5. Newton's law of cooling a)  $d T(t) = -x (T(t) - T_e)$ 3 dtd) = 5 - 16 dt [en (TH)-Te)] = [-x+]  $\ln \left( \frac{T(t) - Te}{T(0) - Te} \right) = -kt$   $\frac{T(t) - Te}{T(0) - Te} = e$  $T(t) = Te + (T(0) - Te)e^{-\kappa t}$   $b) (1) dT(t) = -\kappa (T(t) - Te)$ FF: (2) dT(A) = T(++1)-T(A) (1)=(2) - Ku(t) = T(t+st) - T(t)- Kult) = u(+st) -u(+) a (test) = u(t) (1- Kst) re substituition (T Const) = Te + (To -Te) (1- KSt) m u(mst) = u(0) (1- Kst)m

c) T(mst) = Te + (To - Te) (1- HS+) m Stability and. 11- Kst 151 052- Kst 52 - 0 Kst 70 1 Kst 52 - conditionally stable! of) see at the Cost part of the poly - X u(++st) = T(++st) - T(+) - K u (++st) = u(++st)-u(+) u(t) = u(t+st) (1+ Kst) u(t+st) = u(t) 1+ Kst u(mst) = u(0)  $(1 + Kst)^{m}$ resub. T (mst) = Te + T(0)-Te (1+ Kst) m stability and. (1+ Ks+ 171 = always fullfilled - I unconditionally stable both positive

f) Taylor expansion  $f(t+st) = f(t) + \frac{d}{dt} f(t) st + \frac{1}{2} \frac{d^2}{dt^2} f(t) st^2 + O(st^3)$   $f(t) = E + (T_0 - T_e)e^{-Kt}$ of f(+) = (To-Te) (-K)e-Kt d2 p(t) = (To-Te) K2 e-Kt Pr (+) = (To-Te)e (1-Ks+)+ Te Pr (+) = (To-Te)e Kt (1-Ks++ K3t2)+ Te As we see in the plot Pr(t) is slightly closer to the analytical solution (higher order = higher accuracy) 6. a) f(N(+)) = 1(++++) - N(+) NG) - NH)+SNA) - N(++3+)+SNA+)-NH)-SNA) f(N(+)+SN(+))= N(++s+)-N(+)+SN(++s+)-SN(+)
= F(N(+)) f (N(+) + SN(+)) - f(N(+)) - SN(+) - SN(+) st f(NA)+SN(H))-f(N(H)) = 8N(H+sH) - 1

N(++st) = N(+) (st df +1) Stability and 14 st 2N 1+ 1+ 21 <1 A+2N <0 N < 0 (2) -1-S+ZN (1 -St2N <2 N > - 1 sf(W) /N° (should be a parabola)

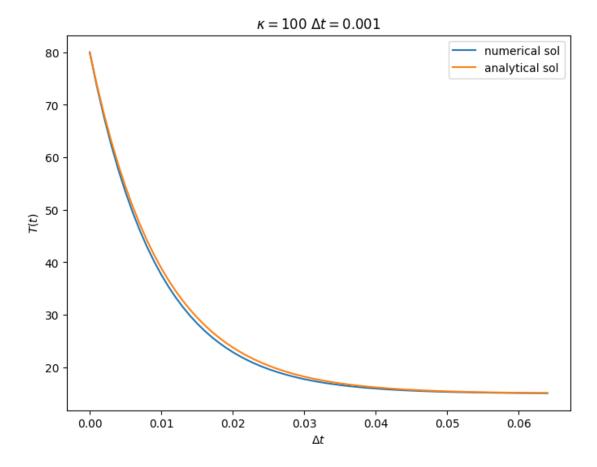
## sheet4

## September 21, 2017

```
1 5.
1.1 d)
In [1]: using PyPlot
In [2]: #ef marching eq
         ef(x,t0,te,dt,k)=te+(t0-te)*(1.-k*dt)^x
         #analytical sol
         analytical(x,t0,te,k)=te+(t0-te)*exp(-k*x)
Out[2]: analytical (generic function with 1 method)
   \kappa \Delta t \geq 0 \wedge \kappa \Delta \leq 2
   e.g. \rightarrow \kappa = 100 \land \Delta t = 0.001
In [3]: #parameters
         k=100.
         dt=0.001
         t0=80
         te=15
         tend=te+0.1
         tvalues=Float64[]
         xvalues=Float64[]
         anavalues=Float64[]
         #x-step, corresponds to m
         x=0.
         #time out of step
         t(x)=x*dt
         #numerical temp
         tnow=ef(x,t0,te,dt,k)
         #analytical
         ana=analytical(x,t0,te,k)
         println(ana)
         #until the analytical is aboth the wished
         while(ana>=tend)
             push!(tvalues,tnow)
             push!(anavalues,ana)
```

```
push!(xvalues,x)
#increases timestep
x+=1
    tnow=ef(x,t0,te,dt,k)
    ana=analytical(t(x),t0,te,k)
end

#plot
figure(1,figsize=(8,6))
plot(xvalues*dt,tvalues,label="numerical sol")
plot(xvalues*dt,anavalues,label="analytical sol")
title(L"$\kappa = 100$ $\Delta t=0.001$")
ylabel(L"$T(t)$")
xlabel(L"$\Delta t$")
legend()
```

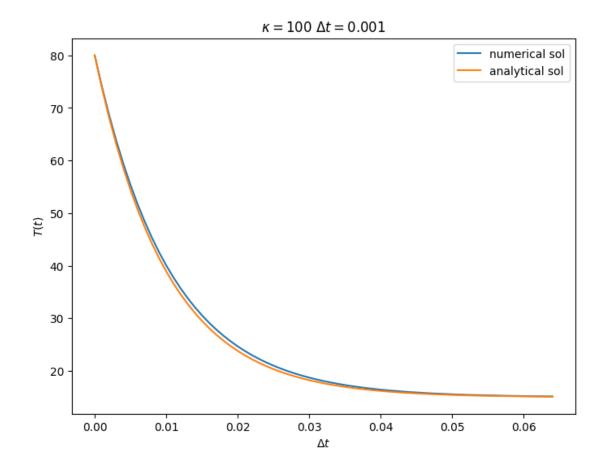


80.0

Out[3]: PyObject <matplotlib.legend.Legend object at 0x7f4729944890>

## 1.2 e)

```
In [4]: #eb-scheme
        eb(x,t0,te,dt,k)=te+(t0-te)/(1+k*dt)^x
        #parameters
        k=100.
        dt=0.001
        t0=80
        te=15
        tend=te+0.1
        tvalues=Float64[]
        xvalues=Float64[]
        anavalues=Float64[]
        #x-step, corresponds to m
        x=0.
        t(x)=x*dt
        tnow=eb(x,t0,te,dt,k)
        ana=analytical(x,t0,te,k)
        println(ana)
        #until the analytical is aboth the wished
        while(ana>=tend)
            push!(tvalues,tnow)
            push!(anavalues,ana)
            push!(xvalues,x)
            x+=1
            tnow=eb(x,t0,te,dt,k)
            ana=analytical(t(x),t0,te,k)
        end
        #plot
        figure(1,figsize=(8,6))
        plot(xvalues*dt,tvalues,label="numerical sol")
        plot(xvalues*dt,anavalues,label="analytical sol")
        title(L"$\kappa = 100$ $\Delta t=0.001$")
        ylabel(L"$T(t)$")
        xlabel(L"$\Delta t$")
        legend()
```



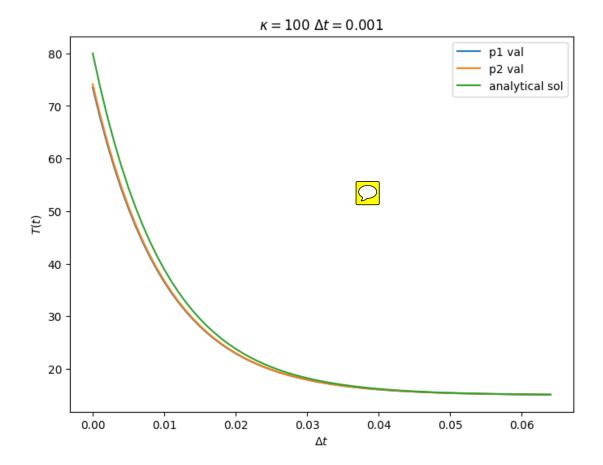
80.0

WARNING: Method definition t(Any) in module Main at In[3]:13 overwritten at In[4]:15.

Out[4]: PyObject <matplotlib.legend.Legend object at 0x7f47297c4e50>

## 1.3 f)

```
#values
xvalues=Float64[]
p1values=Float64[]
p2values=Float64[]
anavalues=Float64[]
x=0.
t(x)=x*dt
t1=p1(t(x),t0,te,dt,k)
t2=p2(t(x),t0,te,dt,k)
ana=analytical(x,t0,te,k)
println(ana)
while(ana>=tend)
    push!(p1values,t1)
    push!(p2values,t2)
    push!(anavalues,ana)
    push!(xvalues,x)
    x+=1
    t1=p1(t(x),t0,te,dt,k)
    t2=p2(t(x),t0,te,dt,k)
    ana=analytical(t(x),t0,te,k)
end
#plot
figure(1,figsize=(8,6))
plot(xvalues*dt,p1values,label="p1 val")
plot(xvalues*dt,p2values,label="p2 val")
plot(xvalues*dt,anavalues,label="analytical sol")
title(L"\$\appa = 100\$ \$\Delta t=0.001\$")
ylabel(L"$T(t)$")
xlabel(L"$\Delta t$")
legend()
```

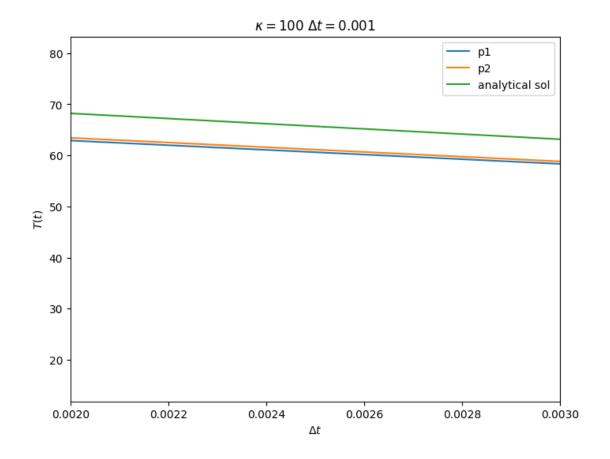


80.0

WARNING: Method definition t(Any) in module Main at In[4]:15 overwritten at In[5]:17.

Out[5]: PyObject <matplotlib.legend.Legend object at 0x7f472971c990>

I want to see, which polynomial is closer to the analytical result, therefore i zoom in a bit



Out[12]: (0.002,0.003)

In []: