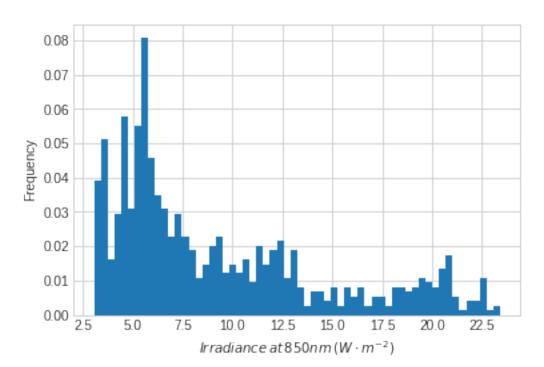


4.1 Determination of the probability law of the streamlines

We can see in the following graphic the probability distribution and the important parameters of the flat plate G distribution. The rate parameter, λ is 0.16, and the minimum value of G over the selected stream is 3.05 W/m2.

```
In [10]: # Determination of Law
         import scipy.stats as stats
         part0 = stream_flat[lst_flat[1][0]:lst_flat[1][1]]
         time0 = streamtime flat[lst flat[1][0]:lst flat[1][1]]
         meanG = np.log(stream_flat[lst_flat[1][0]:lst_flat[1][1]]).mean()
         stdG = np.log(part0).std()
         maxG = part0.max()
         particle_flat = part0
         minG_flat = particle_flat.min()
         time_flat = time0
         loc_flat,scale_flat = stats.expon.fit(particle_flat, floc=minG_flat)
         x_flat = np.linspace(0,30)
         dist_flat = stats.expon.pdf(x_flat, loc_flat, scale_flat)
         weights_flat = np.ones_like(particle_flat)/float(len(particle_flat))
         plt.hist(particle flat, 60, weights=weights flat)
         plt.xlabel('${Irradiance\, at\, 850 nm\, (W\cdot m^{-2}})$')
         plt.ylabel('Frequency')
         plt.show()
         print(loc_flat, scale_flat)
         print(f'rate = {1/scale_flat}')
         print(f'minG = {minG_flat}')
         time0.index = pd.RangeIndex(len(time0.index))
         part1 = np.concatenate(60*24*3*[part0])
         time1 = np.concatenate(60*24*3*[time0])
         # Create a deltaTime list
         deltaTime = []
         for i in range(len(time0)-1):
             deltaTime.append(time0[i+1] - time0[i])
         time1 = [0]
         # Turn deltaTime into a np.array
         deltaTime = np.concatenate((60*24*3+4)*[np.array(deltaTime)])
         for i in range(len(deltaTime)):
             time1.append(time1[-1]+deltaTime[i])
         from scipy.interpolate import interp1d
```

```
time2 = np.asarray(time1)/86400
f = interp1d(time2, part1[0:len(time1)])
Nsteps = 800
```



```
3.0435 6.172651881720432
rate = 0.1620049241657998
minG = 3.0435
```

In [11]: # ODE3 (dynamic input)

num_steps = 10000

days = 172800/3600/24

SS0 = 200.0

SACO = 500.0

SICO = 7.0E-6

SH20 = 0.0

SINO = 45 # 45

SIPO = 8.15 # 8.15

SIO = 150.0

XPB0 = 500.0

```
XSO = 300.0
XIO = 100.0
y0 = [SS0, SAC0, SIC0, SH20, SIN0, SIP0, SI0, XPB0, XS0, XI0,]
# Interpolate the irradiance array
t_new = np.linspace(0, time2.max(), num_steps)
G850 = f(t_new)
# Store our state outputs here
y_out = [y0]
# Loop over each point in the dataset
for step in range(num_steps-1):
   t = [t_new[step], t_new[step+1]]
    # Get G850 for this step
   u = G850[step:step+1]
    # Solve the ODEs
   soln = odeint(pam_batch, y0, t, args=(u,))
   y_inter = np.float64(soln[1])
   y_out.append(y_inter)
   y0 = y_inter.tolist()
y_out = np.asarray(y_out)
df_dyn = pd.DataFrame((y_out))
df_dyn['SSout'] = y_out[:,0]
df_dyn['SACout'] = y_out[:,1]
df_dyn['SICout'] = y_out[:,2]
df_dyn['SH2out'] = y_out[:,3]
df_dyn['SINout'] = y_out[:,4]
df_dyn['SIPout'] = y_out[:,5]
df_dyn['SIout'] = y_out[:,6]
df_dyn['XPBout'] = y_out[:,7]
df_dyn['XSout'] = y_out[:,8]
df_dyn['XIout'] = y_out[:,9]
df_dyn['time'] = t_new
```

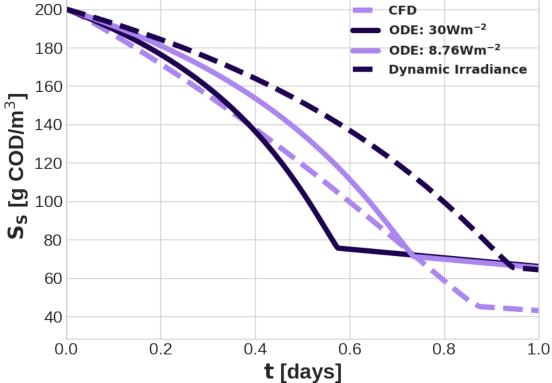
5 S.4 Importing of CFD Data

Here we impport the CFD results from the FPR and CR. Once the simulations are run on the ODE system, one must import the results from the no-flow solution from the OpenFOAM solver. In this case, the irradiance was assumed as equal throughout the entire domain at $30 \, \mathrm{Wm}^{-2}$.

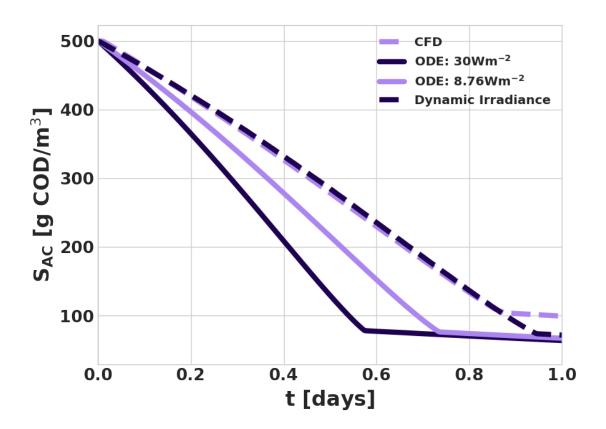
N.B. the medium and coarse solutions overlapped, with the generated graphics being available in the supplementary information directory.

```
In [13]: fig1 = plt.figure(figsize=(8,6), dpi=120)
         ax1 = fig1.add_subplot(111)
         ax1.tick_params(axis="both",
                        reset=False,
                        labelsize=16)
         font = {'family' : 'DejaVu Sans',
                 'weight' : 'bold',
                 'size' : 14}
         plt.rc('font', **font)
         plt.plot(data["time"]/86400,
                  data["SS"]*1000,
                  "--",
                  label='CFD',
                  linewidth=5,
                  color="#ac86ef")
         plt.plot(df_30["time"],
                  df_30["SSout"],
                  label='ODE: 30$\mathbf{Wm^{-2}}$',
                  linewidth=5,
                  color="#1d004f")
```

```
plt.plot(df_10["time"],
         df_10["SSout"],
         label='ODE: 8.76\\mathbf{\m^{-2}}\\,
         linewidth=5,
         color="#ac86ef")
plt.plot(df_dyn["time"],
         df_dyn["SSout"],
         "--",
         label='Dynamic Irradiance',
         linewidth=5,
         color="#1d004f"
         )
plt.legend(fontsize=12)
plt.xlabel(r'$\mathbf{t}$ [days]', fontweight='bold', fontsize=20)
plt.ylabel(r'$\mathbf{S_S}$ [g COD/m$^3$]', fontweight='bold', fontsize=20)
plt.xlim([0, 1])
fig1.savefig("graphs/SS_evol.eps")
200
                                                 CFD
                                                 ODE: 30Wm<sup>-2</sup>
180
                                                 ODE: 8.76Wm<sup>-2</sup>
                                                 Dynamic Irradiance
160
```

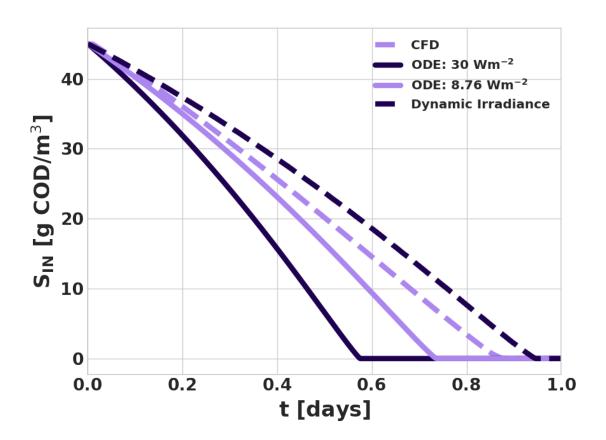


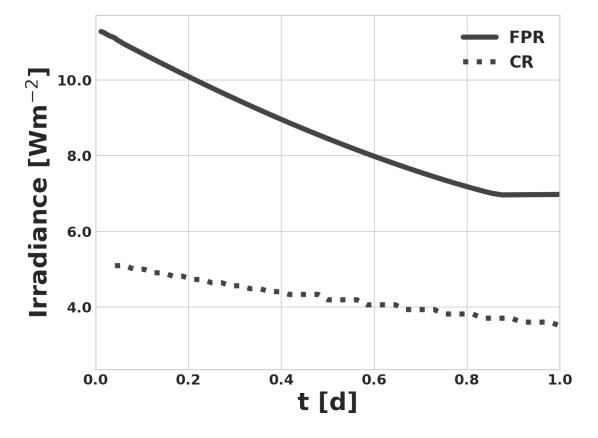
```
In [14]: fig2 = plt.figure(figsize=(8,6), dpi=120)
         ax1 = fig2.add_subplot(111)
         ax1.tick_params(axis="both",
                        reset=False,
                        labelsize=16)
         font = {'family' : 'DejaVu Sans',
                 'weight' : 'bold',
                 'size' : 14}
         plt.rc('font', **font)
         plt.plot(data["time"]/86400,
                  data["SAC"]*1000,
                  "--",
                  label='CFD',
                  linewidth=5,
                  color="#ac86ef")
         plt.plot(df_30["time"],
                  df 30["SACout"],
                  label='ODE: 30$\mathbf{Wm^{-2}}$',
                  linewidth=5,
                  color="#1d004f")
         plt.plot(df_10["time"],
                  df_10["SACout"],
                  label='ODE: 8.76\\mathbf{\m^{-2}}\\,
                  linewidth=5,
                  color="#ac86ef")
         plt.plot(df_dyn["time"],
                  df_dyn["SACout"],
                  label='Dynamic Irradiance',
                  linewidth=5,
                  color="#1d004f")
         plt.legend(fontsize=12)
         plt.xlabel(r'$\mathbf{t}\$ [days]', fontweight='bold', fontsize=20)
         plt.ylabel(r'$\mathbf{S_{AC}}$ [g COD/m$^3$]', fontweight='bold', fontsize=20)
         plt.xlim([0, 1])
         fig2.savefig("graphs/SAC_evol.eps")
```



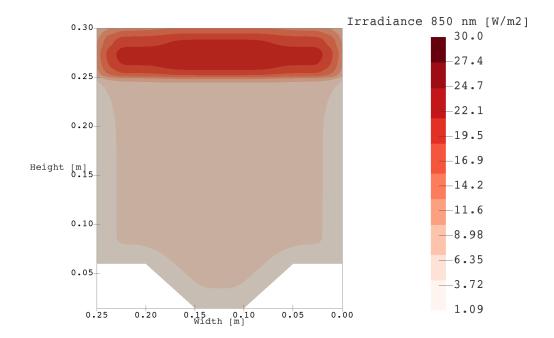
```
In [15]: fig3 = plt.figure(figsize=(8,6), dpi=120)
         ax1 = fig3.add_subplot(111)
         ax1.tick_params(axis="both",
                        reset=False,
                        labelsize=16)
         font = {'family' : 'DejaVu Sans',
                 'weight' : 'bold',
                 'size' : 14}
         plt.rc('font', **font)
         plt.plot(data["time"]/86400,
                  data["SIN"]*1000,
                  "--",
                  label='CFD',
                  linewidth=5,
                  color="#ac86ef")
         plt.plot(df_30["time"],
```

```
df_30["SINout"],
         label='ODE: 30 $\mathbf{Wm^{-2}}$',
         linewidth=5,
         color="#1d004f")
plt.plot(df_10["time"],
         df_10["SINout"],
         label='ODE: 8.76 \mathbf{Wm^{-2}}',
         linewidth=5,
         color="#ac86ef")
plt.plot(df_dyn["time"],
         df_dyn["SINout"],
         "--",
         label='Dynamic Irradiance',
         linewidth=5,
         color="#1d004f"
plt.legend(fontsize=12)
plt.xlabel(r'$\mathbf{t}$ [days]', fontweight='bold', fontsize=20)
plt.ylabel(r'\$\mathbb{S}_{IN}) \ [g\ COD/m\$^3\$]', \ fontweight='bold', \ fontsize=20)
plt.xlim([0, 1])
fig3.savefig("graphs/SIN_evol.eps")
```





Out[18]:



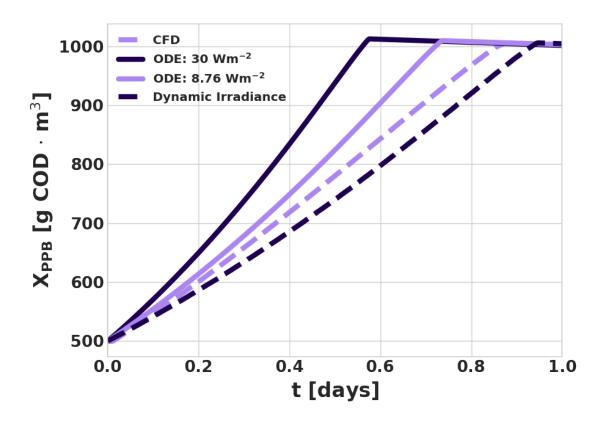
```
In [19]: import matplotlib as mpl
         mpl.rcParams['lines.linewidth'] = 5
         ## Plotting
         # plot for substrate concentration over time
         fig = plt.figure(figsize=(8,6), dpi=100)
         ax = fig.add_subplot(221)
         ax.tick_params(axis='both',
                        reset=False,
                        labelsize=12)
         plt.ylabel("\$\mathbb{S}_{AC})$ (\$\mathbb{C}D\, \cdot L^{-1})$)",
                   fontsize=16, fontweight='bold')
         #plt.xlabel("time (d)", fontsize=20, fontweight='bold')
         plt.plot(data['time']/3600/24, data['SAC']*1000, '--',
                 label='CFD solution',
                 color='#ac86ef')
         plt.plot(df_dyn['time'], df_dyn['SACout'], '--',
```

```
label='dynamic irradiance',
      color='#1d004f')
plt.plot(df_30['time'], df_30['SACout'],
      label='maximum nominal irradiance',
      color='#1d004f')
plt.plot(df_10['time'], df_10['SACout'],
      label='average irradiance',
      color='#ac86ef')
####### SAC PLOT
ax = fig.add_subplot(222)
ax.yaxis.tick_right()
ax.yaxis.set_label_position('right')
ax.tick_params(axis='both',
            reset=False,
            labelsize=12)
plt.ylabel("\$\mathbb{S}_{S}) (\mathbb{S}_{mathbf\{mgCOD\}, \cdot L^{-1}\})",
        fontsize=16, fontweight='bold')
plt.plot(data['time']/3600/24, data['SS']*1000, '--',
      label='CFD solution',
      color='#ac86ef')
plt.plot(df_dyn['time'], df_dyn['SSout'], '--',
      label='dynamic irradiance',
      color='#1d004f')
plt.plot(df_30['time'], df_30['SSout'],
      label='maximum nominal irradiance',
      color='#1d004f')
plt.plot(df_10['time'], df_10['SSout'],
      label='average irradiance',
      color='#ac86ef')
####### SIN PLOT
ax = fig.add_subplot(223)
ax.tick_params(axis='both',
```

```
reset=False,
             labelsize=12)
plt.ylabel("\$\mathbb{S}_{IN})$ (\$\mathbb{mgN}, \cdot L^{-1})$)",
         fontsize=16, fontweight='bold')
#plt.xlabel("time (d)", fontsize=20, fontweight='bold')
plt.plot(data['time']/3600/24, data['SIN']*1000, '--',
       label='CFD solution',
       color='#ac86ef')
plt.plot(df_dyn['time'], df_dyn['SINout'], '--',
       label='dynamic irradiance',
       color='#1d004f')
plt.plot(df_30['time'], df_30['SINout'],
       label='maximum nominal irradiance',
       color='#1d004f')
4
plt.plot(df_10['time'], df_10['SINout'],
       label='average irradiance',
       color='#ac86ef')
####### STP PI.OT
ax = fig.add_subplot(224)
ax.yaxis.tick_right()
ax.yaxis.set_label_position('right')
ax.tick_params(axis='both',
             reset=False,
             labelsize=12)
plt.ylabel("$\mathbf{S {IP}}$ ($\mathbf{mgP\, \cdot L^{-1}}$)",
         fontsize=16, fontweight='bold')
#plt.xlabel("time (d)", fontsize=20, fontweight='bold')
plt.plot(data['time']/3600/24, data['SIP']*1000, '--',
       label='CFD solution',
       color='#ac86ef')
plt.plot(df_dyn['time'], df_dyn['SIPout'], '--',
       label='dynamic irradiance',
       color='#1d004f')
plt.plot(df_30['time'], df_30['SIPout'],
```

```
label='maximum nominal irradiance',
               color='#1d004f')
    plt.plot(df_10['time'], df_10['SIPout'],
               label='average irradiance',
               color='#ac86ef')
    fig.text(0.5, -0.05, 'time (d)', fontsize=20, fontweight='bold', ha='center')
     #plt.xlabel("time (d)", fontsize=20, fontweight='bold')
    fig.tight_layout()
    fig.savefig('graphs/substrate_evolution.eps', bbox_inches='tight')
    500
                                                                                  200
\mathsf{S}_{\mathsf{AC}} (mgCOD \cdot \mathsf{L}^{-1})
    400
                                                                                  150
                                                                                       S<sub>s</sub> (mgCOD
    300
                                                                                  100
    200
    100
                                                                                  50
         0.0
                   0.5
                             1.0
                                               0.0
                                                         0.5
                                                                    1.0
                                                                              1.5
      40
 S<sub>IN</sub> (mgN·L<sup>-1</sup>)
     30
     20
     10
                                                         0.5
         0.0
                   0.5
                              1.0
                                        1.5
                                               0.0
                                                                    1.0
                                                                              1.5
                                     time (d)
```

```
font = {'family' : 'DejaVu Sans',
        'weight' : 'bold',
        'size' : 14}
plt.rc('font', **font)
plt.plot(data["time"]/86400,
         data["XPB"]*1000,
         "--",
         label='CFD',
         linewidth=5,
         color="#ac86ef"
plt.plot(df_30["time"],
         df_30["XPBout"],
         label='ODE: 30 $\mathbf{Wm^{-2}}$',
         linewidth=5,
         color="#1d004f")
plt.plot(df_10["time"],
         df_10["XPBout"],
         label='ODE: 8.76 $\mathbf{Wm^{-2}}$',
         linewidth=5,
         color="#ac86ef")
plt.plot(df_dyn["time"],
         df_dyn["XPBout"],
         label='Dynamic Irradiance',
         linewidth=5,
         color="#1d004f"
         )
plt.legend(fontsize=12)
plt.xlabel(r'$\mathbf{t}\$ [days]', fontweight='bold', fontsize=20)
plt.ylabel(r'$\mathbf{X_{PPB}}$ [g COD $\cdot$ m$^3$]', fontweight='bold', fontsize=2
plt.xlim([0, 1])
fig3.savefig("graphs/XPB_evol.eps")
```



```
In [21]: df_dyn.to_csv("./data/flat_dyn.csv")
         df_10.to_csv("./data/flat_10.csv")
         df_30.to_csv("./data/flat_30.csv")
         data.to_csv("./data/flat_cfd.csv")
In [22]: # Streamlines for the stirred vessel
         stream = pd.read_csv("./data/cylinder.csv")["GLambda_0"]
         streamtime = pd.read_csv("./data/cylinder.csv")["IntegrationTime"]
         print(f"Length of G850 is {len(stream)}.")
         print(f"Length of time array is {len(streamtime)}.")
         # This is for the flat plate reactor
         lst = []
         M = 500
         j = 0
         starter = 30
         j = starter
         while starter < len(streamtime)-2:</pre>
             while streamtime[j] < streamtime[j+1]:</pre>
                 j = j + 1
             if j - starter > M:
```

```
lst.append([starter, j])
             starter = j + 1
             j = starter
Length of G850 is 223647.
Length of time array is 223647.
In [23]: # Determination of Law
         import scipy.stats as stats
         part0 = stream[lst[0][0]:lst[0][1]]
         time0 = streamtime[lst[0][0]:lst[0][1]]
         meanG = np.log(stream[lst[0][0]:lst[0][1]]).mean()
         stdG = np.log(part0).std()
         maxG = part0.max()
         minG = part0.min()
         particle_cyl = part0
         time_cyl = time0
         minG_cyl = minG
         loc_cyl, scale_cyl = stats.expon.fit(particle_cyl, floc=minG)
         x_{cyl} = np.linspace(0,30)
         dist_cyl = stats.expon.pdf(x_cyl, loc_cyl, scale_cyl)
         weights_cyl = np.ones_like(particle_cyl)/float(len(particle_cyl))
         #print(loc_cyl, scale_cyl)
         #print(f'rate = {1/scale_cyl}')
         #print(minG cyl)
```

6 S.5 Summary of Short Term Dynamic Irradiance

The short term evolution of the radiative field, based on the flow field is summarised below. The first row corresponds to the time series evolution of the radiative field and the probability distribution (expon distribution) of an experienced irradiance for a particle for the flat plate reactor. The bottom row is the same, but for the cylinder.

```
In [24]: # Generate Graph for Publication
    import matplotlib as mpl
    mpl.rcParams['lines.linewidth'] = 5
    from matplotlib.ticker import MaxNLocator
    set_color = "#444444"
    plt.style.use("seaborn-whitegrid")
    font = {'family' : 'DejaVu Sans',
```

```
'weight' : 'bold',
                      'size' : 16}
plt.rc('font', **font)
def format_ticks():
          plt.gca().yaxis.set_major_locator(MaxNLocator(prune='lower'))
             plt.gca().xaxis.set\_major\_formatter(StrMethodFormatter('\{x:,.1f\}')) \ \# \ 1 \ decimal \ formatter('\{x:,.1f\}')
          plt.locator_params(nbins=5)
          plt.locator_params(numticks=5)
            if x == "xy":
                        plt.gca().yaxis.set\_major\_formatter(StrMethodFormatter('\{x:,.1f\}')) \ \# \ 1 \ decinonic formatter(strMethodFormatter('\{x:,.1f\}')) \ \# \ 1 \ decinonic formatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter(strMethodFormatter
fig2 = plt.figure(figsize=(10,7.5), frameon=False, dpi=200)
ax = fig2.add_subplot(223)
ax.tick_params(axis='both',
                                      reset=False,
                                      labelsize=16)
plt.hist(particle_flat, 60, weights=weights_flat, color=set_color)
plt.ylabel('Frequency', fontsize=16, fontweight='bold', color="black")
plt.xlim([0, 25])
ax.text(20, -0.015, "$\mathbf{Irradiance\, \, [W m^{-2}]}$", fontsize=16)
format_ticks()
ax = fig2.add_subplot(224)
ax.tick_params(axis='both',
                                      reset=False,
                                      labelsize=16)
plt.hist(particle_cyl, 60, weights=weights_cyl, color=set_color)
plt.xlim([0, 25])
format_ticks()
ax = fig2.add_subplot(222)
ax.tick_params(axis='both',
                                      reset=False,
                                      labelsize=16)
plt.plot(streamtime[lst[0][0]:lst[0][1]], stream[lst[0][0]:lst[0][1]],
                      color=set_color, linewidth=4)
plt.title("CR", fontweight='bold')
plt.xlim([0, 30])
plt.ylim([0, 25])
format_ticks()
```

```
ax = fig2.add_subplot(221)
    ax.tick_params(axis='both',
                   reset=False,
                   labelsize=16)
    plt.plot(streamtime_flat[lst_flat[1][0]:lst_flat[1][1]], stream_flat[lst_flat[1][0]:lst_flat[1][0]
             color=set_color, linewidth=4)
    plt.title("FPR", fontweight='bold')
    plt.xlim([0, 30])
    plt.ylim([0, 25])
    plt.ylabel('$\mathbf{Irradiance\, \, [W m^{-2}}]$', fontsize=16, color="black")
    ax.text(30, -4.0, "t [s]", fontsize=16)
    format_ticks()
    ax.text(0.5,21.5,"a", fontsize=24, fontweight="bold")
    ax.text(63,21.5,"b", fontsize=24, fontweight="bold")
    ax.text(0.5,-8.5,"c", fontsize=24, fontweight="bold")
    ax.text(63,-8.5,"d", fontsize=24, fontweight="bold")
    fig2.savefig("graphs/G850_short.pdf")
    plt.show()
                                                           CR
                     FPR
    25
                                          25
                                                                          b
 Irradiance [Wm<sup>-2</sup>]
                                          20
    20
    15
                                          15
                                          10
    10
      5
                                           5
                                     30
t [s]
                 10
                                                                 20
       0
                           20
                                            0
                                                                            30
  0.08 C
                                                                          d
                                        0.20
Preduency
0.04
0.02
                                        0.15
                                        0.10
                                        0.05
```

25

tu 25 0 5 Irradiance [Wm⁻²]

10

15

20

25

0

5

10

15

20

```
In [25]: time0.index = pd.RangeIndex(len(time0.index))
       part1 = np.concatenate(60*24*3*[part0])
       time1 = np.concatenate(60*24*3*[time0])
        # Create a deltaTime list
       deltaTime = []
       for i in range(len(time0)-1):
           deltaTime.append(time0[i+1] - time0[i])
       time1 = [0]
        # Turn deltaTime into a np.array
       deltaTime = np.concatenate((60*24*3+4)*[np.array(deltaTime)])
       for i in range(len(deltaTime)):
           time1.append(time1[-1]+deltaTime[i])
        #for i in range(3*60*24):
       len(deltaTime)
       time2 = np.asarray(time1)/86400
       f = interp1d(time2, part1[0:len(time1)])
       Nsteps = 800
# USER INPUT
       num\_steps = 10000
       days = 172800/3600/24
       SS0 = 200.0
       SACO = 500.0
       SICO = 7.0E-6
       SH20 = 0.0
       SINO = 45 # 45
       SIPO = 8.15 # 8.15
       SIO = 150.0
       XPBO = 500.0
       XSO = 300.0
       XIO = 100.0
        y0 = [SSO, SACO, SICO, SH2O, SINO, SIPO, SIO, XPBO, XSO, XIO,]
        # Interpolate the irradiance array
       t_new = np.linspace(0, time2.max(), num_steps)
       G850 = f(t_new)
```

```
y_out = [y0]
        # Loop over each point in the dataset
        for step in range(num_steps-1):
            t = [t_new[step], t_new[step+1]]
            # Get G850 for this step
            u = G850[step:step+1]
            # Solve the ODEs
            soln = odeint(pam_batch, y0, t, args=(u,))
            y_inter = np.float64(soln[1])
            y_out.append(y_inter)
            y0 = y_inter.tolist()
        y_out = np.asarray(y_out)
        cyl_dyn = pd.DataFrame((y_out))
        cyl_dyn['SSout'] = y_out[:,0]
        cyl_dyn['SACout'] = y_out[:,1]
        cyl_dyn['SICout'] = y_out[:,2]
        cyl_dyn['SH2out'] = y_out[:,3]
        cyl_dyn['SINout'] = y_out[:,4]
        cyl_dyn['SIPout'] = y_out[:,5]
        cyl_dyn['Slout'] = y_out[:,6]
        cyl_dyn['XPBout'] = y_out[:,7]
        cyl_dyn['XSout'] = y_out[:,8]
        cyl_dyn['XIout'] = y_out[:,9]
        cyl_dyn['time'] = t_new
# USER INPUT
        num_steps = 800
        days = 120604/3600/24
        SS0 = 200.0
        SACO = 500.0
        SICO = 7.0E-6
        SH20 = 0.0
        SINO = 45 # 45
        SIP0 = 8.15
                    # 8.15
        SIO = 150.0
```

Store our state outputs here

```
XPB0 = 500.0
        XSO = 300.0
        XIO = 100.0
        y0 = [SSO, SACO, SICO, SH2O, SINO, SIPO, SIO, XPBO, XSO, XIO,]
        time_ode = np.linspace(0,days,num_steps)
        #u_t = interp1d(time_input.squeeze(), full_G850.squeeze())
        # Store our state outputs here
        y_out = [y0]
        # Loop over each point in the dataset
        for step in range(num_steps - 1):
           t = [time_ode[step], time_ode[step+1]]
           # Get G850 for this step
           G850 = 30
           # Solve the ODEs
           soln = odeint(pam_batch, y0, t, args=(G850,))
           y_inter = np.float64(soln[1])
           y_out.append(y_inter)
           y0 = y_inter.tolist()
        y_out = np.asarray(y_out)
        cyl_30 = pd.DataFrame(y_out)
        cyl_30['SSout'] = y_out[:,0]
        cyl 30['SACout'] = y out[:,1]
        cyl_30['SICout'] = y_out[:,2]
        cyl_30['SH2out'] = y_out[:,3]
        cyl_30['SINout'] = y_out[:,4]
        cyl_30['SIPout'] = y_out[:,5]
        cyl_30['SIout'] = y_out[:,6]
        cyl_30['XPBout'] = y_out[:,7]
        cyl_30['XSout'] = y_out[:,8]
        cyl_30['XIout'] = y_out[:,9]
        cyl_30['time'] = time_ode
# USER INPUT
        num_steps = 800
```

```
days = 120604/3600/24
SS0 = 200.0
SACO = 500.0
SICO = 7.0E-6
SH20 = 0.0
SINO = 45 # 45
SIP0 = 8.15
            # 8.15
SIO = 150.0
XPB0 = 500.0
XSO = 300.0
XIO = 100.0
y0 = [SS0, SAC0, SIC0, SH20, SIN0, SIP0, SI0, XPB0, XS0, XI0,]
time_ode = np.linspace(0,days,num_steps)
#u_t = interp1d(time_input.squeeze(), full_G850.squeeze())
# Store our state outputs here
y_{out} = [y0]
# Loop over each point in the dataset
for step in range(num_steps - 1):
   t = [time_ode[step], time_ode[step+1]]
    # Get G850 for this step
   G850 = beer_lambert(30, 53+15, 0.015)
    # Solve the ODEs
   soln = odeint(pam_batch, y0, t, args=(G850,))
   y_inter = np.float64(soln[1])
   y_out.append(y_inter)
   y0 = y_inter.tolist()
y_out = np.asarray(y_out)
cyl_10 = pd.DataFrame(y_out)
cyl_10['SSout'] = y_out[:,0]
cyl_10['SACout'] = y_out[:,1]
cyl_10['SICout'] = y_out[:,2]
cyl_10['SH2out'] = y_out[:,3]
cyl_10['SINout'] = y_out[:,4]
cyl_10['SIPout'] = y_out[:,5]
```

```
cyl_10['SIout'] = y_out[:,6]
        cyl_10['XPBout'] = y_out[:,7]
        cyl_10['XSout'] = y_out[:,8]
        cyl_10['XIout'] = y_out[:,9]
        cyl 10['time'] = time ode
In [29]: # Import CFD Solution of the stirred vessel.
        cyl_cfd = pd.read_csv("./data/round_cfd.csv")
        cyl_cfd.columns = ["index", "time",
                      "SS", "SAC", "SIN", "SIP",
                      "XPB"
######
        plt.rcParams['lines.linewidth'] = 3.0
        dark = "#444444"
        # Set the plot styles here
        def CFDplot(x, y, shade):
           plt.plot(x, y, label="CFD", color=shade)
        def ODE1plot(x, y, shade):
           style = "^"
           plt.plot(x.iloc[::25], y.iloc[::25], style,
                   label="$\mathregular{ODE1\, \, [I_{max}]}$",
                   color=shade)
        def ODE2plot(x, y, shade):
           style = "D"
           plt.plot(x.iloc[::25], y.iloc[::25], style,
                   label="$\mathregular{ODE2\, \, [\overline{I}]]$",
                   color=shade)
        def ODE3plot(x, y, shade):
           style = "X"
           plt.plot(x.iloc[::250], y.iloc[::250], style,
                   label="ODE3 [I=f(t)]",
                    color=shade)
        fig1 = plt.figure(figsize=(15, 15), dpi=180)
        fig1.set_edgecolor("black")
        ax1 = fig1.add subplot(321)
        ax1.tick_params(axis='both',
                     reset=False,
```

```
labelsize=20)
plt.ylabel("\$\mathbb{X}_{PB}) [\$\mathbb{COD}, \cdot L^{-1}}",
        fontsize=24, fontweight='bold')
ODE1plot(df_30['time'], df_30['XPBout'], dark)
ODE2plot(df 10['time'], df 10['XPBout'], dark)
ODE3plot(df_dyn['time'], df_dyn['XPBout'], dark)
CFDplot(data['time']/3600/24, data['XPB']*1000, dark)
plt.title("FPR", fontsize=24, fontweight='bold')
plt.xlim([0,1])
plt.text(0.05, 950, "a", fontsize=30)
######
         ax2 = fig1.add_subplot(322, sharex=ax1, sharey=ax1)
ax2.yaxis.tick_right()
ax2.yaxis.set label position('right')
ODE1plot(cyl 30['time'], cyl 30['XPBout'], dark)
ODE2plot(cyl_10['time'], cyl_10['XPBout'], dark)
ODE3plot(cyl_dyn['time'], cyl_dyn['XPBout'], dark)
CFDplot(cyl_cfd['time']/3600/24, cyl_cfd['XPB']*1000, dark)
plt.xlim([0,1])
ax2.tick_params(axis='both',
            reset=False,
            labelsize=20)
plt.title("CR", fontsize=24, fontweight='bold')
plt.text(0.05, 950, "b", fontsize=30)
######
          ax3 = fig1.add subplot(323)
ax3.tick_params(axis='both',
            reset=False,
            labelsize=20)
plt.ylabel("$\mathbf{S {AC}}$ [$\mathbf{mg COD\, \cdot L^{-1}}$]",
        fontsize=24, fontweight='bold')
ODE1plot(df_30['time'], df_30['SACout'], dark)
ODE2plot(df_10['time'], df_10['SACout'], dark)
ODE3plot(df_dyn['time'], df_dyn['SACout'], dark)
```

```
CFDplot(data['time']/3600/24, data['SAC']*1000, dark)
plt.xlim([0,1])
plt.text(0.9, 425, "c", fontsize=30)
####### PLOT SAC FOR CR
ax4 = fig1.add_subplot(324, sharex=ax2)
ax4.yaxis.tick right()
ax4.yaxis.set_label_position('right')
ODE1plot(cyl_30['time'], cyl_30['SACout'], dark)
ODE2plot(cyl_10['time'], cyl_10['SACout'], dark)
ODE3plot(cyl_dyn['time'], cyl_dyn['SACout'], dark)
CFDplot(cyl_cfd['time']/3600/24, cyl_cfd['SAC']*1000, dark)
plt.xlim([0,1])
ax4.tick_params(axis='both',
           reset=False,
           labelsize=20)
plt.text(0.9, 425, "d", fontsize=30)
######## PLOT SS FOR FPR
ax5 = fig1.add_subplot(325)
ax5.tick_params(axis='both',
           reset=False,
           labelsize=20)
plt.ylabel("\$\mathbb{S}_{S}) [\$\mathbb{m}_{mgCOD}, \cdot L^{-1}}",
       fontsize=24, fontweight='bold')
ODE1plot(df 30['time'], df 30['SSout'], dark)
ODE2plot(df_10['time'], df_10['SSout'], dark)
ODE3plot(df_dyn['time'], df_dyn['SSout'], dark)
CFDplot(data['time']/3600/24, data['SS']*1000, dark)
plt.xlim([0, 1])
plt.text(0.9, 175, "e", fontsize=30)
######## PLOT SS FOR CF
ax6 = fig1.add_subplot(326)
ax6.yaxis.tick_right()
```

```
ax6.yaxis.set_label_position('right')
     ax6.tick_params(axis='both',
                        reset=False,
                        labelsize=20)
     ODE1plot(cyl_30['time'], cyl_30['SSout'], dark)
     ODE2plot(cyl_10['time'], cyl_10['SSout'], dark)
     ODE3plot(cyl_dyn['time'], cyl_dyn['SSout'], dark)
     CFDplot(cyl_cfd['time']/3600/24, cyl_cfd['SS']*1000, dark)
     plt.xlim([0, 1])
     plt.text(0.9, 175, "f", fontsize=30)
     plt.legend(loc=(-1.15, -0.6), ncol=4, fontsize=20)
     fig1.text(0.45, 0.060, "t [d]", fontsize=24)
     fig1.savefig("graphs/growth_kinetics.pdf")
                       FPR
                                                                  CR
X<sub>PB</sub> [mgCOD · L<sup>-1</sup>]
   1000
                                                    b
          a
    800
                                                                                      800
    600
                                                                                      600
      0.0
              0.2
                     0.4
                            0.6
                                   0.8
                                          1.0
                                                        0.2
                                                               0.4
                                                                      0.6
                                                                             0.8
                                                                                    1.0
S<sub>AC</sub> [mgCOD·L<sup>-1</sup>]
                                                                                  d
                                        C
                                                                                      400
                                                                                      200
                                          1.0
      0.0
              0.2
                            0.6
                                   0.8
                                                 0.0
                                                        0.2
                                                               0.4
                                                                      0.6
                                                                             0.8
                                                                                    1.0
                     0.4
 S<sub>5</sub> [mgCOD · L<sup>-1</sup>]
                                                                                      200
                                        e
                                                                                      100
                                                                                      50
      0.0
              0.2
                     0.4
                            0.6
                                   0.8
                                          1.0
                                                 0.0
                                                        0.2
                                                               0.4
                                                                      0.6
                                                                             8.0
                                                                                    1.0
                                        t [d]
                                    ODE2 [Ī]
                ODE1 [I<sub>max</sub>]
                                                      ODE3 [I=f(t)]
                                                                            CFD
```

7 S.6 Long-term dynamic irradiance

One of the final mechanisms to discuss is the long term evolution of the irradiance field as the biomass grows. The figures have been exported to the paper document, and also pasted below. Note that there have only been 10 levels of discretisation for the colour scheme. It has been anecdotally found that having too fine a resolution when displaying color graphs can detract from the physics of the system.

