5.3 Transient Heat Conduction in Various Geometries

Rescaling

We'll assume the thermal conductivity k is a constant, and define thermal diffusivity in the conventional way

$$lpha = rac{k}{
ho C_p}$$

We will further assume symmetry with respect to all spatial coordinates except r where r extends from -R to +R. The boundary conditions are

$$T(t,R) = T_{\infty} \qquad orall t > 0 \
abla T(t,0) = 0 \qquad orall t \geq 0$$

where we have assumed symmetry with respect to r and uniform initial conditions $T(0,r)=T_0$ for all $0 \le r \le R$. Following standard scaling procedures, we introduce the dimensionless variables

$$T'=rac{T-T_0}{T_\infty-T_0} \ r'=rac{r}{R} \ t'=trac{lpha}{R^2}$$

Dimensionless Model

Under these conditions the problem reduces to

$$rac{\partial T'}{\partial t'} =
abla^2 T'$$

with auxiliary conditions

$$T'(0,r') = 0 \qquad \forall 0 \le r' \le 1 \ T'(t',1) = 1 \qquad \forall t' > 0 \
abla T'(t',0) = 0 \qquad \forall t' \ge 0$$

which we can specialize to specific geometries.

Preliminary Code

```
In [ ]:
```

```
!pip install -q pyomo
!wget -N -q "https://ampl.com/dl/open/ipopt/ipopt-linux64.zip"
!unzip -o -q ipopt-linux64
ipopt_executable='/content/ipopt'
```

In []:

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d.axes3d import Axes3D
def model_plot(m):
    r = sorted(m.r)
    t = sorted(m.t)
   rgrid = np.zeros((len(t), len(r)))
    tgrid = np.zeros((len(t), len(r)))
    Tgrid = np.zeros((len(t), len(r)))
    for i in range(0, len(t)):
       for j in range(0, len(r)):
            rgrid[i,j] = r[j]
            tgrid[i,j] = t[i]
            Tgrid[i,j] = m.T[t[i], r[j]].value
   fig = plt.figure(figsize=(10,6))
    ax = fig.add subplot(1, 1, 1, projection='3d')
    ax.set_xlabel('Distance r')
    ax.set_ylabel('Time t')
    ax.set zlabel('Temperature T')
    p = ax.plot_wireframe(rgrid, tgrid, Tgrid)
```

Planar Coordinates

Suppressing the prime notation, for a slab geometry the model specializes to

$$\frac{\partial T}{\partial t} = \frac{\partial^2 T}{\partial r^2}$$

with auxiliary conditions

$$egin{array}{ll} T(0,r) &= 0 & \forall 0 \leq r \leq 1 \\ T(t,1) &= 1 & \forall t > 0 \\ rac{\partial T}{\partial r}(t,0) &= 0 & \forall t \geq 0 \end{array}$$

```
In [11]:
```

```
!ls -a /content
```

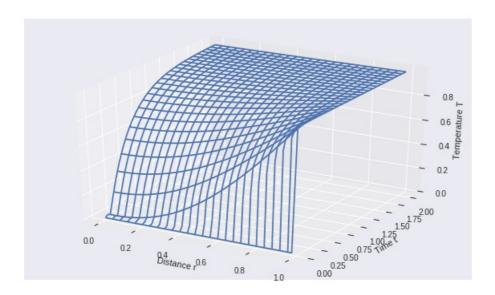
. .. coin-license.txt .config ipopt ipopt-linux64.zip sample_data

In [12]:

```
from pyomo.environ import *
from pyomo.dae import *
m = ConcreteModel()
m.r = ContinuousSet(bounds=(0,1))
m.t = ContinuousSet(bounds=(0,2))
m.T = Var(m.t, m.r)
m.dTdt = DerivativeVar(m.T, wrt=m.t)
        = DerivativeVar(m.T, wrt=m.r)
m.d2Tdr2 = DerivativeVar(m.T, wrt=(m.r, m.r))
@m.Constraint(m.t, m.r)
def pde(m, t, r):
    if t == 0:
        return Constraint.Skip
    if r == 0 or r == 1:
       return Constraint.Skip
    return m.dTdt[t,r] == m.d2Tdr2[t,r]
m.obj = Objective(expr=1)
m.ic = Constraint(m.r, rule=lambda m, r:
                                            m.T[0,r] == 0 if r > 0 and r < 1 else
Constraint.Skip)
m.bc1 = Constraint(m.t, rule=lambda m, t:
                                           m.T[t,1] == 1)
m.bc2 = Constraint(m.t, rule=lambda m, t: m.dTdr[t,0] == 0)
TransformationFactory('dae.finite_difference').apply_to(m, nfe=50, scheme='FORWARD', wr
TransformationFactory('dae.finite_difference').apply_to(m, nfe=50, scheme='FORWARD', wr
SolverFactory('ipopt', executable=ipopt_executable).solve(m, tee=True).write()
model_plot(m)
Ipopt 3.12.8:
This program contains Ipopt, a library for large-scale nonlinear optimization.
Ipopt is released as open source code under the Eclipse Public License (EPL).
        For more information visit http://projects.coin-or.org/Ipopt
*************************
This is Ipopt version 3.12.8, running with linear solver mumps.
NOTE: Other linear solvers might be more efficient (see Ipopt documentation).
Number of nonzeros in equality constraint Jacobian...:
                                                           30347
Number of nonzeros in inequality constraint Jacobian.:
                                                              0
Number of nonzeros in Lagrangian Hessian....:
Total number of variables....:
                                                           10299
                     variables with only lower bounds:
                variables with lower and upper bounds:
                                                               0
                    variables with only upper bounds:
                                                              0
Total number of equality constraints....:
                                                           10200
Total number of inequality constraints....:
                                                              0
        inequality constraints with only lower bounds:
                                                               0
   inequality constraints with lower and upper bounds:
                                                               0
        inequality constraints with only upper bounds:
                             inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
iter
        objective
                     inf pr
   0 1.0000000e+00 1.00e+00 0.00e+00 -1.0 0.00e+00 - 0.00e+00 0.00e+00 0 1.0000000e+00 1.50e-12 2.50e-01 -1.7 2.50e+03 -4.0 1.00e+00 1.00e+00h 1 2 1.0000000e+00 1.82e-12 1.53e-10 -1.7 8.16e-10 -4.5 1.00e+00 1.00e+00h 1
```

```
Number of iterations...: 2
                              (scaled)
                                                   (unscaled)
Objective....: 1.000000000000000e+00 1.000000000000000e+00
Dual infeasibility....: 1.5268200213081659e-10 1.5268200213081659e-10
Constraint violation...: 3.6364522504328498e-14
Complementarity.....: 0.000000000000000e+00
Overall NLP error....: 1.5268200213081659e-10
                                              1.8182261252164267e-12
                                              0.0000000000000000e+00
                                              1.5268200213081659e-10
Number of objective function evaluations
                                             = 3
Number of objective gradient evaluations
Number of equality constraint evaluations
Number of inequality constraint evaluations
Number of equality constraint Jacobian evaluations = 3
Number of inequality constraint Jacobian evaluations = 0
Number of Lagrangian Hessian evaluations
                                             = 2
Total CPU secs in IPOPT (w/o function evaluations) =
                                                   0.207
Total CPU secs in NLP function evaluations
                                                   0.003
EXIT: Optimal Solution Found.
# = Solver Results
# ______
  Problem Information
# _______
Problem:
- Lower bound: -inf
 Upper bound: inf
 Number of objectives: 1
 Number of constraints: 10200
 Number of variables: 10299
 Sense: unknown
# -----
  Solver Information
# -----
Solver:
- Status: ok
 Message: Ipopt 3.12.8\x3a Optimal Solution Found
 Termination condition: optimal
 Id: 0
 Error rc: 0
 Time: 0.30750346183776855
  Solution Information
- number of solutions: 0
```

number of solutions displayed: 0



Cylindrical Coordinates

Suppressing the prime notation, for a cylindrical geometry the model specializes to

$$\frac{\partial T}{\partial t} = \frac{1}{r} \frac{\partial}{\partial r} \left(r \frac{\partial T}{\partial r} \right)$$

Expanding,

$$\frac{\partial T}{\partial t} = \frac{\partial^2 T}{\partial t^2} + \frac{1}{r} \frac{\partial T}{\partial r}$$

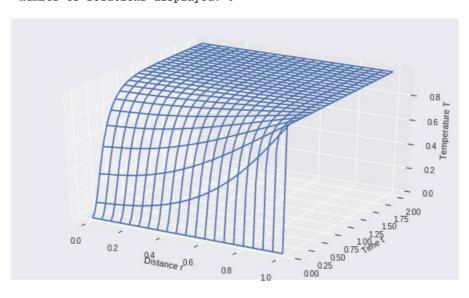
with auxiliary conditions

$$egin{array}{ll} T(0,r) &= 0 & & orall 0 \leq r \leq 1 \ T(t,1) &= 1 & & orall t > 0 \ rac{\partial T}{\partial r}(t,0) = 0 & & orall t \geq 0 \end{array}$$

In [13]:

```
from pyomo.environ import *
from pyomo.dae import *
m = ConcreteModel()
m.r = ContinuousSet(bounds=(0,1))
m.t = ContinuousSet(bounds=(0,2))
m \cdot T = Var(m \cdot t, m \cdot r)
m.dTdt = DerivativeVar(m.T, wrt=m.t)
m.dTdr = DerivativeVar(m.T, wrt=m.r)
m.d2Tdr2 = DerivativeVar(m.T, wrt=(m.r, m.r))
 \texttt{m.pde} = \texttt{Constraint}(\texttt{m.t}, \ \texttt{m.r}, \ \texttt{rule=lambda} \ \texttt{m}, \ \texttt{t}, \ \texttt{r:} \ \texttt{m.dTdt}[\texttt{t,r}] \ == \ \texttt{m.d2Tdr2}[\texttt{t,r}] \ + \ (1/r) *\texttt{m.r} 
dTdr[t,r]
         if r > 0 and r < 1 and t > 0 else Constraint.Skip)
m.ic = Constraint(m.r, rule=lambda m, r: m.T[0,r] == 0)
m.bc2 = Constraint(m.t, rule=lambda m, t: m.dTdr[t,0] == 0)
TransformationFactory('dae.finite_difference').apply_to(m, nfe=20, wrt=m.r, scheme='CEN
TransformationFactory('dae.finite_difference').apply_to(m, nfe=50, wrt=m.t, scheme='BAC
KWARD')
SolverFactory('ipopt', executable=ipopt_executable).solve(m).write()
model_plot(m)
```

```
# = Solver Results
Problem Information
Problem:
- Lower bound: -inf
 Upper bound: inf
 Number of objectives: 1
 Number of constraints: 4060
 Number of variables: 4110
 Sense: unknown
  Solver Information
Solver:
- Status: ok
 Message: Ipopt 3.12.8\x3a Optimal Solution Found
 Termination condition: optimal
 Error rc: 0
 Time: 0.16563725471496582
  Solution Information
# -----
Solution:
- number of solutions: 0
 number of solutions displayed: 0
```



Spherical Coordinates

Suppressing the prime notation, for a cylindrical geometry the model specializes to

$$\frac{\partial T}{\partial t} = \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \frac{\partial T}{\partial r} \right)$$

Expanding,

$$rac{\partial T}{\partial t} = rac{\partial^2 T}{\partial t^2} + rac{2}{r}rac{\partial T}{\partial r}$$

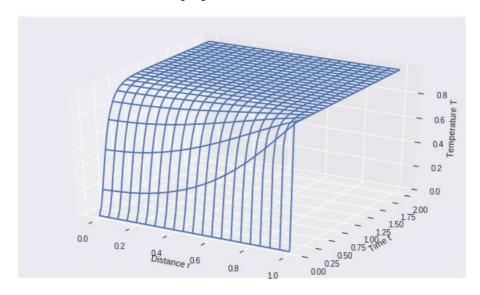
with auxiliary conditions

$$egin{array}{ll} T(0,r) &= 0 & \forall 0 \leq r \leq 1 \\ T(t,1) &= 1 & \forall t > 0 \\ rac{\partial T}{\partial r}(t,0) &= 0 & \forall t \geq 0 \end{array}$$

In [14]:

```
from pyomo.environ import *
from pyomo.dae import '
m = ConcreteModel()
m.r = ContinuousSet(bounds=(0,1))
m.t = ContinuousSet(bounds=(0,2))
m.T = Var(m.t, m.r)
m.dTdt = DerivativeVar(m.T, wrt=m.t)
       = DerivativeVar(m.T, wrt=m.r)
m.d2Tdr2 = DerivativeVar(m.T, wrt=(m.r, m.r))
m.pde = Constraint(m.t, m.r, rule=lambda m, t, r: m.dTdt[t,r] == m.d2Tdr2[t,r] + (2/r)*m.
dTdr[t,r]
        if r > 0 and r < 1 and t > 0 else Constraint.Skip)
m.ic = Constraint(m.r, rule=lambda m, r:
                                            m.T[0,r] == 0)
m.bc1 = Constraint(m.t, rule=lambda m, t: m.T[t,1] == 1 if t > 0 else Constraint.Skip
m.bc2 = Constraint(m.t, rule=lambda m, t: m.dTdr[t,0] == 0)
TransformationFactory('dae.finite_difference').apply_to(m, nfe=20, wrt=m.r, scheme='CEN
TransformationFactory('dae.finite_difference').apply_to(m, nfe=50, wrt=m.t, scheme='BAC
KWARD')
SolverFactory('ipopt', executable=ipopt_executable).solve(m).write()
model_plot(m)
```

```
# = Solver Results
Problem Information
Problem:
- Lower bound: -inf
 Upper bound: inf
 Number of objectives: 1
 Number of constraints: 4060
 Number of variables: 4110
 Sense: unknown
  Solver Information
Solver:
- Status: ok
 Message: Ipopt 3.12.8\x3a Optimal Solution Found
 Termination condition: optimal
 Error rc: 0
 Time: 0.15887713432312012
  Solution Information
# -----
Solution:
- number of solutions: 0
 number of solutions displayed: 0
```



```
In [ ]:
```