

Tutorial Dynamic Inflow

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Learning objectives:

The student will:

- Learn an approach for the implementation of three engineering models for the simulation of dynamic inflow: Pitt-Peters, Øye and Larsen-Madsen
- Compare how the three models result in different predictions of the dynamic induction

References for additional study of dynamic inflow

- W. Yu. The wake of an unsteady actuator disc. PhD thesis, Delft University of Technology, 2018. URL <http://resolver.tudelft.nl/uuid:0e3a2402-585c-41b1-81cf-a35753076dfc> (<http://resolver.tudelft.nl/uuid:0e3a2402-585c-41b1-81cf-a35753076dfc>)
- Wessel van der Deijl. Dynamic wind speed in dynamic inflow models. Master's thesis, Delft University of Technology, July 2018. URL <http://resolver.tudelft.nl/uuid:da585fef-bd98-4519-962b-e1ccc62c906a> (<http://resolver.tudelft.nl/uuid:da585fef-bd98-4519-962b-e1ccc62c906a>)
- Delphine De Tavernier and Carlos S. Ferreira. A new dynamic inflow model for vertical-axis wind turbines. Wind Energy, 23(5):1196–1209, 2020. doi: 10.1002/we.2480. URL <https://onlinelibrary.wiley.com/doi/abs/10.1002/we.2480> (<https://onlinelibrary.wiley.com/doi/abs/10.1002/we.2480>)
- David A. Peters. How dynamic inflow survives in the competitive world of rotorcraft aerodynamics. Journal of the American Helicopter Society, 54(1), 2009. ISSN 2161-6027. doi: doi:10.4050/JAHS.54.011001. URL <https://www.ingentaconnect.com/content/ahs/jahs/2009/00000054/00000001/art00003> (<https://www.ingentaconnect.com/content/ahs/jahs/2009/00000054/00000001/art00003>)

```
In [1]: # Here we import the python libraries we will use
import matplotlib.pyplot as plt # import pyplot library
import matplotlib.colors as colors # import colors library
import numpy as np # import numpy
# import matplotlib
```

The equations below express the induction factor a as a function of thrust coefficient C_T , and C_T as a function of a .

Our original assumption of the relation between loading on the actuator and induction is given by the equations below.

$$C_T = 4a(1 - a)$$

and

$$a = \frac{1}{2}(1 - \sqrt{1 - C_T})$$

Above a certain loading on the streamtube, the wake will enter "turbulent wake state" or even "vortex ring state". The relations above are no longer valid, and should be corrected. ### Glauert's correction for heavily loaded rotors

One proposed correction was defined by Glauert, and it is given by the equations below.

$$C_T = \begin{cases} 4a(1 - a), & \text{for } a < 1 - \frac{\sqrt{C_{T_1}}}{2} \\ C_{T_1} - 4(\sqrt{C_{T_1}} - 1)(1 - a), & \text{for } a \geq 1 - \frac{\sqrt{C_{T_1}}}{2} \end{cases}$$

$$a = \begin{cases} \frac{1}{2} - \frac{\sqrt{1 - C_T}}{2}, & \text{for } C_T < C_{T_2} \\ 1 + \frac{C_T - C_{T_1}}{4\sqrt{C_{T_1}} - 4}, & \text{for } C_T \geq C_{T_2} \end{cases}$$

and

$$C_{T_1} = 1.816 \text{ and } C_{T_2} = 2\sqrt{C_{T_1}} - C_{T_1}$$

In [2]: # induction as a function of thrust coefficient

```
def ainduction(CT,glauert=False):
    """
    This function calculates the induction factor 'a' as a function of thrust coefficient CT
    including Glauert's correction. Notice that here we are using the wind turbine notation of 'a'
    """
    if glauert:
        CT1=1.816;
        CT2=2*np.sqrt(CT1)-CT1
    else:
        CT1=0
        CT2=100

    a=np.zeros(np.shape(CT))
    a[CT>=CT2] = 1 + (CT[CT>=CT2]-CT1)/(4*(np.sqrt(CT1)-1))
    a[CT<CT2] = 0.5-0.5*np.sqrt(1-CT[CT<CT2])
    return a

def CTfunction(a, glauert = False):
    """
    This function calculates the thrust coefficient as a function of induction factor 'a'
    'glauert' defines if the Glauert correction for heavily loaded rotors should be used; default value is false
    """
    CT = np.zeros(np.shape(a))
    CT = 4*a*(1-a)
    if glauert:
        CT1=1.816;
        a1=1-np.sqrt(CT1)/2;
        CT[a>a1] = CT1-4*(np.sqrt(CT1)-1)*(1-a[a>a1])

    return CT
```

The Pitt-Peters dynamic inflow model

The linear Pitt-Peters dynamic-inflow model is shown in the equation below for the case of axial flow in x -direction

- for an annulus at radius r_j with area A_j
- for a loading on the annulus T
- where v_x is the perturbation velocity at the rotor (induced velocity)

$$\left[\frac{8}{3\pi} r_j \rho A_j \frac{dv_x}{dt} + 2\rho A_j v_x (U_\infty + v_x) \right] = T$$

The Pitt-Peters dynamic inflow model: application to heavy loaded cases

The equation for the Pitt-Peters dynamic-inflow model can be rewritten as

$$\frac{dv_x}{dt} = \frac{3\pi U_\infty^2}{16r_j} \left[\frac{T}{1/2\rho A_j U_\infty^2} - 4 \frac{v_x (U_\infty + v_x)}{U_\infty^2} \right]$$

and further as

$$\frac{dv_x}{dt} = \frac{3\pi U_\infty^2}{16r_j} [C_{T_{load}} - C_{T_{(v_x/U_\infty)}}]$$

where $C_{T_{load}}$ is the non-dimensional thrust applied to the annulus and $C_{T_{(v_x/U_\infty)}}$ is the expression for an equivalent quasi-steady loading based on the induction v_x/U_∞ and actuator disk momentum theory.

```
In [3]: # The Pitt-Peters dynamic inflow model

def pitt_peters(Ct,vind,Uinf,R,dt,glauert=False):
    # this function determines the time derivative of the induction at the annuli
    # Ct is the thrust coefficient on the actuator, vind is the induced velocity,
    # Uinf is the unperturbed velocity and R is the radial scale of the flow, nd dt is
    # the time step
    # it returns the new value of induction vindnew and the time derivative dvind_dt
    # glauert defines if glauert's high load correction is applied

    a=-vind/Uinf # determine the induction coefficient for the time step {i-1}
    Ctn= -CTfunction(a, glauert) # calculate the thrust coefficient from the induction
    # for the time step {i-1}

    dvind_dt = ( Ct-Ctn )/(16/(3*np.pi))*(Uinf**2/R) # calculate the time derivative
    # of the induction velocity
    vindnew = vind + dvind_dt*dt # calculate the induction at time {i} by time integration
    return vindnew, dvind_dt
```

```
In [4]: # we will now calculate and plot the solution of a step change in thrust coefficient

# we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

# define Ct and induction at t0
Ct0=np.array([-0.5]) # this is the value of thrust coefficient at time t0, the start
of the time array
vind0=-ainduction(-Ct0)*Uinf # this is the quasi-steady value of induction at time t
0, calculated from Ct0

# define quasi-steady solution of Ct and induction at t>=t1
Ct1=np.array([-0.85]) # this is the value of thrust coefficient at time t1
vind1=-ainduction(-Ct1)*Uinf # this is the quasi-steady value of induction at time t
1, calculated from Ct1

# define Ct as a function of time
Ct = np.zeros(np.shape(time))+Ct0 # we initialize the array of thrust coefficient, se
tting all initial values at Ct0

# change Ct for time above t1
t1=5 # we define t1, when Ct experiences a step change
Ct[time>=t1] = Ct1 # we correct the values of Ct for the instants that time is after
t1, to a value of Ct1. We therefore
# define the step change from Ct0 to Ct1

#set arrays for induction
vind = np.zeros(np.shape(time)) # we create the array of induction velocity
vind[0]=vind0 # we initialize the first value to the induction velocity at t0, calcul
ated from Ct0

# solve the equation in time of the value of induction by using the Pitt-Peters model
for i,val in enumerate(time[:-1]):
    vind[i+1]=pitt_peters(Ct[i+1],vind[i],Uinf,R,dt )[0]
```

```

In [5]: # plot figure the change in induction factor  $a$  calculated by the Pitt-peters model
        # for a step change in thrust coefficient
        # from  $C_{t_0}=0.5$  to  $C_{t_0}=0.85$ . The values are made non-dimensioned by the steady-solution values of  $a$ 
        # for the two values of  $C_{t_0}$ . Time  $t$  is made non-dimensional by the radius of the actuator  $R$  and
        # the unperturbed wind speed  $U_{\infty}$ , and set to zero at the moment of the step change.

plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn') # define a colormap for the figure
fig,ax1 = plt.subplots(figsize=[6,6]) # create figure, axis and define the size of the figure
ax2=ax1.twinx() # we twin the axis to create a secondary y-axis. therefore, we twin the x-axis
# we now plot the evolution of  $C_t$  over time
lns1=ax1.plot((time-t1)*R/Uinf, (Ct-Ct0)/(Ct1-Ct0), color='green',linestyle='dashdot', label=r' $C_{t_0}$ ') # notice the negative
                                                    # value as we will be using the notation for wind turbines
# we now plot the evolution of induction velocity over time in the secondary axis
lns2=ax2.plot((time-t1)*R/Uinf, (vind-vind0)/(vind1-vind0), color='blue',linestyle='-', label= r' $a$ ') # notice the negative

# define properties of the primary y-axis
# ax1.set_aspect(aspect=20.0) # set aspect ratio of the figure
ax1.set_xlabel(r' $t \frac{R}{U_{\infty}}$ ') # Label of x-axis
ax1.set_ylabel(r' $\frac{C_{t_0}-C_{t_1}}{C_{t_1}-C_{t_0}}$ ', color='green', fontsize=20) # Label of y-axis
ax1.set_xlim(-t1,10) # set limits of x-axis
ax1.set_ylim(-0.05,1.05) # set limits of x-axis
ax1.tick_params(axis='y', labelcolor='green') # set the color of the axis

# define properties of secondary axis
ax2.set_ylabel(r' $\frac{a-a_{\left(C_{t_0}\right)}}{a_{\left(C_{t_1}\right)}-a_{\left(C_{t_0}\right)}}$ ', color='blue', fontsize=20) # Label of y-axis
ax1.set_ylim(-0.05,1.05) # set limits of x-axis
ax2.tick_params(axis='y', labelcolor='blue')# set the color of the axis

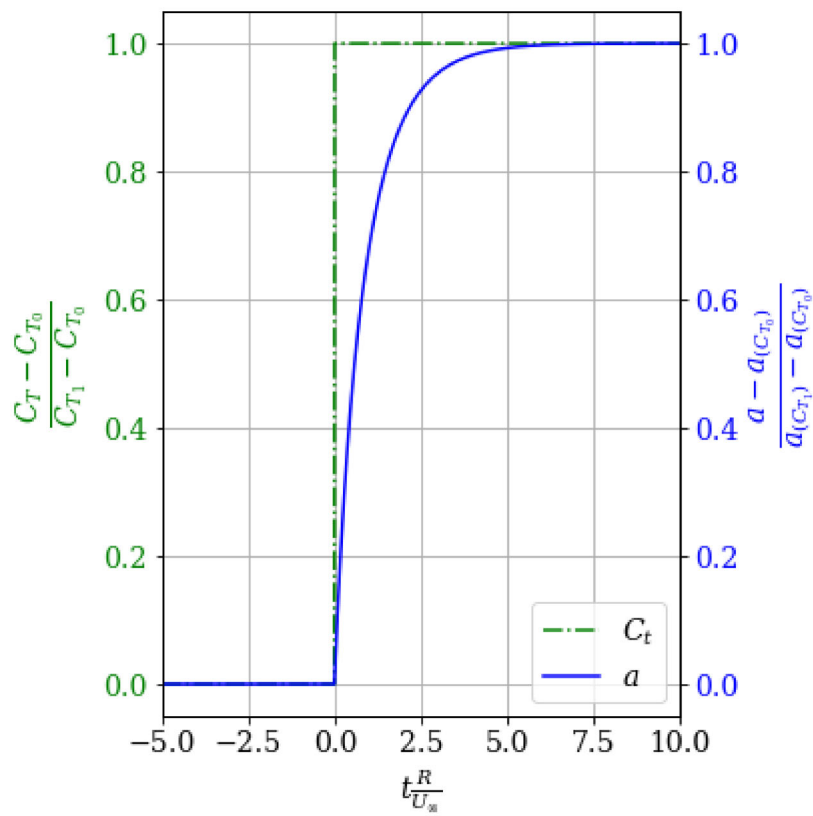
# here we plot the Legend of the figure
lns = lns1+lns2 # add legends
labs = [l.get_label() for l in lns] # extract labels
plt.legend(lns, labs, loc='lower right') # plot legend

ax1.grid(axis='both',which='both') # add a grid

plt.tight_layout() # tight layout to avoid cropping labels
plt.show() # show figure

# save figure in three formats: svg, pdf and png
filename = 'figures_tutorial_dynamic_inflow/step_change_ct_induction_pittpeters' # directory and filename
fig.savefig(filename+'.svg') # save figure in svg
fig.savefig(filename+'.pdf') # save figure in pdf
fig.savefig(filename+'.png', dpi=300) # save figure in png

```



The figure above shows the change in induction factor a calculated by the Pitt-peters model for a step change in thrust coefficient from $C_{t_0} = 0.5$ to $C_{t_0} = 0.85$. The values are made non-dimensional by the steady-solution values of a for the two values of C_t . Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ , and set to zero at the moment of the step change.

Effect of reduced frequency

We will now calculate and plot the solution of the Pitt-Peters dynamic inflow model for different ranges of reduced frequency.

```

In [6]: ## we will now plot the solution of the Pitt-Peters dynamic inflow model for differen
t ranges of reduced frequency

# we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

## we now define the frequencies and amplitudes, takign an array of reduced frequenci
es and scaling with Uinf and R
omega=np.arange(0.5,2.1,.5) *Uinf/R

# we define the variation of thrust coefficient in time
Ct0=np.array([-0.5]) # this it the average value of thrust coefficient at time

# we now define the amplitude of thrust coefficient at time
deltaCt=np.array([-0.35])

# initialize array for induction solution
vind = np.zeros([np.size(omega),np.size(time)])

# initialize array for Ct solution
Ct = np.zeros([np.size(omega),np.size(time)])

# we now initialize all time series for all frequencies
for j,omegaval in enumerate(omega):
    Ct[j,0] = Ct0
    vind[j,0] = -ainduction(-Ct0)*Uinf

# we now solve the equation in time for all frequencies,
for i,timeval in enumerate(time[:-1]):
    for j,omegaval in enumerate(omega):
        Ct[j,i+1] = Ct0 + deltaCt*np.sin(omegaval*Uinf/R*timeval) # calculate Ct at t
ime {i+1} for the case of frequency {j}
        vind[j,i+1]=pitt_peters(Ct[j,i+1],vind[j,i],Uinf,R,dt )[0] # calculate induct
ion at time {i+1} for frequency {j}

```



```

In [7]: # plot figure the change in induction factor  $a$  calculated by the Pitt-peters model
for a sinusoidal change in thrust coefficient
#  $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$ , with  $C_{T_0} = 0.5$  and  $\Delta C_T = 0.35$ .
# The values are made non-dimensioned by the steady-solution values of  $a$ 
# for the minimum and maximum values of  $C_T$ . Time  $t$  is made non-dimensional by the
# radius of the actuator  $R$  and
# the unperturbed wind speed  $U_\infty$ , and set to zero at the moment of the step change.

# plot figure
plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn')
fig1, ax1 = plt.subplots(figsize=[6,6]) # create figure, axis and define the size of the figure

# plot steady solution of induction as a function of  $C_T$ 
Ctsteady=np.arange(-(Ct0-deltaCt),-(Ct0+deltaCt)+.005,.01) # define an array of  $C_T$ 
asteady= ainduction(Ctsteady) # calculate steady solution of induction as a function of  $C_T$ 

# we plot the steady solution of induction as a function of  $C_T$ 
ax1.plot( Ctsteady, asteady, label='Steady', color='blue')

# we will now plot the unsteady solution
for j,omegaval in enumerate(omega):
    ind=(-np.floor(2*np.pi/(omegaval*R/Uinf)/dt)-1).astype(int) # indices of the last full cycle to only plot 1 cycle
    label1=r' $\omega \frac{R}{U_\infty} =$ ' + np.str(omegaval) + '$' # define label for the legend
    # plot unsteady solution
    plt1=ax1.plot(-Ct[j,ind:], -vind[j,ind:]/Uinf, label=label1, linestyle=(0,(j+1,j+1)), linewidth = (6/(j+2)) )
    color = plt1[0].get_color()

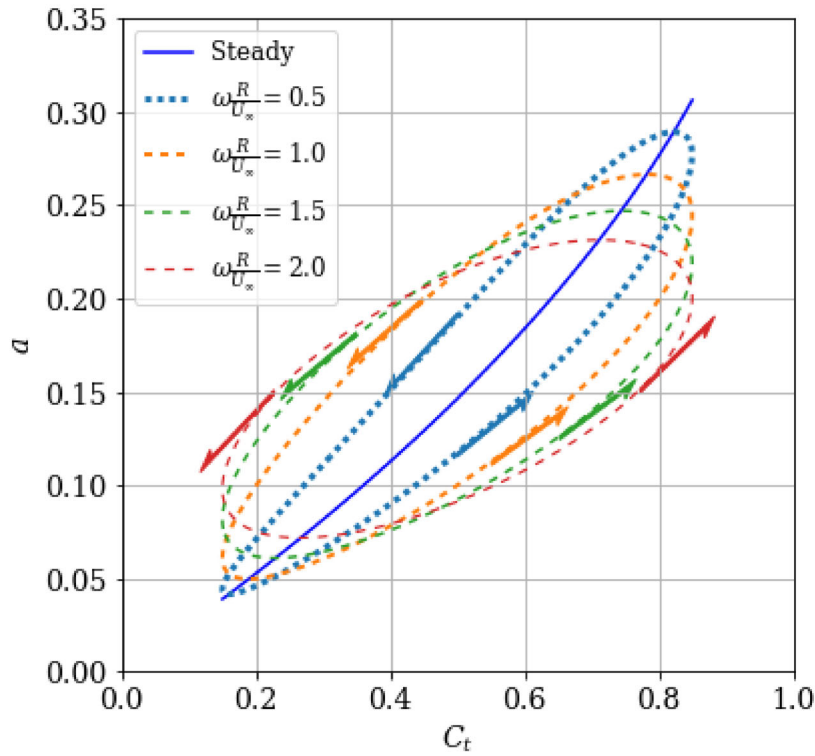
    # we will plot arrows to see the direction of the cycle
    phase_of_cycle = np.mod(time[ind:]*omegaval*R/Uinf,2*np.pi) # calculate the phase of the different points of the cycle
    i1=np.argmin(np.abs(phase_of_cycle-0))+j*30 # index of start of cycle plotted
    i2=np.argmin(np.abs(phase_of_cycle-np.pi))+j*30 # index of 180 degrees
    scale_arrow=.1 # scale of arrow
    dx = -(Ct[j,ind+i1+1]-Ct[j,ind+i1]) # dx of arrow
    dy = -(vind[j,ind+i1+1]-vind[j,ind+i1])/Uinf # dy of arrow
    ax1.arrow(-Ct[j,ind+i1], -vind[j,ind+i1]/Uinf,
              scale_arrow*dx/np.sqrt(dx**2+dy**2) , scale_arrow*dy/np.sqrt(dx**2+dy**2),
              color=color, width=scale_arrow*.04, shape='left') # plot arrow at 0 degrees of cycle
    dx = -(Ct[j,ind+i2+1]-Ct[j,ind+i2]) # dx of arrow
    dy = -(vind[j,ind+i2+1]-vind[j,ind+i2])/Uinf # dy of arrow
    ax1.arrow(-Ct[j,ind+i2], -vind[j,ind+i2]/Uinf, scale_arrow*dx/np.sqrt(dx**2+dy**2) ,
              scale_arrow*dy/np.sqrt(dx**2+dy**2),
              color=color, width=scale_arrow*.04, shape='left') # plot arrow at 190 degrees of cycle

# define properties of axis, plot grid and show figure
ax1.set_xlabel(r' $C_t$ ') # label of the x-axis
ax1.set_ylabel(r' $a$ ') # label of the y-axis
ax1.set_xlim(0,1) # set limits of x-axis
ax1.set_ylim(0,.35) # set limits of x-axis

```

```
plt.legend(fontsize=12) # plot the legend, change fontsize to fit better
plt.grid() # plot grid
plt.show() # show figure

filename = 'figures_tutorial_dynamic_inflow/sinusoidal_ct_induction_pitpetters'
fig1.savefig(filename+'.svg') # save figure
fig1.savefig(filename+'.pdf') # save figure
fig1.savefig(filename+'.png', dpi=300) # save figure
```



The figure above shows the induction factor a calculated by the Pitt-peters model for a sinusoidal change in thrust coefficient $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$, with $C_{T_0} = 0.5$ and $\Delta C_T = 0.35$. Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ . The arrows show the direction of the cycle.

The Øye dynamic inflow model

The Øye dynamic-inflow model differs from the Pitt-Peters model. Instead of expressing a linear form of the momentum equation, it expresses a filtering of the quasi-steady values through two first-order differential equations

$$\begin{aligned}
 v_{int} + \tau_1 \frac{dv_{int}}{dt} &= v_{qs} + 0.6\tau_1 \frac{dv_{qs}}{dt} \\
 v_x + \tau_2 \frac{dv_x}{dt} &= v_{int} \\
 \tau_1 &= \frac{1.1}{1 - 1.3a} \frac{R}{U_\infty} \\
 \tau_2 &= \left(\frac{r_j}{R}\right)^2 \tau_1
 \end{aligned}$$

In [8]: # The Øye dynamic inflow model

```
def oye_dynamic_inflow(vz, Ct1, Ct2, vint, Uinf, R, r,dt,glauert=False):
    # this function determines the time derivative of the induction at the annuli
    # using the Øye dynamic inflow model
    # Ct is the thrust coefficient on the actuator, vind is the induced velocity,
    # Uinf is the unperturbed velocity and R is the radial scale of the flow,
    # r is the radial position of the annulus. vqs is the quasi-steady value from BE
    M,
    #vint is an intermediate value and vz is the induced velocity

    # calculate quasi-steady induction velocity
    vqst1=-ainduction(-Ct1)*Uinf

    # calculate current induction factor
    a=-vz/Uinf

    # calculate time scales of the model
    t1 = 1.1/(1-1.3*a)*R/Uinf
    t2 = (0.39-0.26*(r/R)**2)*t1

    # calculate next-time-step quasi-steady induction velocity
    vqst2=-ainduction(-Ct2)*Uinf

    #calculate time derivative of intermediate velocity
    dvint_dt= (vqst1+ (vqst2-vqst1)/dt*0.6*t1 - vint)/t1

    # calculate new intermediate velocity
    vint2 = vint +dvint_dt*dt

    #calculate time derivaive of the induced velocity
    dvz_dt = ((vint+vint2)/2-vz)/t2

    #calculate new induced velocity
    vz2 = vz +dvz_dt*dt
    return vz2, vint2
```

```

In [9]: # we will now calculate and plot the solution of a step change in thrust coefficient
        using the Oye model

# we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

# define Ct and induction at t0
Ct0=np.array([-0.5]) # this is the value of thrust coefficient at time t0, the start
of the time array
vind0=-ainduction(-Ct0)*Uinf # this is the quasi-steady value of induction at time t
0, calculated from Ct0

# define quasi-steady solution of Ct and induction at t>=t1
Ct1=np.array([-0.85]) # this is the value of thrust coefficient at time t1
vind1=-ainduction(-Ct1)*Uinf # this is the quasi-steady value of induction at time t
1, calculated from Ct1

# define Ct as a function of time
Ct = np.zeros(np.shape(time))+Ct0 # we initialize the array of thrust coefficient, se
tting all initial values at Ct0

# change Ct for time above t1
t1=5 # we define t1, when Ct experiences a step change
Ct[time>=t1] = Ct1 # we correct the values of Ct for the instants that time is after
t1, to a value of Ct1. We therefore
        # define the step change from Ct0 to Ct1

#set arrays for induction
vind = np.zeros(np.shape(time)) # we create the array of induction velocity
vind[0]=vind0 # we initialize the first value to the induction velocity at t0, calcul
ated from Ct0

# the Oye model requires the use of a temporary value of induction vint. Here, we init
ialize the value of vint
vint=vind0

# solve the equation in time of the value of induction by using the Oye model
for i, val in enumerate(time[:-1]):
    vind[i+1], vint=oye_dynamic_inflow(vind[i], Ct[i], Ct[i+1], vint, Uinf, R, 0.95*R,
dt)

```

```

In [10]: # plot figure the change in induction factor  $a$  calculated by the Oye model for a step change in thrust coefficient
# from  $C_{t_0}=0.5$  to  $C_{t_0}=0.85$ . The values are made non-dimensioned by the steady-solution values of  $a$ 
# for the two values of  $C_{t_0}$ . Time  $t$  is made non-dimensional by the radius of the actuator  $R$  and
# the unperturbed wind speed  $U_{\infty}$ , and set to zero at the moment of the step change.

plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn') # define a colormap for the figure
fig,ax1 = plt.subplots(figsize=[6,6]) # create figure, axis and define the size of the figure
ax2=ax1.twinx() # we twin the axis to create a secondary y-axis. therefore, we twin the x-axis
# we now plot the evolution of  $C_t$  over time
lns1=ax1.plot((time-t1)*R/Uinf, (Ct-Ct0)/(Ct1-Ct0), color='green',linestyle='dashdot', label=r' $C_{t_0}$ ') # notice the negative
# value as we will be using the notation for wind turbines
# we now plot the evolution of induction velocity over time in the secondary axis
lns2=ax2.plot((time-t1)*R/Uinf, (vind-vind0)/(vind1-vind0), color='blue',linestyle='-', label= r' $a$ ') # notice the negative

# define properties of the primary y-axis
# ax1.set_aspect(aspect=20.0) # set aspect ratio of the figure
ax1.set_xlabel(r' $t \frac{R}{U_{\infty}}$ ') # Label of x-axis
ax1.set_ylabel(r' $\frac{C_{t_1}-C_{t_0}}{C_{t_1}-C_{t_0}}$ ', color='green', fontsize=20) # Label of y-axis
ax1.set_xlim(-t1,10) # set limits of x-axis
ax1.set_ylim(-0.05,1.05) # set limits of x-axis
ax1.tick_params(axis='y', labelcolor='green') # set the color of the axis

# define properties of secondary axis
ax2.set_ylabel(r' $\frac{a-a_{\left(C_{t_0}\right)}}{a_{\left(C_{t_1}\right)}-a_{\left(C_{t_0}\right)}}$ ', color='blue', fontsize=20) # Label of y-axis
ax1.set_ylim(-0.05,1.05) # set limits of x-axis
ax2.tick_params(axis='y', labelcolor='blue') # set the color of the axis

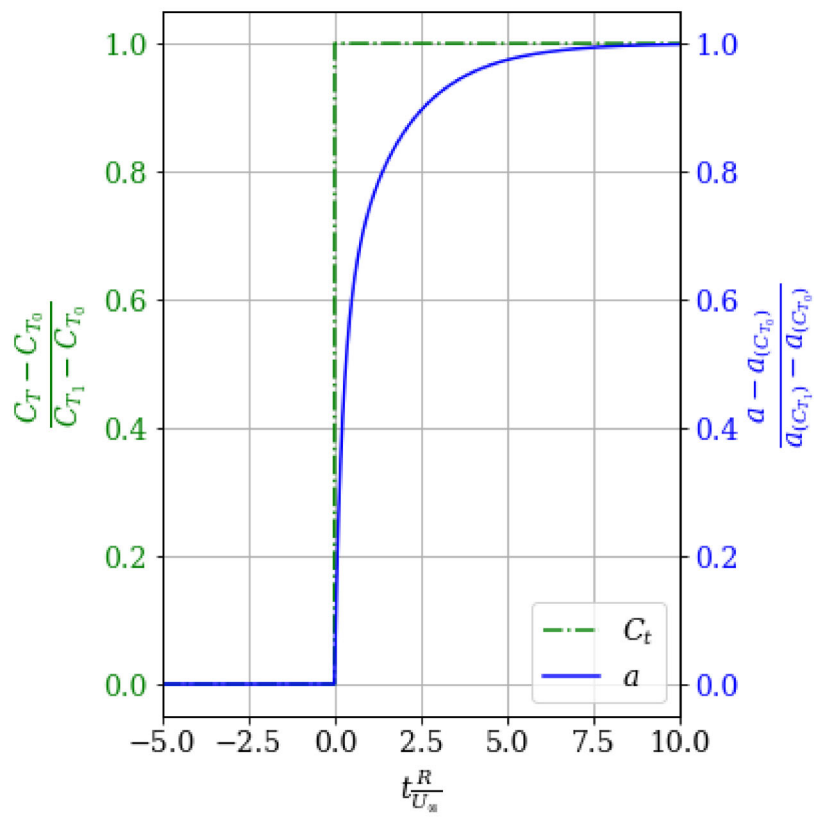
# here we plot the Legend of the figure
lns = lns1+lns2 # add legends
labs = [l.get_label() for l in lns] # extract labels
plt.legend(lns, labs, loc='lower right') # plot legend

ax1.grid(axis='both',which='both') # add a grid

plt.tight_layout() # tight layout to avoid cropping labels
plt.show() # show figure

# save figure in three formats: svg, pdf and png
filename = 'figures_tutorial_dynamic_inflow/step_change_ct_induction_oye' # directory and filename
fig.savefig(filename+'.svg') # save figure in svg
fig.savefig(filename+'.pdf') # save figure in pdf
fig.savefig(filename+'.png', dpi=300) # save figure in png

```



The figure above shows the change in induction factor a calculated by the Øye model for a step change in thrust coefficient from $C_{t_0} = 0.5$ to $C_{t_0} = 0.85$. The values are made non-dimensional by the steady-solution values of a for the two values of C_t . Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ , and set to zero at the moment of the step change.

```

In [11]: ## we will now plot the solution of the Oye dynamic inflow model for different ranges
of reduced frequency

# we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

## we now define the frequencies and amplitudes, takign an array of reduced frequenci
es and scaling with Uinf and R
omega=np.arange(0.5,2.1,.5) *Uinf/R

# we define the variation of thrust coefficient in time
Ct0=np.array([-0.5]) # this it the average value of thrust coefficient at time

# we now define the amplitude of thrust coefficient at time
deltaCt=np.array([-0.35])

# initialize array for induction solution
vind = np.zeros([np.size(omega),np.size(time)])

# initialize array for Ct solution
Ct = np.zeros([np.size(omega),np.size(time)])

# the Oye model requires the use of atemporaty value of induction vint. Here, we init
ialize the value of vint
vint = np.zeros([np.size(omega)])

# we now initialize all time series for all frequencies
for j,omegaval in enumerate(omega):
    Ct[j,0] = Ct0
    vind[j,0] = -ainduction(-Ct0)*Uinf
    vint[j] = vind[j,0]

# we now solve the equation in time for all frequencies,
for i,timeval in enumerate(time[:-1]):
    for j,omegaval in enumerate(omega):
        Ct[j,i+1] = Ct0 + deltaCt*np.sin(omegaval*Uinf/R*timeval) # calculate Ct at t
ime {i+1} for the case of frequency {j}
        vind[j,i+1],vint[j]=oye_dynamic_inflow(vind[j,i], Ct[j,i], Ct[j,i+1], vint[j
], Uinf, R, 0.95*R,dt) # calculate induction at
# time {i+1} for frequency {j}

```



```

In [12]: # plot figure the change in induction factor  $a$  calculated by the Pitt-peters model
for a sinusoidal change in thrust coefficient
#  $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$ , with  $C_{T_0} = 0.5$  and  $\Delta C_T = 0.35$ .
# The values are made non-dimensioned by the steady-solution values of  $a$ 
# for the minimum and maximum values of  $C_T$ . Time  $t$  is made non-dimensional by the
# radius of the actuator  $R$  and
# the unperturbed wind speed  $U_\infty$ , and set to zero at the moment of the step change.

# plot figure
plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn')
fig1, ax1 = plt.subplots(figsize=[6,6]) # create figure, axis and define the size of the figure

# plot steady solution of induction as a function of  $C_T$ 
Ctsteady=np.arange(-(Ct0-deltaCt),-(Ct0+deltaCt)+.005,.01) # define an array of  $C_T$ 
asteady= ainduction(Ctsteady) # calculate steady solution of induction as a function of  $C_T$ 

# we plot the steady solution of induction as a function of  $C_T$ 
ax1.plot( Ctsteady, asteady, label='Steady', color='blue')

# we will now plot the unsteady solution
for j,omegaval in enumerate(omega):
    ind=(-np.floor(2*np.pi/(omegaval*R/Uinf)/dt)-1).astype(int) # indices of the last full cycle to only plot 1 cycle
    label1=r' $\omega \frac{R}{U_\infty} =$ ' + np.str(omegaval) + '$' # define label for the legend
    # plot unsteady solution
    plt1=ax1.plot(-Ct[j,ind:], -vind[j,ind:]/Uinf, label=label1, linestyle=(0,(j+1,j+1)), linewidth = (6/(j+2)) )
    color = plt1[0].get_color()

    # we will plot arrows to see the direction of the cycle
    phase_of_cycle = np.mod(time[ind:]*omegaval*R/Uinf,2*np.pi) # calculate the phase of the different points of the cycle
    i1=np.argmin(np.abs(phase_of_cycle-0))+j*30 # index of start of cycle plotted
    i2=np.argmin(np.abs(phase_of_cycle-np.pi))+j*30 # index of 180 degrees
    scale_arrow=.1 # scale of arrow
    dx = -(Ct[j,ind+i1+1]-Ct[j,ind+i1]) # dx of arrow
    dy = -(vind[j,ind+i1+1]-vind[j,ind+i1])/Uinf # dy of arrow
    ax1.arrow(-Ct[j,ind+i1], -vind[j,ind+i1]/Uinf,
              scale_arrow*dx/np.sqrt(dx**2+dy**2) , scale_arrow*dy/np.sqrt(dx**2+dy**2),
              color=color, width=scale_arrow*.04, shape='left') # plot arrow at 0 degrees of cycle
    dx = -(Ct[j,ind+i2+1]-Ct[j,ind+i2]) # dx of arrow
    dy = -(vind[j,ind+i2+1]-vind[j,ind+i2])/Uinf # dy of arrow
    ax1.arrow(-Ct[j,ind+i2], -vind[j,ind+i2]/Uinf, scale_arrow*dx/np.sqrt(dx**2+dy**2) ,
              scale_arrow*dy/np.sqrt(dx**2+dy**2),
              color=color, width=scale_arrow*.04, shape='left') # plot arrow at 190 degrees of cycle

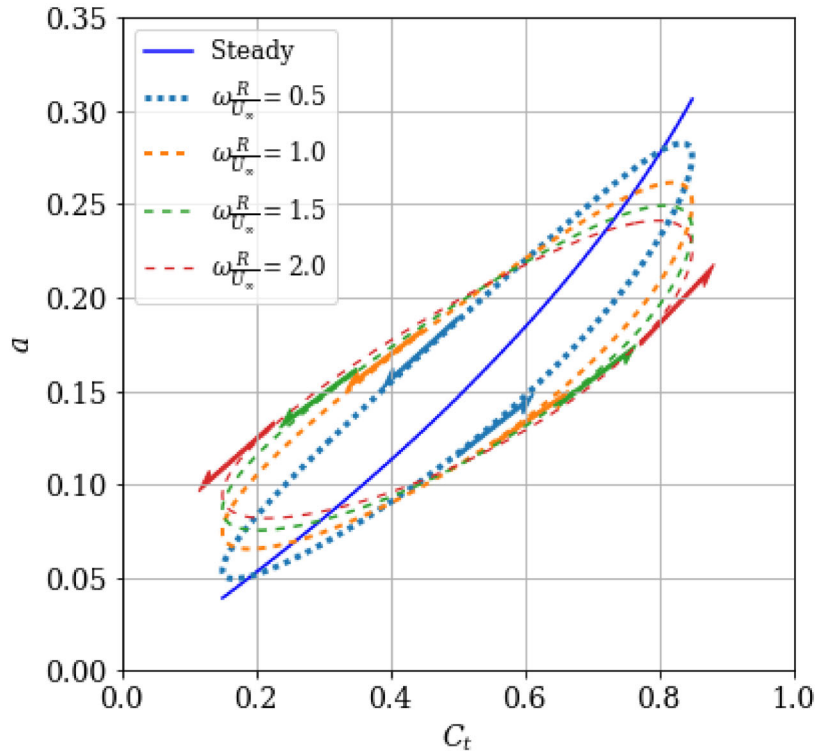
# define properties of axis, plot grid and show figure
ax1.set_xlabel(r' $C_t$ ') # label of the x-axis
ax1.set_ylabel(r' $a$ ') # label of the y-axis
ax1.set_xlim(0,1) # set limits of x-axis
ax1.set_ylim(0,.35) # set limits of x-axis

```



```
plt.legend(fontsize=12) # plot the legend, change fontsize to fit better
plt.grid() # plot grid
plt.show() # show figure

filename = 'figures_tutorial_dynamic_inflow/sinusoidal_ct_induction_oye'
fig1.savefig(filename+'.svg') # save figure
fig1.savefig(filename+'.pdf') # save figure
fig1.savefig(filename+'.png', dpi=300) # save figure
```



The figure above shows the induction factor a calculated by the Øye model for a sinusoidal change in thrust coefficient $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$, with $C_{T_0} = 0.5$ and $\Delta C_T = 0.35$. Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ . The arrows show the direction of the cycle.

The Larsen-Madsen model

In this model, as in Øye's model, dynamic inflow is modelled using a low pass filtering of the steady state induced velocities.

$$a_{t_{i+1}} = a_{t_i} e^{-\frac{\Delta t}{\tau}} + a_{qs} \left(1 - e^{-\frac{\Delta t}{\tau}}\right)$$

$$\tau = 0.5 \frac{R}{U_{wake}} \simeq 0.5 \frac{R}{a U_\infty}$$

In [13]: *# the Larsen-Madsen low pass filter model*

```
def larsenmadsen(vz, Ct2, Uinf, R,dt,glauert=False):  
    # this function determines the time derivative of the induction at the annuli  
    # using the Larsen-Madsen dynamic inflow model  
    # Ct2 is the thrust coefficient on the actuator at the next time step,  
    # vind is the induced velocity,  
    # Uinf is the unperturbed velocity and R is the radial scale of the flow,  
    # R is the radius. vqst2 is the quasi-steady value from momentum theory,  
  
    # calculate velocity wake  
    Vwake=Uinf+vz  
  
    # calculate time scales of the model  
    t1 = 0.5*R/Vwake  
  
    # calculate next-time-step quasi-steady induction velocity  
    vqst2=-ainduction(-Ct2)*Uinf  
  
    #calculate new induced velocity  
    vz2 = vz*np.exp(-dt/t1)+vqst2*(1-np.exp(-dt/t1))  
  
    return vz2
```

```

In [14]: # we will now calculate and plot the solution of a step change in thrust coefficient

# we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

# define Ct and induction at t0
Ct0=np.array([-0.5]) # this is the value of thrust coefficient at time t0, the start
of the time array
vind0=-a*induction(-Ct0)*Uinf # this is the quasi-steady value of induction at time t
0, calculated from Ct0

# define quasi-steady solution of Ct and induction at t>=t1
Ct1=np.array([-0.85]) # this is the value of thrust coefficient at time t1
vind1=-a*induction(-Ct1)*Uinf # this is the quasi-steady value of induction at time t
1, calculated from Ct1

# define Ct as a function of time
Ct = np.zeros(np.shape(time))+Ct0 # we initialize the array of thrust coefficient, se
tting all initial values at Ct0

# change Ct for time above t1
t1=5 # we define t1, when Ct experiences a step change
Ct[time>=t1] = Ct1 # we correct the values of Ct for the instants that time is after
t1, to a value of Ct1. We therefore
# define the step change from Ct0 to Ct1

#set arrays for induction
vind = np.zeros(np.shape(time)) # we create the array of induction velocity
vind[0]=vind0 # we initialize the first value to the induction velocity at t0, calcul
ated from Ct0

# solve the equation in time of the value of induction by using the Larsen-Madsen mod
el
for i,val in enumerate(time[:-1]):
    vind[i+1]=larsenmadsen(vind[i], Ct[i+1], Uinf, R,dt)

```

```

In [15]: # plot figure the change in induction factor  $a$  calculated by the Pitt-peters model
          # for a step change in thrust coefficient
          # from  $C_{t_0}=0.5$  to  $C_{t_0}=0.85$ . The values are made non-dimensioned by the steady-solution values of  $a$ 
          # for the two values of  $C_{t_0}$ . Time  $t$  is made non-dimensional by the radius of the actuator  $R$  and
          # the unperturbed wind speed  $U_{\infty}$ , and set to zero at the moment of the step change.

          plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
          plt.rcParams["font.family"] = "serif" # a nice font for the figures
          plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
          # cmap = plt.get_cmap('BuGn') # define a colormap for the figure
          fig,ax1 = plt.subplots(figsize=[6,6]) # create figure, axis and define the size of the figure
          ax2=ax1.twinx() # we twin the axis to create a secondary y-axis. therefore, we twin the x-axis
          # we now plot the evolution of  $C_t$  over time
          lns1=ax1.plot((time-t1)*R/Uinf, (Ct-Ct0)/(Ct1-Ct0), color='green',linestyle='dashdot',
            , label=r' $C_{t_0}$ ') # notice the negative
                                     # value as we will be using the notation for wind turbines
          # we now plot the evolution of induction velocity over time in the secondary axis
          lns2=ax2.plot((time-t1)*R/Uinf, (vind-vind0)/(vind1-vind0), color='blue',linestyle='-',
            , label= r' $a$ ') # notice the negative

          # define properties of the primary y-axis
          # ax1.set_aspect(aspect=20.0) # set aspect ratio of the figure
          ax1.set_xlabel(r' $t \frac{R}{U_{\infty}}$ ') # Label of x-axis
          ax1.set_ylabel(r' $\frac{C_{t_0}-C_{t_0}}{C_{t_1}-C_{t_0}}$ ',
            color='green', fontsize=20) # Label of y-axis
          ax1.set_xlim(-t1,10) # set limits of x-axis
          ax1.set_ylim(-0.05,1.05) # set limits of x-axis
          ax1.tick_params(axis='y', labelcolor='green') # set the color of the axis

          # define properties of secondary axis
          ax2.set_ylabel(r' $\frac{a-a_{\left(C_{t_0}\right)}}{a_{\left(C_{t_1}\right)}-a_{\left(C_{t_0}\right)}}$ ',
            color='blue', fontsize=20) # Label of y-axis
          ax1.set_ylim(-0.05,1.05) # set limits of x-axis
          ax2.tick_params(axis='y', labelcolor='blue')# set the color of the axis

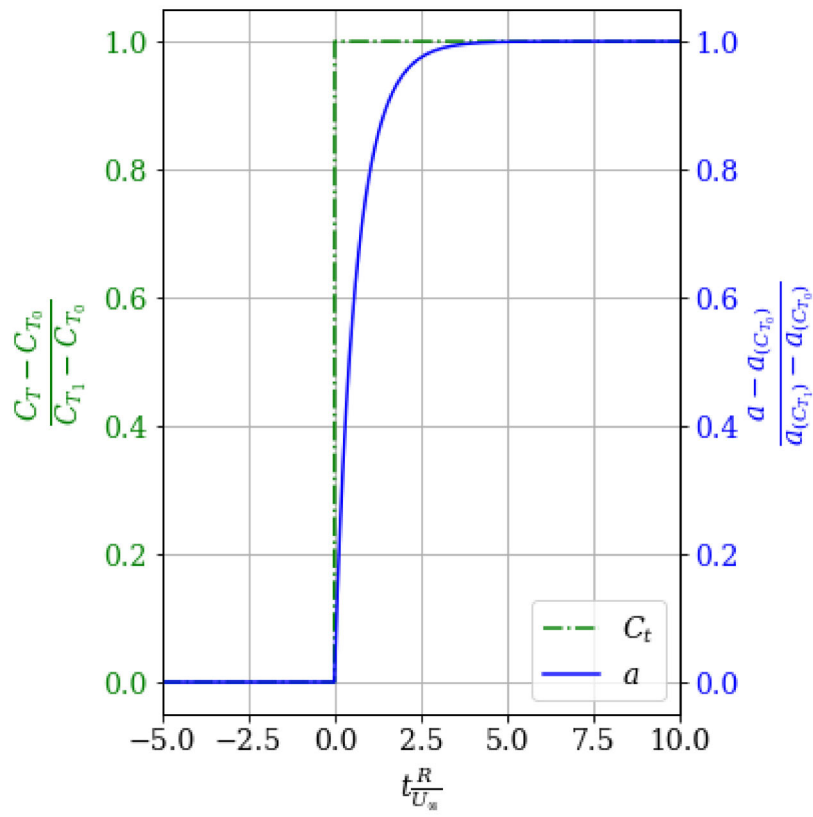
          # here we plot the Legend of the figure
          lns = lns1+lns2 # add legends
          labs = [l.get_label() for l in lns] # extract labels
          plt.legend(lns, labs, loc='lower right') # plot legend

          ax1.grid(axis='both',which='both') # add a grid

          plt.tight_layout() # tight layout to avoid cropping labels
          plt.show() # show figure

          # save figure in three formats: svg, pdf and png
          filename = 'figures_tutorial_dynamic_inflow/step_change_ct_induction_Larsen_Madsen' #
          directory and filename
          fig.savefig(filename+'.svg') # save figure in svg
          fig.savefig(filename+'.pdf') # save figure in pdf
          fig.savefig(filename+'.png', dpi=300) # save figure in png

```



The figure above shows the change in induction factor a calculated by the Larsen-Madsen model for a step change in thrust coefficient from $C_{t_0} = 0.5$ to $C_{t_0} = 0.85$. The values are made non-dimensioned by the steady-solution values of a for the two values of C_t . Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ , and set to zero at the moment of the step change.

```

In [16]: ## we will now plot the solution of the Larsen-Madsen dynamic inflow model for different ranges of reduced frequency

# we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

## we now define the frequencies and amplitudes, taken an array of reduced frequencies and scaling with Uinf and R
omega=np.arange(0.5,2.1,.5) *Uinf/R

# we define the variation of thrust coefficient in time
Ct0=np.array([-0.5]) # this is the average value of thrust coefficient at time

# we now define the amplitude of thrust coefficient at time
deltaCt=np.array([-0.35])

# initialize array for induction solution
vind = np.zeros([np.size(omega),np.size(time)])

# initialize array for Ct solution
Ct = np.zeros([np.size(omega),np.size(time)])

# we now initialize all time series for all frequencies
for j,omegaval in enumerate(omega):
    Ct[j,0] = Ct0
    vind[j,0] = -ainduction(-Ct0)*Uinf

# we now solve the equation in time for all frequencies,
for i,timeval in enumerate(time[:-1]):
    for j,omegaval in enumerate(omega):
        Ct[j,i+1] = Ct0 + deltaCt*np.sin(omegaval*Uinf/R*timeval) # calculate Ct at time {i+1} for the case of frequency {j}
        vind[j,i+1]=larsenmadsen(vind[j,i], Ct[j,i+1], Uinf, R,dt) # calculate induction at time {i+1} for frequency {j}

```

```

In [17]: # plot figure the change in induction factor  $a$  calculated by the Larsen-Madsen model for a sinusoidal change in thrust coefficient
#  $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$ , with  $C_{T_0} = 0.5$  and  $\Delta C_T = 0.35$ .
# The values are made non-dimensioned by the steady-solution values of  $a$ 
# for the minimum and maximum values of  $C_T$ . Time  $t$  is made non-dimensional by the radius of the actuator  $R$  and
# the unperturbed wind speed  $U_\infty$ , and set to zero at the moment of the step change.

# plot figure
plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn')
fig1, ax1 = plt.subplots(figsize=[6,6]) # create figure, axis and define the size of the figure

# plot steady solution of induction as a function of  $C_T$ 
Ctsteady = np.arange(-(Ct0-deltaCt), -(Ct0+deltaCt)+.005, .01) # define an array of  $C_T$ 
asteady = ainduction(Ctsteady) # calculate steady solution of induction as a function of  $C_T$ 

# we plot the steady solution of induction as a function of  $C_T$ 
ax1.plot(Ctsteady, asteady, label='Steady', color='blue')

# we will now plot the unsteady solution
for j, omegaval in enumerate(omega):
    ind = (-np.floor(2*np.pi/(omegaval*R/Uinf)/dt)-1).astype(int) # indices of the last full cycle to only plot 1 cycle
    label1 = r' $\omega \frac{R}{U_\infty} =$ ' + np.str(omegaval) + '$' # define label for the legend
    # plot unsteady solution
    plt1 = ax1.plot(-Ct[j, ind:], -vind[j, ind:]/Uinf, label=label1, linestyle=(0, (j+1, j+1)), linewidth=(6/(j+2)))
    color = plt1[0].get_color()

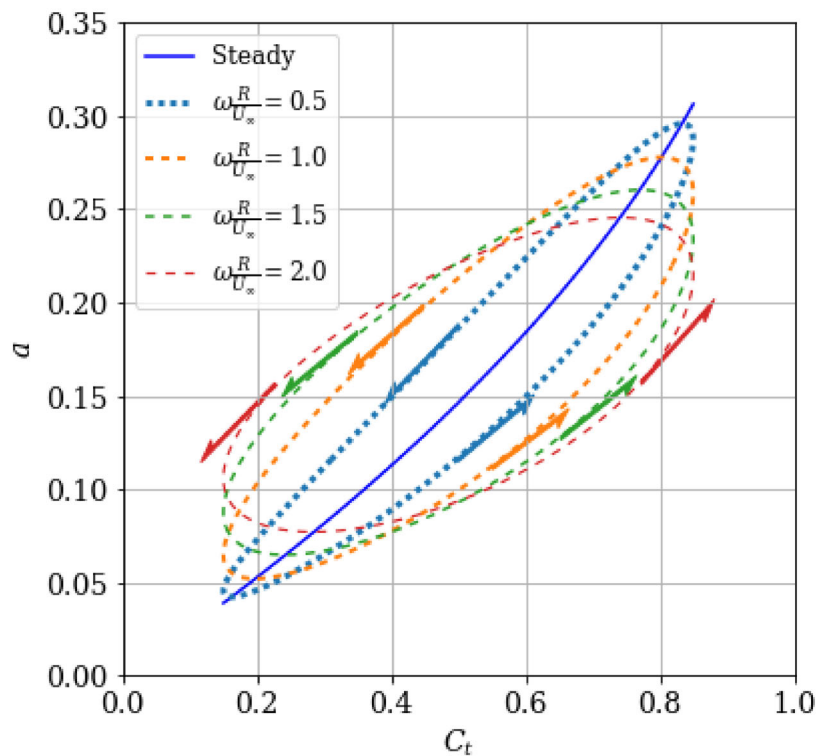
    # we will plot arrows to see the direction of the cycle
    phase_of_cycle = np.mod(time[ind:]*omegaval*R/Uinf, 2*np.pi) # calculate the phase of the different points of the cycle
    i1 = np.argmin(np.abs(phase_of_cycle-0))+j*30 # index of start of cycle plotted
    i2 = np.argmin(np.abs(phase_of_cycle-np.pi))+j*30 # index of 180 degrees
    scale_arrow = .1 # scale of arrow
    dx = -(Ct[j, ind+i1+1]-Ct[j, ind+i1]) # dx of arrow
    dy = -(vind[j, ind+i1+1]-vind[j, ind+i1])/Uinf # dy of arrow
    ax1.arrow(-Ct[j, ind+i1], -vind[j, ind+i1]/Uinf,
              scale_arrow*dx/np.sqrt(dx**2+dy**2), scale_arrow*dy/np.sqrt(dx**2+dy**2),
              color=color, width=scale_arrow*.04, shape='left') # plot arrow at 0 degrees of cycle
    dx = -(Ct[j, ind+i2+1]-Ct[j, ind+i2]) # dx of arrow
    dy = -(vind[j, ind+i2+1]-vind[j, ind+i2])/Uinf # dy of arrow
    ax1.arrow(-Ct[j, ind+i2], -vind[j, ind+i2]/Uinf, scale_arrow*dx/np.sqrt(dx**2+dy**2),
              scale_arrow*dy/np.sqrt(dx**2+dy**2),
              color=color, width=scale_arrow*.04, shape='left') # plot arrow at 190 degrees of cycle

# define properties of axis, plot grid and show figure
ax1.set_xlabel(r' $C_t$ ') # label of the x-axis
ax1.set_ylabel(r' $a$ ') # label of the y-axis
ax1.set_xlim(0,1) # set limits of x-axis
ax1.set_ylim(0,.35) # set limits of y-axis

```

```
plt.legend(fontsize=12) # plot the legend, change fontsize to fit better
plt.grid() # plot grid
plt.show() # show figure

filename = 'figures_tutorial_dynamic_inflow/sinusoidal_ct_induction_Larsen_Madsen'
fig1.savefig(filename+'.svg') # save figure
fig1.savefig(filename+'.pdf') # save figure
fig1.savefig(filename+'.png', dpi=300) # save figure
```



The figure above shows the induction factor a calculated by the Larsen-Madsen model for a sinusoidal change in thrust coefficient $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$, with $C_{T_0} = 0.5$ and $\Delta C_T = 0.35$. Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ . The arrows show the direction of the cycle.

Comparison of the three models

We will now calculate and plot the values of the three models


```

In [18]: # we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.005 # we define the time step
time=np.arange(0, 20, dt) # we create the array "time"

## we now define the frequencies and amplitudes, takign an array of reduced frequenci
es and scaling with Uinf and R
omega=np.arange(0.5,2.1,.5) *Uinf/R

# we define the variation of thrust coefficient in time
Ct0=np.array([-0.5]) # this it the average value of thrust coefficient at time

# we now define the amplitude of thrust coefficient at time
deltaCt=np.array([-0.35])

# initialize array for induction solution for the three models
vind = np.zeros([np.size(omega),np.size(time),3])

# initialize array for Ct solution
Ct = np.zeros([np.size(omega),np.size(time)])

# the Oye model requires the use of atemporaty value of induction vint. Here, we init
ialize the value of vint
vint = np.zeros([np.size(omega)])

# we now initialize all time series for all frequencies
for j,omegaval in enumerate(omega):
    Ct[j,0] = Ct0
    vind[j,0,:] = -ainduction(-Ct0)*Uinf
    vint[j] = vind[j,0,0]

# we now solve the equation in time for all frequencies,
for i,timeval in enumerate(time[:-1]):
    for j,omegaval in enumerate(omega):
        Ct[j,i+1] = Ct0 + deltaCt*np.sin(omegaval*Uinf/R*timeval) # calculate Ct at t
ime {i+1} for the case of frequency {j}
        vind[j,i+1,0]=pitt_peters(Ct[j,i+1],vind[j,i,0],Uinf,R,dt )[0] # calculate in
duction at time {i+1} for frequency {j}
        vind[j,i+1,1],vint[j]=oye_dynamic_inflow(vind[j,i,1], Ct[j,i], Ct[j,i+1], vin
t[j], Uinf, R, .95*R,dt) # calculate induction at

# time {i+1} for frequency {j}
        vind[j,i+1,2]=larsenmadsen(vind[j,i,2], Ct[j,i+1], Uinf, R,dt) # calculate in
duction at time {i+1} for frequency {j}

```

```

In [19]: # plot figure the change in induction factor  $a$  calculated by the Pitt-Peters, Oye,
Larsen-Madsen model for a sinusoidal change in thrust coefficient
#  $C_{T_0} = C_{T_0} + \Delta C_T \sin(\omega t)$ , with  $C_{T_0} = 0.5$  and  $\Delta C_T = 0.35$ .
# The values are made non-dimensioned by the steady-solution values of  $a$ 
# for the minimum and maximum values of  $C_T$ . Time  $t$  is made non-dimensional by the
radius of the actuator  $R$  and
# the unperturbed wind speed  $U_{\infty}$ , and set to zero at the moment of the step change.

# plot figure
plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn')
fig1, ax1 = plt.subplots(figsize=[8,6]) # create figure, axis and define the size of the figure

# plot steady solution of induction as a function of  $C_T$ 
Ctsteady=np.arange(-(Ct0-deltaCt),-(Ct0+deltaCt)+.005,.01) # define an array of  $C_T$ 
asteady= ainduction(Ctsteady) # calculate steady solution of induction as a function
of  $C_T$ 

# we plot the steady solution of induction as a function of  $C_T$ 
ax1.plot( Ctsteady, asteady, label='Steady', color='blue')

models = ['Pitt-P.', '\O ye', 'Larsen-M.']
# we will now plot the unsteady solution
colors = ['red', 'orange', 'green']

# for j, omegaval in enumerate(omega):
for j in [0,3]: # we will only plot two sets as nto to overload the plot
    omegaval = omega[j]
    ind=(-np.floor(2*np.pi/(omegaval*R/Uinf)/dt)-1).astype(int) # indices of the last
full cycle to only plot 1 cycle

    for k in range(3):
        label1=r'\omega \frac{R}{U_{\infty}}='+np.str(omegaval)+'$ $'+models[k]+'$' #
define label for the legend
        # plot unsteady solution
        ax1.plot(-Ct[j,ind:], -vind[j,ind:,k]/Uinf, label=label1,
linestyle=(0,(j+1,j+1)), linewidth = (6/(j+2)) , color=colors[k])

        # we will plot arrows to see the direction of the cycle
        phase_of_cycle = np.mod(time[ind:]*omegaval*R/Uinf,2*np.pi) # calculate the phase
of the different points of the cycle
        i1=np.argmin(np.abs(phase_of_cycle-0))+j*30 # index of start of cycle plotted
        i2=np.argmin(np.abs(phase_of_cycle-np.pi))+j*30 # index of 180 degrees
        scale_arrow=.1 # scale of arrow
        dx = -(Ct[j,ind+i1+1]-Ct[j,ind+i1]) # dx of arrow
        dy = -(vind[j,ind+i1+1,k]-vind[j,ind+i1,k])/Uinf # dy of arrow
        ax1.arrow(-Ct[j,ind+i1], -vind[j,ind+i1,k]/Uinf,
scale_arrow*dx/np.sqrt(dx**2+dy**2) , scale_arrow*dy/np.sqrt(dx**2+
dy**2),
color=colors[k], width=scale_arrow*.04, shape='left') # plot arrow
at 0 degrees of cycle
        dx = -(Ct[j,ind+i2+1]-Ct[j,ind+i2]) # dx of arrow
        dy = -(vind[j,ind+i2+1,k]-vind[j,ind+i2,k])/Uinf # dy of arrow
        ax1.arrow(-Ct[j,ind+i2], -vind[j,ind+i2,k]/Uinf, scale_arrow*dx/np.sqrt(dx**2
+dy**2) ,
scale_arrow*dy/np.sqrt(dx**2+dy**2),
color=colors[k], width=scale_arrow*.04, shape='left') # plot arrow

```

at 190 degrees of cycle

```
# define properties of axis, plot grid and show figure
```

```
ax1.set_xlabel(r'$C_t$') # label of the x-axis
```

```
ax1.set_ylabel(r'$a$') # label of the y-axis
```

```
ax1.set_xlim(0,1.5) # set limits of x-axis
```

```
ax1.set_ylim(0,.35) # set limits of y-axis
```

```
plt.legend(fontsize=12, loc='lower right') # plot the legend, change fontsize to fit better
```

```
plt.grid() # plot grid
```

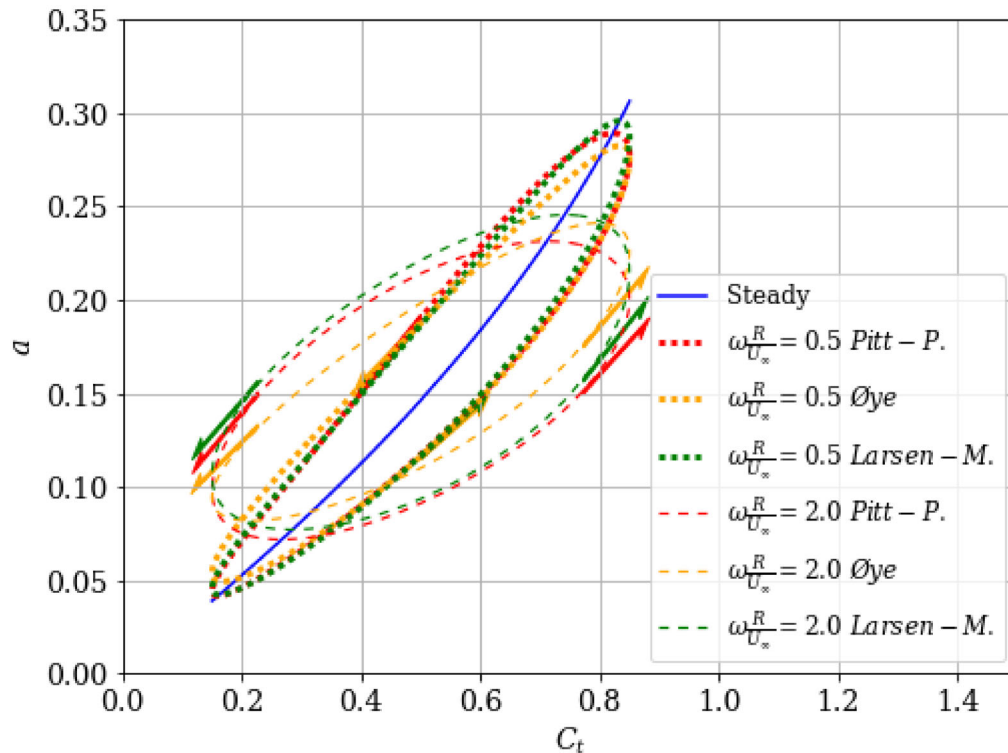
```
plt.show() # show figure
```

```
filename = 'figures_tutorial_dynamic_inflow/sinusoidal_ct_induction_PP_Oye_LM'
```

```
fig1.savefig(filename+'.svg') # save figure
```

```
fig1.savefig(filename+'.pdf') # save figure
```

```
fig1.savefig(filename+'.png', dpi=300) # save figure
```



The figure above shows the induction factor a calculated by the Pitt-Peters, Øye and Larsen-Madsen models for a sinusoidal change in thrust coefficient $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$, with $C_{T_0} = 0.5$ and $\Delta C_T = 0.35$. Time t is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ . The arrows show the direction of the cycle.

Calculate the amplitude damping and the phase delay for all models for different frequencies

Amplitude damping and the phase delay are a function of the reduced frequency of the perturbation. We will now calculate these values for the different models for different reduced frequencies

```

In [20]: # we define the value of U_infinity and the radius of the actuator
Uinf=1 # U_infinity
R=1 # radius of the actuator

# define time array
dt=0.01 # we define the time step
time=np.arange(0, 200, dt) # we create the array "time"

## we now define the frequencies and amplitudes, takign an array of reduced frequenci
es and scaling with Uinf and R
omega=np.arange(0.05,2.51,.25) *Uinf/R

# we define the variation of thrust coefficient in time
Ct0=np.array([-0.5]) # this it the average value of thrust coefficient at time

# we now define the amplitude of thrust coefficient at time
deltaCt=np.array([-0.35])

# initialize array for induction solution for the three models
vind = np.zeros([np.size(omega),np.size(time),3])

# initialize array for Ct solution
Ct = np.zeros([np.size(omega),np.size(time)])

# the Oye model requires the use of atemporaty value of induction vint. Here, we init
ialize the value of vint
vint = np.zeros([np.size(omega)])

# we now initialize all time series for all frequencies
for j,omegaval in enumerate(omega):
    Ct[j,0] = Ct0
    vind[j,0,:] = -ainduction(-Ct0)*Uinf
    vint[j] = vind[j,0,0]

# we now solve the equation in time for all frequencies,
for i,timeval in enumerate(time[:-1]):
    for j,omegaval in enumerate(omega):
        Ct[j,i+1] = Ct0 + deltaCt*np.sin(omegaval*Uinf/R*timeval) # calculate Ct at t
ime {i+1} for the case of frequency {j}
        vind[j,i+1,0]=pitt_peters(Ct[j,i+1],vind[j,i,0],Uinf,R,dt )[0] # calculate in
duction at time {i+1} for frequency {j}
        vind[j,i+1,1],vint[j]=oye_dynamic_inflow(vind[j,i,1], Ct[j,i], Ct[j,i+1], vin
t[j], Uinf, R, .5*R,dt) # calculate induction at

# time {i+1} for frequency {j}
        vind[j,i+1,2]=larsenmadsen(vind[j,i,2], Ct[j,i+1], Uinf, R,dt) # calculate in
duction at time {i+1} for frequency {j}

```

```

In [21]: # plot steady solution of induction as a function of  $C_T$ 
Ctsteady=np.arange(-(Ct0-deltaCt),-(Ct0+deltaCt)+.005,.01) # define an array of  $C_T$ 
asteady= ainduction(Ctsteady) # calculate steady solution of induction as a function
of  $C_T$ 

# steady amplitude
amplstd=asteady[-1]-asteady[0]

# we now calculate the amplitude and phase delay for each case

# initialize array for amplitude
ampl = np.zeros([np.size(omega),3])

# initialize array for phase_delay
phase_delay = np.zeros([np.size(omega),3])

for j,omegaval in enumerate(omega): # for-cycle over all frequencies
    ind=(-np.floor(2*np.pi/(omegaval*R/Uinf)/dt)-1).astype(int) # indices of the last
    full cycle to only plot 1 cycle

        for k in range(3):
            # we will plot arrows to see the direction of the cycle
            atemp = -vind[j,ind:,k]/Uinf # create temporary array of induction factor for
            the cycle
            phase_of_cycle = np.mod(time[ind:]*omegaval*R/Uinf,2*np.pi) # calculate the p
            hase of the different points of the cycle

            i1=np.argmax(atemp) # index of start of cycle plotted
            i2=np.argmin(atemp) # index of 180 degrees

            ampl[j,k]=atemp[i1]-atemp[i2] # determine 2*amplitude of induction factor
            phase_delay[j,k] = (phase_of_cycle[i1]-np.pi/2)*180/np.pi # determine phase d
            elay at maximum induction

```

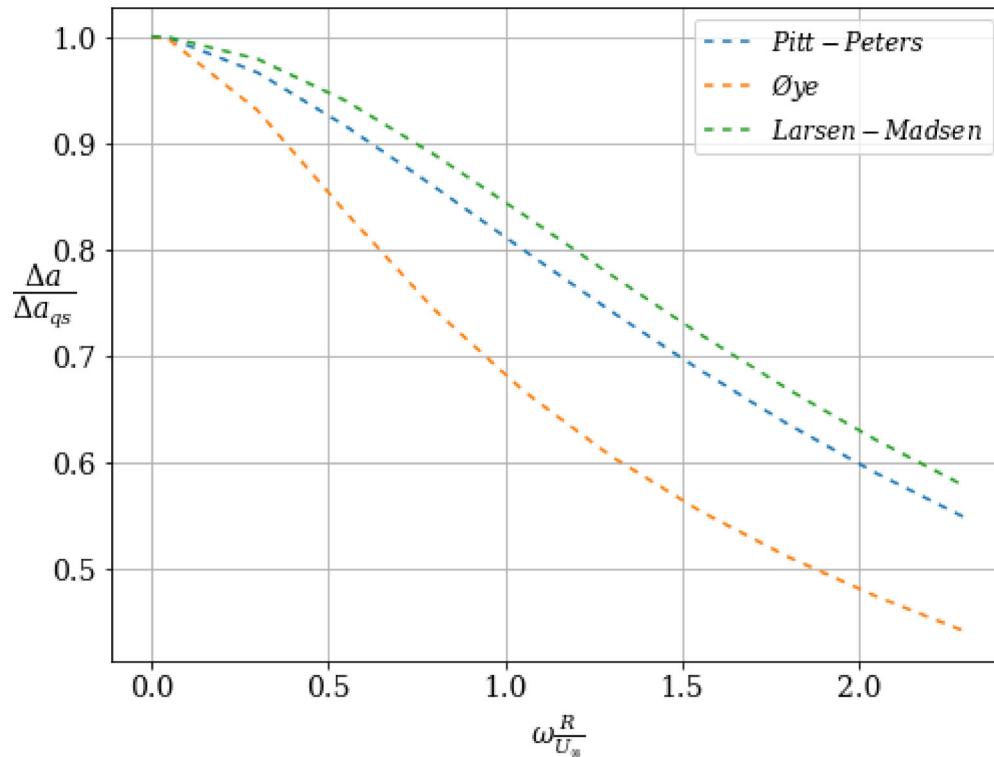
```
In [22]: models = ['Pitt-Peters', '\O ye', 'Larsen-Madsen']

# plot figure
plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn')
fig1, ax1 = plt.subplots(figsize=[8,6]) # create figure, axis and define the size of the figure

for i in range(3):
    label1=r'$'+models[i]+'$' # define label for the legend
    # we plot the steady solution of induction as a function of $C_T$
    ax1.plot( np.append(0, omega), np.append(1, ampl[:,i]/amplstd), label=label1, linestyle=(0,(k+1,k+1)), linewidth = (6/(k+2)))

# define properties of axis, plot grid and show figure
ax1.set_xlabel(r'$\omega \frac{R}{U_{\infty}}$') # Label of the x-axis
ax1.set_ylabel(r'$\frac{\Delta a}{\Delta a_{qs}}$', rotation=0, fontsize=20) # Label of the y-axis
# ax1.set_xlim(0,1.5) # set limits of x-axis
# ax1.set_ylim(0,.35) # set limits of x-axis
plt.legend(fontsize=12, loc='upper right') # plot the legend, change fontsize to fit better
plt.grid() # plot grid
plt.show() # show figure

filename = 'figures_tutorial_dynamic_inflow/sinusoidal_amplitude_PP_Oye_LM'
fig1.savefig(filename+'.svg') # save figure
fig1.savefig(filename+'.pdf') # save figure
fig1.savefig(filename+'.png', dpi=300) # save figure
```



The figure above shows the change in amplitude of induction factor Δa as a function of the frequency of the sinusoidal variation of the thrust coefficient, as calculated by the Pitt-Peters, Øye and Larsen-Madsen models for a sinusoidal change in thrust coefficient $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$, with $C_{T_0} = 0.5$ and $\Delta C_T = 0.35$. Δa is made non-dimensional by the quasi-steady value. The frequency ω is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_{∞} .

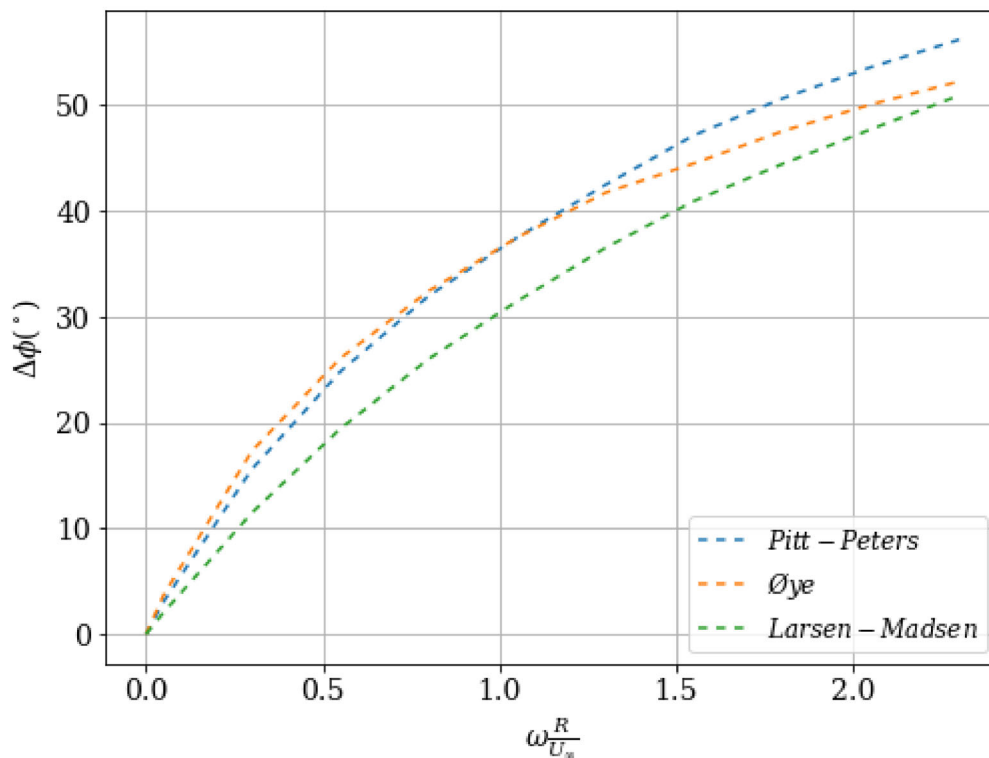
```
In [23]: models = ['Pitt-Peters', '\O ye', 'Larsen-Madsen']

# plot figure
plt.rcParams.update({'font.size': 14}) # define fontsize for the figures
plt.rcParams["font.family"] = "serif" # a nice font for the figures
plt.rcParams["mathtext.fontset"] = "dejavuserif" # a nice font for the latex expressions
# cmap = plt.get_cmap('BuGn')
fig1, ax1 = plt.subplots(figsize=[8,6]) # create figure, axis and define the size of the figure

for i in range(3):
    label1=r'$'+models[i]+'$' # define label for the legend
    # we plot the steady solution of induction as a function of $C_T$
    ax1.plot( np.append(0, omega), np.append(0, phase_delay[:,i]), label=label1, linestyle=(0,(k+1,k+1)), linewidth = (6/(k+2)))

# define properties of axis, plot grid and show figure
ax1.set_xlabel(r'$\omega \frac{R}{U_\infty}$') # Label of the x-axis
ax1.set_ylabel(r'$\Delta \phi$ (^\circ)$') # Label of the y-axis
# ax1.set_xlim(0,1.5) # set limits of x-axis
# ax1.set_ylim(0,.35) # set limits of x-axis
plt.legend(fontsize=12, loc='lower right') # plot the legend, change fontsize to fit better
plt.grid() # plot grid
plt.show() # show figure

filename = 'figures_tutorial_dynamic_inflow/sinusoidal_phase_delay_PP_Oye_LM'
fig1.savefig(filename+'.svg') # save figure
fig1.savefig(filename+'.pdf') # save figure
fig1.savefig(filename+'.png', dpi=300) # save figure
```



The figure above shows the phase delay of induction factor Δa (in degrees) as a function of the frequency of the sinusoidal variation of the thrust coefficient, as calculated by the Pitt-Peters, Øye and Larsen-Madsen models for a sinusoidal change in thrust coefficient $C_T = C_{T_0} + \Delta C_T \sin(\omega t)$, with $C_{T_0} = 0.5$ and $\Delta C_T = 0.35$. The frequency ω is made non-dimensional by the radius of the actuator R and the unperturbed wind speed U_∞ .

In []: