BPL_IEC_operation

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In this notebook we show operation of a typical ion-exchange chromatography step. The impact of pH is also illustrated.

The model is based on the simplified model [1].

```
In [1]: run -i BPL IEC fmpy explore.py
        Linux - run FMU pre-comiled OpenModelica 1.21.0
        Model for bioreactor has been setup. Key commands:
         par()change of parameters and initial values

    change initial values only

         - init()
                    - simulate and plot
         - simu()
         - newplot() - make a new plot
         - show() - show plot from previous simulation
         - disp()
                      - display parameters and initial values from the last sim
        ulation
         - describe() - describe culture, broth, parameters, variables with val
        ues/units
        Note that both disp() and describe() takes values from the last simulati
        Brief information about a command by help(), eg help(simu)
        Key system information is listed with the command system info()
In [2]: plt.rcParams['figure.figsize'] = [30/2.54, 24/2.54]
```

1 Typical parameters an ion exchange chromatography column step

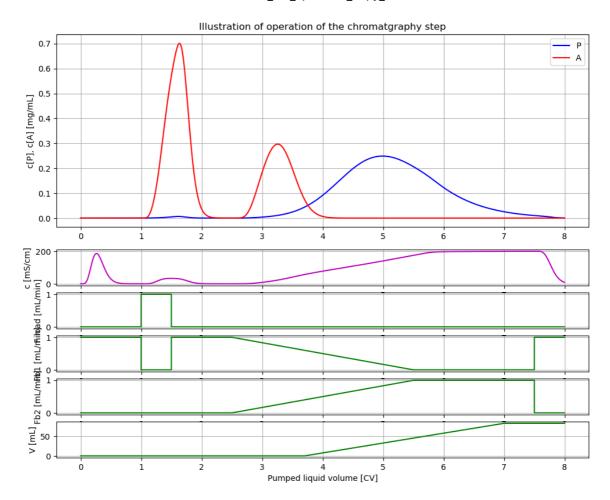
```
In [3]: # From given column height (h) diameter (d) and linear flow rate (lfr)
# actual column volume (V) and volume flow rate (VFR) are calculated below

from numpy import pi
h = 20.0
d = 1.261
a = pi*(d/2)**2
V = h*a
print('V =', np.round(V,1), '[mL]')

lfr = 48
VFR = a*lfr/60
print('VFR =', np.round(VFR,1), '[mL/min]')

V = 25.0 [mL]
VFR = 1.0 [mL/min]
In [4]: # Sample concentration product P_in and antagonist A_in
par(P_in = 1.0)
```

```
par(A in = 1.0)
par(E in = 0.0)
# Column properties are described by the size and binding capacity of the
par(height = h)
par(diameter = d)
par(Q av = 6.0)
# Resin parameters - default values used
# Remaining salt koncentration in the column from prvious batch and elimi
init(E 0 = 50)
# Salt concentration of the desorption buffer
par(E in desorption buffer = 8.0)
# Flow rate rate through the
par(LFR=lfr)
# Switching points during operation are conveniently described in terms of
CV \text{ ekv} = 1.0
CV_ads = 0.5
CV wash = 1.0
CV desorb = 3.0
CV_start_pool = 1.2
CV stop pool = 4.5
CV wash2 = 2.5
par(scale_volume=True, start_adsorption=CV_ekv*V, stop_adsorption=(CV_ekv
par(start desorption=(CV ekv+CV ads+CV wash)*V, stationary desorption=(CV
par(stop desorption=7.5*V)
par(start_pooling=(CV_ekv+CV_ads+CV_wash+CV_start_pool)*V, stop_pooling=(
# Simulation and plot of results
newplot(title='Illustration of operation of the chromatgraphy step', plot
simu((CV ekv+CV ads+CV wash+CV desorb+CV wash2)*V/VFR)
```



Comments of steps of operations:

- 1. Time: 0-1 hours equilibration. Just to illustrate the equilibration process the first part of the column is given an initial value of salt concentration.
- 2. Time: 1-1.5 hours sample is loaded on the column. The product P is adsorbed to the columne and just a small amount passes through and goes to the waste. The antagonist A is much less adsrobed.
- 3. Time: 1.5-2.5 hours washing 1. The column comes to equilibrium and both antagonist and product comes down to low levels.
- 4. Time: 2.5-5.5 hours desorption. A linear gradient of increaseing salt concentration is applied. First the antagonist and later the product comes out.
- 5. Time: 5.5-7.5 hours washing 2 The The column has constant salt concentration and stationary desorption.
- 6. Time: 3.7-7.0 hours pooling of product. The start- and stop of pooling are chosen with trade-off between maximizing the product pooled and minimize the amount of antagonist in the pooling.
- 7. Time: 7.5-8.0 hours desorption stopped and salt is washed out and preparation of the next batch to come.

Note that step 4 and 5 is parallel to step 6.

```
In [5]: # Check mass-balance of P and A
P_mass = model_get('tank_harvest.m[1]') + model_get('tank_waste.m[1]')
A_mass = model_get('tank_harvest.m[2]') + model_get('tank_waste.m[2]')
```

```
print('P_mass [mg] =', P_mass)
print('A_mass [mg] =', A_mass)

P_mass [mg] = 12.422130344156677
A mass [mg] = 12.488781164504793
```

2 The impact of the slope of the desorption gradient

```
In [6]:
         # Simulations showing the impact of change of slope of the desorption gra
         newplot(title='Impact of the slope of the gradient', plotType='Elution-co
         # Same gradienet as before
         par(start desorption=(CV ekv+CV ads+CV wash)*V, stationary desorption=(CV
         par(stop desorption=7.5*V)
         simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)
         # Gradeint finishes after 0.5 of the volume
         par(stationary desorption = (CV ekv + CV ads + CV wash + 0.5*CV desorb)*V
         simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)
         # Fradient finishes after 0.25 of the volume
         par(stationary desorption = (CV_ekv + CV_ads + CV_wash + 0.25*CV_desorb)*
         simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)
                                         Impact of the slope of the gradient
            0.7
            0.6
           0.5
          c[A] [mg/mL]
           0.4
           0.3
           0.2
            0.1
            0.0
           200
         c [mS/cm]
           Fb2 [mL/mftb]1 [mL/rffiltb]ad [mL/min]
             0
             1
             0
          []<sub>50</sub>
```

3 The impact of salt concentration in the sample

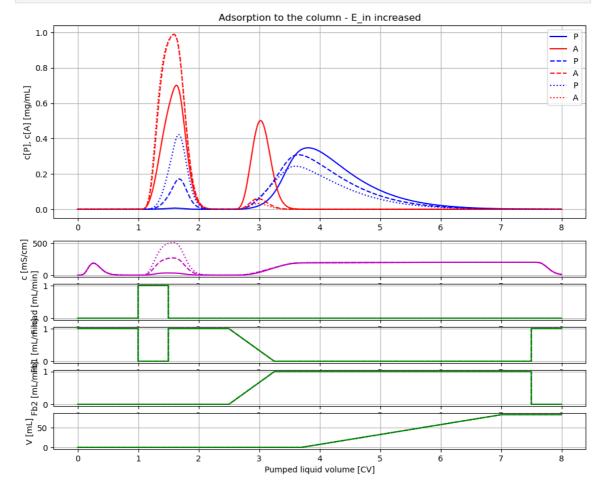
Pumped liquid volume [CV]

These values should be compared with the expected value 12.5 mg, i.e. half a column volume with sample concentration 1 mg/L. The difference is due to numerical errors during simulation.

```
In [7]: # Let us investigate the impact of increasing salt concetration in the sa
# Simulate and plot the results
newplot(title='Adsorption to the column - E_in increased', plotType='Elut

for value in [0, 10, 20]:
    par(E_in=value)
    simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)

# Restore default values
par(k2=0.05, k4=0.3, E_in=0)
```



Note, that increased salt concentration in the sample affect binding of both proteins. During adsorption less is bound. During desoprtion less product P can be harvested but the fraction of antagonist A may be lowered. Thus, some product is lost but the quality in terms of purity is improved.

4 The impact of change of binding strength due to pH

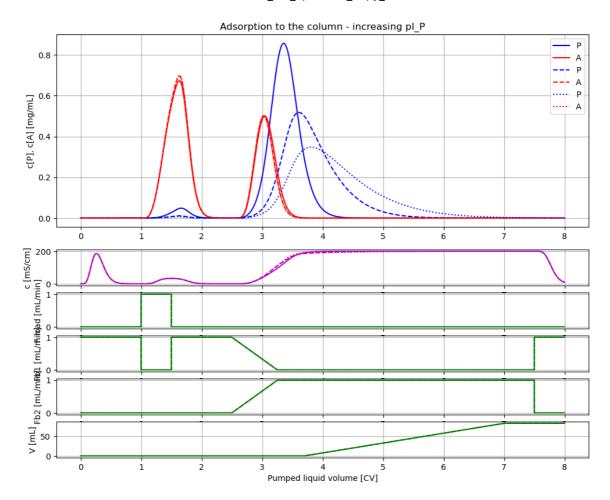
There are many factors that contribute to the binding strength. A most important factor is the pH-value of the resin and the characteristic iso-electric point of the protein. The

binding strenght can be seen as proportional to the difference.

The binding strength of the resin is described by the quotient KP=k1/k2 for the protein P and similarly KA=k3/k4 for the protein A.

Below a few help-functions that describe this idea of the pH difference and its impact on binding strength in terms of the parameters k1, k2, k3, and k4 of the protein-resin interaction.

```
In [8]: # Define function that describe the proportionality of binding strength of
         # the pH difference of the iso-electric point and the resin
         def KP pH sensitivity(pI P=8.0, pH resin=7.0):
             K P nom = 0.0
             coeff_pH = 6.0
             return K P nom + coeff pH*(pI P-pH resin)
         def KA pH sensitivity(pI A=7.1667, pH resin=7.0):
             K A nom = 0.0
             coeff pH = 1.0
             return K_A_nom + coeff_pH*(pI_A-pH_resin)
         def par pH(pI P=8.0, pI A=7.1667, pH resin=7.0, TP=3.33, TA=20.0):
             if (pI_P > pH_resin) & (pI_A > pH_resin):
                 par(k2 = 1/(TP*KP pH sensitivity(pI P=pI P, pH resin=pH resin)))
                 par(k4 = 1/(TA*KA_pH_sensitivity(pI_A=pI_A, pH_resin=pH_resin)))
                 print('Both pI P > pH resin and pI A > pH resin must hold - no pa
 In [9]: # The default parameters of the column
         disp('column')
         diameter: 1.261
         height: 20.0
         x m : 0.3
         k1 : 0.3
         k2 : 0.05
         k3 : 0.05
         k4 : 0.3
         Q av : 6.0
         E 0 : 50
In [10]: # Let us investigate the impact of change of the iso-electric pH for prot
         # Simulate and plot the results
         newplot(title='Adsorption to the column - increasing pI_P', plotType='Elu
         for value in [7.2, 7.6, 8.0]:
             par pH(pI P=value, pI A=7.1667, pH resin=7.0)
             simu((CV ekv+CV ads+CV wash+CV desorb+CV wash2)*V/VFR)
         # Restore default values
         par(k2 = 0.05, k4 = 0.3)
```



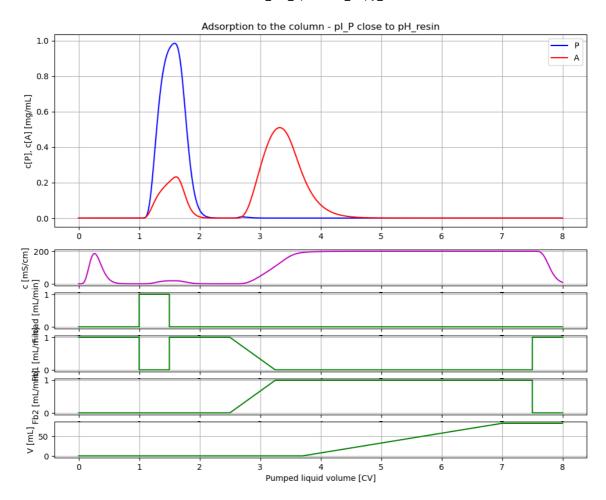
Note, with increasing pI_P the binding of P increase which leads less loss of product during adsorption. During desorption the peak height is lower with increasing binding strenght, but the total amoiunt of product P that can be harvested is higher, due to the smaller loss during adsorption.

```
In [11]: # Let us investigate the impact of pI_P close to pH_resin

# Simulate and plot the results
newplot(title='Adsorption to the column - pI_P close to pH_resin', plotTy

for value in [7.0001]:
    par_pH(pI_P=value, pI_A=8)
    simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)

# Restore default values
par(k2=0.05, k4=0.3)
```

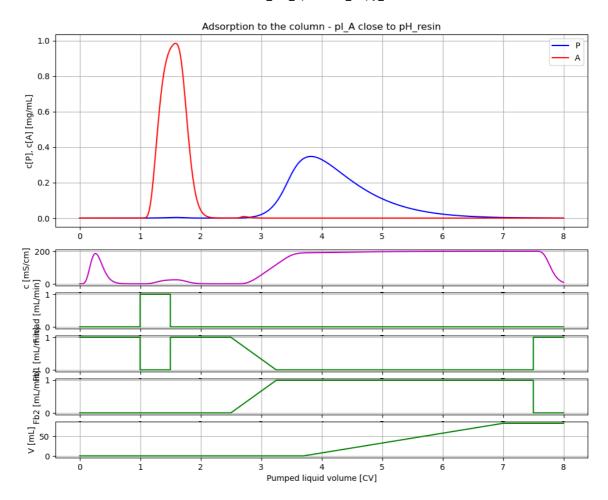


```
In [12]: # Let us investigate the impact of pI_A close to pH_resin

# Simulate and plot the results
newplot(title='Adsorption to the column - pI_A close to pH_resin', plotTy

for value in [7.001]:
    par_pH(pI_P=8.0, pI_A=value)
    simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)

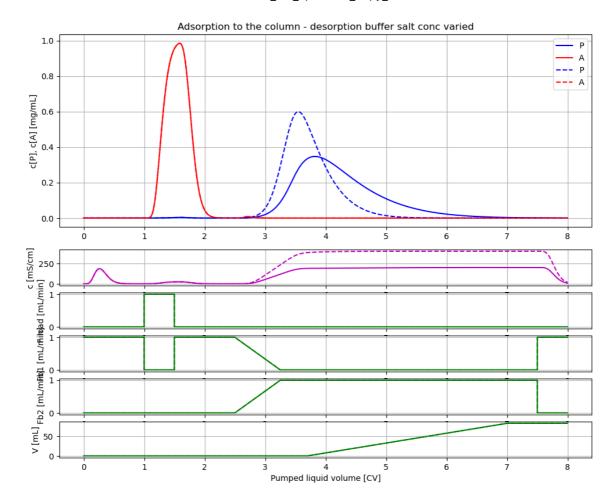
# Restore default values
par(k2=0.05, k4=0.3)
```



```
In [13]: # Let us also investigate the impact of salt concentration of the desorpt
# Simulate and plot the results
newplot(title='Adsorption to the column - desorption buffer salt conc var

for value in [8.0, 16.0]:
    par(E_in_desorption_buffer=value)
    par_pH(pI_P=8.0, pI_A=7.001, pH_resin=7.0)
    simu((CV_ekv+CV_ads+CV_wash+CV_desorb+CV_wash2)*V/VFR)

# Restore default values
par(E_in_desorption_buffer=8.0)
par(k2=0.05, k4=0.3)
```



5 Breakthrough curve often used during process development

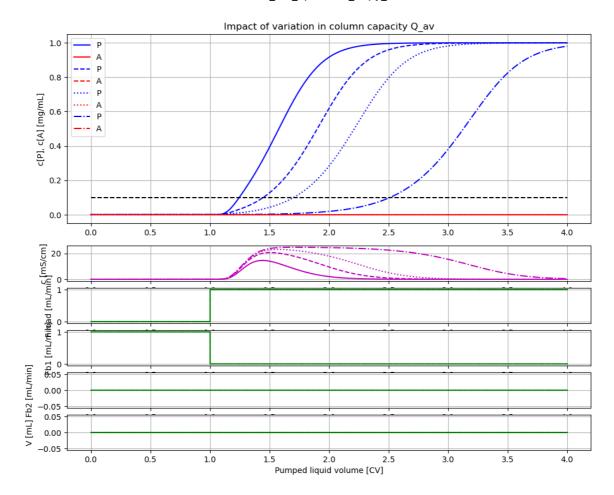
```
In [14]: # Experiment to check column capacity Q_av often called breakthrough curv
    par(P_in=1, A_in=0, E_in=0)
    init(E_0 = 0)
    par(Q_av=6.0)

par(scale_volume=True, start_adsorption=1*V, stop_adsorption=4.01*V)
    par(start_desorption=10*V, stationary_desorption=10.5*V, stop_desorption=
    par(start_pooling=11*V, stop_pooling=12*V)

newplot(title='Impact of variation in column capacity Q_av', plotType='Elfor value in [1, 2, 3, 6]: par(Q_av=value); simu(4.0*V/VFR)

# Linje för 10% UV
    ax1.plot([0,4], [0.1,0.1],'k--')

# Restore default parameters
    par(Q_av=6.0, A_in=1.0)
```



With greater column capacity Q_av the longer it takes before the concentration of protein start to increase. Note, that the salt concentration increase initially during adsorption but then go back to low levels. This phenomenon ia also seen experimentally.

6 Summary

The simplified simulation model was found useful to describe operational aspects of ion exchange chromtography.

- The model describe qualitatively well the impact of typical operational changes in the flow rate.
- The model also describe qualtively well the impact of changes in iso-electric point of the proteins relative the pH of the resin.
- The small deviations in salt concentration from linear increase during the gradient in the salt buffer is also what you see in reality.

References

- 1. Månsson, Jonas, "Control of chromatography comlumn in production scale", Master thesis TFRT-5599, Department of Automatic Control, LTH, Lund Sweden, 1998.
- 2. Pharmacia LKB Biotechnology. "Ion Exchange chromatography. Principles and Methods.", 3rd edition, 1991.

Appendix

```
In [15]:
        system info()
         System information
          -OS: Linux
          -Python: 3.8.16
          -Scipy: not installed in the notebook
          -FMPy: 0.3.15
          -FMU by: OpenModelica Compiler OpenModelica 1.21.0
          -FMI: 2.0
          -Type: ME
          -Name: BPL_IEC.Column_system
          -Generated: 2023-04-21T10:58:26Z
          -MSL: 3.2.3
          -Description: Bioprocess Library version 2.1.1
          -Interaction: FMU-explore for FMPy version 0.9.8
 In [ ]:
```