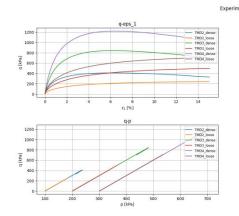
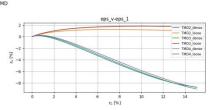
Assignment 2

Task A: Determination of the Mohr-Coulomb model parameters

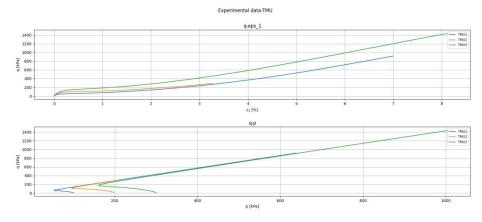
1. Experimental data plot

(1) Drained triaxial test





(2) Undrained triaxial test



(3) Plot code

Plot the experimental data

import numpy as np

import matplotlib.pyplot as plt

from plottingTaskA import plotting testData

from dataExperiment import *

from scipy.optimize import curve_fit

import math

Data group

expData TMD =

[expData()[0],expData()[1],expData()[2],expData()[3],expData()[4],expData()[5]]

labels_TMD = ('TMD2_dense', 'TMD2_loose', 'TMD3_dense', 'TMD3_loose', 'TMD4_dense', 'TMD4_loose')

expData_TMU = [expData()[6],expData()[7],expData()[8]]

labels_TMU = ('TMU1', 'TMU2', 'TMU3')

Emperimental data plotting

print("The experimental data you want to plot: ",

```
"1: txd, 2: txu"
)

testType = input("which kind of plot you want to realize: ")

if testType == '1':
    plotting_testData(expData_TMD,labels_TMD,'txd',None,None,None)

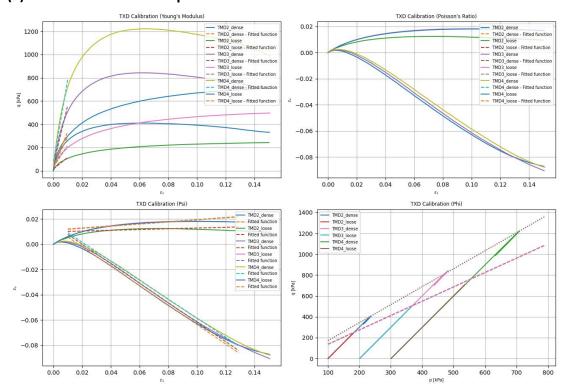
elif testType == '2':
    plotting_testData(expData_TMU,labels_TMU,'txu',None,None,None)
```

2. Parameters calibration and Plot results

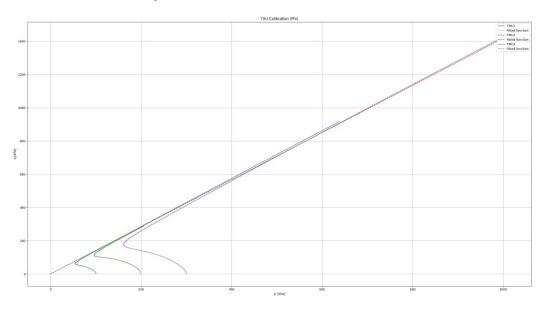
(1) Parameters calibration results:

	E [kPa]	nu [-]	Phi [°]	Psi [°]
TMD2_dense	25630.5	0.325	42	16.2
TMD3_dense	51363.0	0.282	42	16.2
TMD4_dense	70741.3	0.260	42	15.9
TMD2_loose	10003.7	0.161	34	-0.9
TMD3_loose	18461.5	0.167	34	-2.5
TMD4_loose	30030.6	0.146	34	-2.5
TMU1	25647.6	0.325	35	16.2
TMU2	51397.3	0.325	35	16.2
TMU3	70788.5	0.325	35	16.2

(2) Parameters calibration plot under drained triaxial test:



(3) Parameters calibration plot under undrained triaxial test:



(4) Calibration code

Calibration of the parameters

def test_func_linear(x, a, b):
 return a * x + b

Data group

TMD = [expData()[0],expData()[1],expData()[2],expData()[3],expData()[4],expData()[5]] labels1 = ['TMD2_dense', 'TMD2_loose', 'TMD3_dense', 'TMD3_loose', 'TMD4_dense', 'TMD4_loose']

```
TMU = [expData()[6],expData()[7],expData()[8]]
labels2 = ['TMU1', 'TMU2', 'TMU3']
# Fit parameters
def plot_phi_fit(ax, phi, p, q, line_style):
  M = 6 * math.sin(math.radians(phi)) / (3 - math.sin(math.radians(phi)))
  ax.plot(np.arange(100, 800, 10), test_func_linear(np.arange(100, 800, 10), M, 0),
line style)
  return phi
def fit params(data type, labels):
  if data_type == 'TMD':
    fig, axs = plt.subplots(2, 2, figsize=(13, 9))
    for idx, (data, label) in enumerate(zip(TMD, labels)):
      eps_1 = data[:, 0]/100
      q = data[:, 5]
      eps_v = data[:, 1]/100
      p = data[:, 6]
      # E-fit
      indices = np.where(eps 1 < 0.01)
      eps 1 fit = eps 1[indices]
      q fit = q[indices]
      p0 = [10000, 0]
      params, params_covariance = curve_fit(test_func_linear, eps_1_fit, q_fit, p0)
      # nu-fit
      indices_nu = np.where(eps_1 < 0.005)
      eps 1 fitNu = eps 1[indices nu]
      eps_v_fitNu = eps_v[indices_nu]
      p1 = [0.5, 0]
      paramsNu, paramsNu covariance = curve fit(test func linear, eps 1 fitNu,
eps_v_fitNu, p1)
      paramsNu = [1/2 * (1-i) \text{ for } i \text{ in paramsNu}]
      # Psi-fit
      eps 1 fitPsi = eps 1[np.where((eps 1 > 0.025) & (eps 1 < 0.1))]
      eps_vPsi = eps_v[np.where((eps_1 > 0.025) & (eps_1 < 0.1))]
      p2 = [1, -1]
      paramsPsi, paramsPsi covariance = curve fit(test func linear, eps 1 fitPsi,
eps_vPsi, p2)
      sin Psi = paramsPsi[0] / (paramsPsi[0] - 2)
      Psi deg = math.degrees(math.asin(sin Psi))
      # Fitting and Plotting
```

```
axs[0, 0].plot(eps 1, q, label=label)
       axs[0, 0].plot(eps_1_fit, test_func_linear(eps_1_fit, *params), '--', label=f'{label} -
Fitted function')
       axs[0, 1].plot(eps_1, eps_v, label=label)
       axs[0, 1].plot(eps 1 fitNu, test func linear(eps 1 fitNu, *paramsNu), '--',
label=f'{label} - Fitted function')
       axs[1, 0].plot(eps 1, eps v, label=label)
       axs[1, 0].plot(np.arange(0.01, 0.13, 0.001), test func linear(np.arange(0.01, 0.13,
0.001), paramsPsi[0], paramsPsi[1]),'--',
               label='Fitted function')
       # Phi-fit
       axs[1,1].plot(p, q, label=label)
       phi loose = plot phi fit(axs[1, 1], 34, p, q, '--')
       phi dense = plot phi fit(axs[1, 1], 42, p, q, ':')
       print(f"Optimized parameters for {label} (E, C):", params)
       print(f"Young's modulus E = {params[0]:.1f} kPa")
       print(f"Optimized parameters for {label} (Nu, C):", paramsNu)
       print(f"Poisson's ratio Nu = {paramsNu[0]:.3f}")
       print(f"Optimized parameters for {label} (Psi, C):", paramsPsi)
       print(f"Dilatancy angle Psi = {Psi_deg:.1f} degrees")
       print(f"Phi for loose sample = {phi loose} degrees")
       print(f"Phi for dense sample = {phi dense} degrees")
    axs[0, 0].set_title('TXD Calibration (Young\'s Modulus)', fontsize='small')
    axs[0, 0].set xlabel('$\\epsilon {1}$', fontsize='x-small')
    axs[0, 0].set ylabel('q [kPa]', fontsize='x-small')
    axs[0, 0].legend(loc='upper right', fontsize='x-small')
    axs[0, 0].grid(which='both')
    axs[0, 1].set title('TXD Calibration (Poisson\'s Ratio)', fontsize='small')
    axs[0, 1].set xlabel('$\\epsilon {1}$', fontsize='x-small')
    axs[0, 1].set_ylabel('$\\epsilon_{v}$', fontsize='x-small')
    axs[0, 1].legend(loc='upper right', fontsize='x-small')
    axs[0, 1].grid(which='both')
    axs[1, 0].set title('TXD Calibration (Psi)', fontsize='small')
    axs[1, 0].set_xlabel('$\\epsilon_{1}$', fontsize='x-small')
    axs[1, 0].set ylabel('$\\epsilon {v}$', fontsize='x-small')
    axs[1, 0].legend(loc='upper right', fontsize='x-small')
    axs[1, 0].grid(which='both')
    axs[1, 1].set title('TXD Calibration (Phi)', fontsize='small')
    axs[1, 1].set xlabel("p [kPa]", fontsize='x-small')
    axs[1, 1].set ylabel('q [kPa]', fontsize='x-small')
```

```
axs[1, 1].legend(loc='upper left', fontsize='x-small')
    axs[1, 1].grid(which='both')
  elif data_type == 'TMU':
    fig, axs = plt.subplots(1, 1, figsize=(10, 7))
    for idx, (data, label) in enumerate(zip(TMU, labels)):
      q = data[:, 7]
      p = data[:, 6]
      # Phi-fit
      phi = 35
      M = 6*math.sin(math.radians(phi))/(3-math.sin(math.radians(phi)))
      # Fitting and Plotting
      axs.plot(p, q, label=label)
      axs.plot(np.arange(0,1000,10), test func linear(np.arange(0,1000,10), M, 0),'--',
label='Fitted function')
    axs.set title('TXU Calibration (Phi)')
    axs.set_xlabel(r"$p'$ [kPa]")
    axs.set ylabel('q [kPa]')
    axs.legend(loc='upper right', fontsize='medium')
    axs.grid(which='both')
    print(f"Friction angle Phi = {phi:.1f} degrees")
  plt.subplots adjust(hspace=0.6, wspace=0.4)
  plt.tight layout()
  plt.show()
fit_params('TMD', labels1)
fit_params('TMU', labels2)
```

3. Original source code:

dataExperiment.py: for combine the given experimental data by defining the function "dataExperiment" taskA.py: for experimental plotting and parameters calibration

taskA.py: for experimental plotting and parameters calibration plottingTaskA.py: for taskA plottin

TaskB: Extend the Python routines and run simulations

1. Implement the Drucker-Prager failure criterion in the python code

```
elif model pl == "dp":
 # converting friction and dilatancy angles from [°] to [rad]
 phi = math.radians(modelParam["phi"])
 psi = math.radians(modelParam["psi"])
 p = stress[0]
 q = stress[1]
 # Drucker-Prager failure criterion
 f = q - (6 * math.sin(phi) / (3 - math.sin(phi))) * p
 # gradient of yield surface
 df_dp = -(6 * math.sin(phi) / (3 - math.sin(phi)))
 df dq = 1
 # gradient of plastic potential
 dg dp = -(6 * math.sin(psi) / (3 - math.sin(psi)))
 dg_dq = 1
 # gradient of the additional state variables
 df dchi = 0
 dchi depsp = 0
 dchi depsq = 0
```

2. Simulate both txd and txu under MC and DP model and compare with experimental results.

Remark:

1) MC: Mohr-Coulomb, DP: Drucker-Prager, txd: drained triaxial test, txu: undrained triaxial test

```
a) MC model_dense soil_txd
```

```
import numpy, sys
# import additional functions
from functions elementTest import *
from functions matModels import *
from plotting routines import *
from dataExperiment import *
# Input
# initial state
epor = 0.697 # void ratio
p = 300
        # [kPa]
q = 2.00 # [kPa]
eps_1 = 0 # [-]
eps 2 = 0 #[-]
```

```
# "linelast" ... linear elasticity
model el = "linelast"
#"mc" ... mohr-coulomb
# "dp" ... drucker-prager
model pl = "mc"
# model pl = "dp"
model={"model el":model el, "model pl":model pl}
# material parameters
E = 70741.3
nu = 0.260
phi = 42
psi = 15.9
modelParam={"E":E, "nu":nu, "phi":phi, "psi":psi}
# test control
# "oed" ... oedometer test
#"txd" ... drained triaxial test
#"txu" ... undrained triaxial test
testType = 'txd'
# stop criteria for calculation
p max = 1000 # [kPa]
epsq \max = 0.2 \# [\%]
i_max = 1000 # max iterations
# numerical parameter
dt = 0.001 # time step (numerical integration)
#-----
=======
# Calculation
=======
# inital stress and strain in volumetric and deviatoric invariants
stress = numpy.array([p, q])
strain = eps2pq(eps_1,eps_2)
stateVar = numpy.array([epor,0])
sigma 1 = pq2sigma(stress[0], stress[1])[0]
sigma_2 = pq2sigma(stress[0], stress[1])[1]
# record the state variables in a numpy array in the following order
# column: eps_p, eps_q, p, q, eps_1, eps_2, sigma_1, sigma_2, epor
# every row represents a time step
data = numpy.array([[strain[0], strain[1], stress[0], stress[1], eps 1, eps 2, sigma 1,
sigma 2, epor, stateVar[1] ]])
# initialise the loop
step = 1
while (stress[0] < p max) and (strain[1] < epsq max) and (step < i max):
```

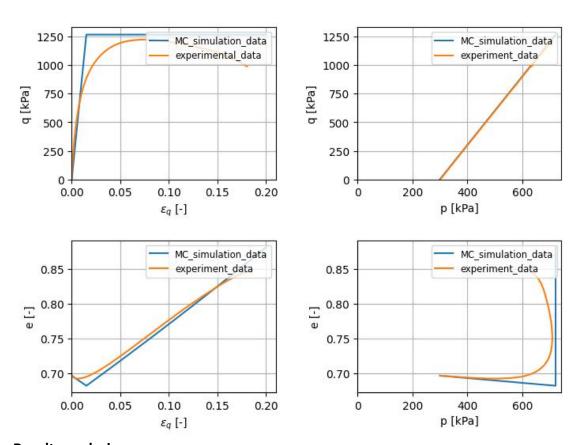
dstress, dstrain, dStateVar = rates(stress,strain,stateVar,testType,model,modelParam) stress, strain, stateVar = integration(stress,strain,dstress,dstrain,stateVar,dStateVar,dt) data = recordData(data,stress,strain,stateVar) step += 1

open output file for saving the data and the plot filename = 'output ' + testType + 'expDataV.S.MC'+' TMD4 dense'

save results to a text file printResults(filename,data) data2 = expData()[4]

plot the results plotting_testData(data, testType='txd', outputName=filename, data2=data2, data3=None)

Drained triaxial test



Results analysis:

The overall simulation of MC model looks good, the reason is the good calibration of parameters and MC model is suitable for this soil situation. In details, for E, we can see at the elastic part of q-eps_q diagram, the experimental data and simulation data is approach, for phi, the plastic part of of q-eps_q diagram, and q-p diagram, the experimental data and simulation data is closed. for nu, we can see the linear part of the diagram e-eps_q, the two lines almost coincided. and the plastic range of e-eps_q, the psi also looks good. The last diagram can get a good simulation results of the dilatancy.

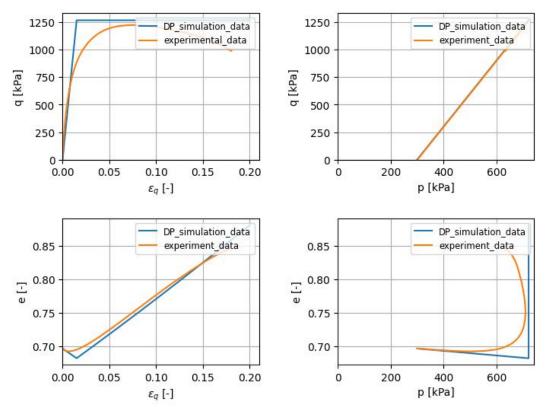
```
b) DP model_dense soil_txd
import numpy, sys
# import additional functions
from functions_elementTest import *
from functions matModels import *
from plotting routines import *
from dataExperiment import *
# Input
# initial state
epor = 0.697 # void ratio
p = 300 # [kPa]
q = 2.00 \# [kPa]
eps 1 = 0 # [-]
eps_2 = 0 #[-]
# material models for the elastic and the plastic part
# "linelast" ... linear elasticity
model el = "linelast"
#"mc" ... mohr-coulomb
# "dp" ... drucker-prager
# model pl = "mc"
model pl = "dp"
model={"model_el":model_el, "model_pl":model_pl}
# material parameters
E = 70741.3
nu = 0.260
phi = 42
psi = 15.9
modelParam={"E":E, "nu":nu, "phi":phi, "psi":psi}
# test control
# "oed" ... oedometer test
#"txd" ... drained triaxial test
#"txu" ... undrained triaxial test
testType = 'txd'
# stop criteria for calculation
p max = 1000 # [kPa]
epsq_max = 0.2 \# [\%]
i_max = 1000 # max iterations
# numerical parameter
```

time step (numerical integration)

dt = 0.001

```
# Calculation
# inital stress and strain in volumetric and deviatoric invariants
stress = numpy.array([p, q])
strain = eps2pq(eps_1,eps_2)
stateVar = numpy.array([epor,0])
sigma_1 = pq2sigma(stress[0], stress[1])[0]
sigma_2 = pq2sigma(stress[0], stress[1])[1]
# record the state variables in a numpy array in the following order
# column: eps_p, eps_q, p, q, eps_1, eps_2, sigma_1, sigma_2, epor
# every row represents a time step
data = numpy.array([[strain[0], strain[1], stress[0], stress[1], eps 1, eps 2, sigma 1,
sigma_2, epor, stateVar[1] ]])
# initialise the loop
step = 1
while (stress[0] < p max) and (strain[1] < epsq max) and (step < i max):
  dstress, dstrain, dStateVar = rates(stress, strain, stateVar, testType, model, modelParam)
  stress, strain, stateVar = integration(stress, strain, dstress, dstrain, stateVar, dStateVar, dt)
  data = recordData(data, stress, strain, stateVar)
  step += 1
# open output file for saving the data and the plot
filename = 'output_' + testType + 'expDataV.S.DP'+' TMD4 dense'
# save results to a text file
printResults(filename,data)
data2 = expData()[4]
# plot the results
plotting testData(data, testType='txd', outputName=filename, data2=data2,
data3=None)
```

Drained triaxial test



Results analysis:

The simulation of DP model also looks good. And the results looks very similar to the MC model, the reason is that we only simulate the compression test.

c) MC model_loose soil_txd

import numpy, sys
import additional functions
from functions_elementTest import *
from functions_matModels import *
from plotting_routines import *
from dataExperiment import *

```
# material models for the elastic and the plastic part
# "linelast" ... linear elasticity
model_el = "linelast"
# "mc" ... mohr-coulomb
# "dp" ... drucker-prager
model_pl = "mc"
```

```
# model pl = "dp"
model={"model el":model el, "model pl":model pl}
# material parameters
E = 30030.6
nu = 0.146
phi = 33
psi = -0.9
modelParam={"E":E, "nu":nu, "phi":phi, "psi":psi}
# test control
# "oed" ... oedometer test
# "txd" ... drained triaxial test
# "txu" ... undrained triaxial test
testType = 'txd'
# stop criteria for calculation
p max = 1000 \# [kPa]
epsq_max = 0.2# [%]
i max = 1000 # max iterations
# numerical parameter
dt = 0.001
               # time step (numerical integration)
# Calculation
# inital stress and strain in volumetric and deviatoric invariants
stress = numpy.array([p, q])
strain = eps2pq(eps 1,eps 2)
stateVar = numpy.array([epor,0])
sigma 1 = pq2sigma(stress[0], stress[1])[0]
sigma 2 = pq2sigma(stress[0], stress[1])[1]
# record the state variables in a numpy array in the following order
# column: eps_p, eps_q, p, q, eps_1, eps_2, sigma_1, sigma_2, epor
# every row represents a time step
data = numpy.array([[strain[0], strain[1], stress[0], stress[1], eps_1, eps_2, sigma_1,
sigma_2, epor, stateVar[1] ]])
# initialise the loop
step = 1
while (stress[0] < p max) and (strain[1] < epsq max) and (step < i max):
  dstress, dstrain, dStateVar =
rates(stress, strain, stateVar, testType, model, modelParam)
  stress, strain, stateVar =
integration(stress, strain, dstress, dstrain, stateVar, dStateVar, dt)
  data = recordData(data, stress, strain, stateVar)
  step += 1
```

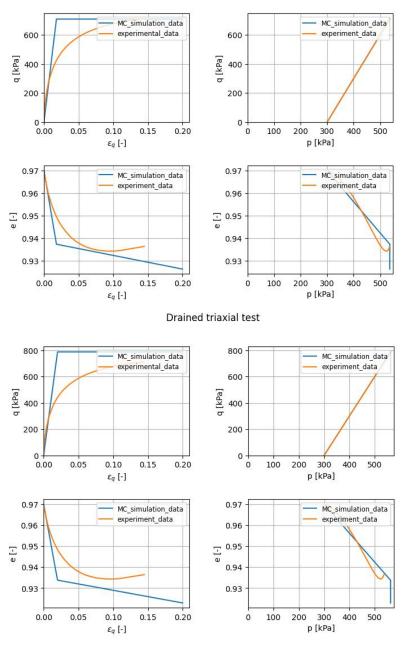
open output file for saving the data and the plot filename = 'output_' + testType + 'expDataV.S.MC'+'_TMD4_loose'

save results to a text file printResults(filename,data) data2 = expData()[5]

plot the results

plotting_testData(data, testType='txd', outputName=filename, data2=data2, data3=None)

Drained triaxial test



Results analysis:

- 1) The only difference about the 2 diagrams above is phi, the upper one phi=33, and the lower one is phi=34(the results from taskA's calibration). Although both simulation results seems not bad for comparison with the experimental results, obviously the lower one seems better than the upper. And the only slight calibration and can get the better simulation.
- 2) Compare with the dense soil simulation results that we got above, from the q-eps_q diagram, the friction angle of dense soil is larger than the loose. And from e-eps_q, the big difference is the plastic part, the dense soil exhibits dialatancy and loose soil exhibits contraction. And seem in e-p diagram, for loose soil, with p increasing, the

void ratio decrease significantly, but for dense soil, with p increasing, the void ratio change little at the beginning, after about p=690kpa, the void ratio increase significantly.

d) DP model_loose soil_txd

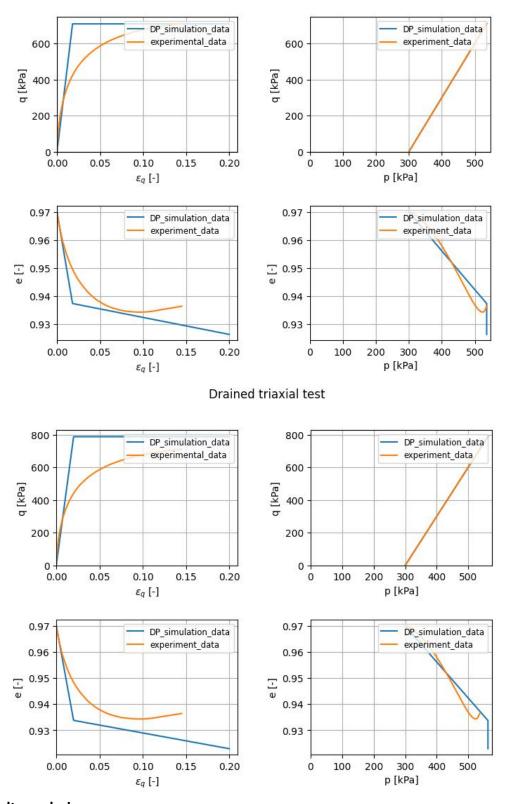
```
import numpy, sys
# import additional functions
from functions elementTest import *
from functions matModels import *
from plotting routines import *
from dataExperiment import *
# Input
# initial state
epor = 0.97 # void ratio
p = 300
             # [kPa]
q = 1.16 # [kPa]
eps 1 = 0 # [-]
eps_2 = 0 # [-]
# material models for the elastic and the plastic part
# "linelast" ... linear elasticity
model el = "linelast"
# "mc" ... mohr-coulomb
# "dp" ... drucker-prager
# model pl = "mc"
model pl = "dp"
model={"model_el":model_el, "model_pl":model_pl}
# material parameters
E = 30030.6
nu = 0.146
phi = 33
psi = -0.9
modelParam={"E":E, "nu":nu, "phi":phi, "psi":psi}
# test control
# "oed" ... oedometer test
# "txd" ... drained triaxial test
# "txu" ... undrained triaxial test
testType = 'txd'
# stop criteria for calculation
p_max = 1000 \# [kPa]
epsq max = 0.2 \# [\%]
i max = 1000 # max iterations
```

numerical parameter

```
# time step (numerical integration)
========
# Calculation
=======
# inital stress and strain in volumetric and deviatoric invariants
stress = numpy.array([p, q])
strain = eps2pq(eps 1,eps 2)
stateVar = numpy.array([epor,0])
sigma 1 = pq2sigma(stress[0], stress[1])[0]
sigma_2 = pq2sigma(stress[0], stress[1])[1]
# record the state variables in a numpy array in the following order
# column: eps p, eps q, p, q, eps 1, eps 2, sigma 1, sigma 2, epor
# every row represents a time step
data = numpy.array([[strain[0], strain[1], stress[0], stress[1], eps 1, eps 2, sigma 1,
sigma_2, epor, stateVar[1] ]])
# initialise the loop
step = 1
while (stress[0] < p max) and (strain[1] < epsq max) and (step < i max):
  dstress, dstrain, dStateVar =
rates(stress, strain, stateVar, testType, model, modelParam)
  stress, strain, stateVar =
integration(stress, strain, dstress, dstrain, stateVar, dStateVar, dt)
  data = recordData(data, stress, strain, stateVar)
  step += 1
# open output file for saving the data and the plot
filename = 'output ' + testType + 'expDataV.S.DP'+' TMD4 loose'
# save results to a text file
printResults(filename,data)
data2 = expData()[1]
# plot the results
plotting testData(data, testType='txd', outputName=filename, data2=data2,
```

dt = 0.001

data3=None)



Results analysis:

The simulation of DP model also looks good. And the results looks very similar to the MC model, the reason is that we only simulate the compression test.

e) MC model_txu

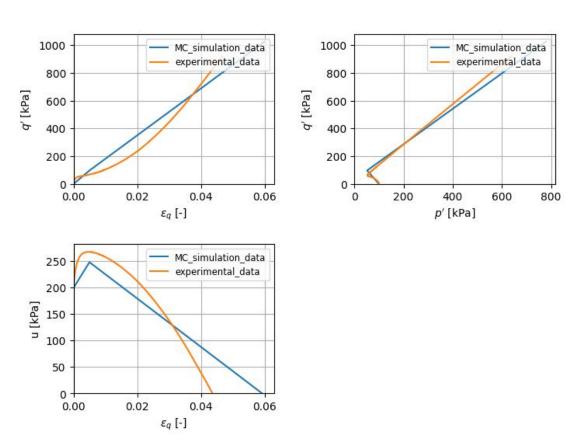
import numpy, sys
import additional functions
from functions_elementTest import *
from functions_matModels import *
from plotting_routines import *

```
from dataExperiment import *
# Input
# initial state
epor = 0.828 # void ratio
p = 100 # [kPa]
q = 0.559 # [kPa]
eps 1 = 0 # [-]
eps 2 = 0 #[-]
# material models for the elastic and the plastic part
# "linelast" ... linear elasticity
model el = "linelast"
# "mc" ... mohr-coulomb
# "dp" ... drucker-prager
model pl = "mc"
# model pl = "dp"
model={"model_el":model_el, "model_pl":model_pl}
# material parameters
E = 25647.6 #[kPa] ... Young's modulus
nu = 0.325 # [-] ... Poisson's ratio
           # [°] ... friction angle
phi = 35
psi = 16.2 # [°] ... dilatancy angle
modelParam={"E":E, "nu":nu, "phi":phi, "psi":psi}
# test control
# "oed" ... oedometer test
# "txd" ... drained triaxial test
# "txu" ... undrained triaxial test
testType = 'txu'
# stop criteria for calculation
p_max = 1000 \# [kPa]
epsq_max = 0.06
                # [%]
i_max = 1000 # max iterations
# numerical parameter
dt = 0.001
         # time step (numerical integration)
========
# Calculation
========
# inital stress and strain in volumetric and deviatoric invariants
stress = numpy.array([p, q])
strain = eps2pq(eps 1,eps 2)
```

```
stateVar = numpy.array([epor,0])
sigma 1 = pq2sigma(stress[0], stress[1])[0]
sigma_2 = pq2sigma(stress[0], stress[1])[1]
# record the state variables in a numpy array in the following order
# column: eps p, eps q, p, q, eps 1, eps 2, sigma 1, sigma 2, epor
# every row represents a time step
data = numpy.array([[strain[0], strain[1], stress[0], stress[1], eps_1, eps_2, sigma_1,
sigma_2, epor, stateVar[1] ]])
# initialise the loop
step = 1
while (stress[0]  and <math>(strain[1] < epsq max) and (step < i max):
  dstress, dstrain, dStateVar =
rates(stress, strain, stateVar, testType, model, modelParam)
  stress, strain, stateVar =
integration(stress, strain, dstress, dstrain, stateVar, dStateVar, dt)
  data = recordData(data,stress,strain,stateVar)
  step += 1
# open output file for saving the data and the plot
filename = 'output '+ testType + 'expDataV.S.MC'+' TMU1'
# save results to a text file
printResults(filename,data)
data2 = expData()[6]
# plot the results
```

plotting_testData(data, testType='txu', outputName=filename, data2=data2, data3=None)

Undrained triaxial test



Results analysis:

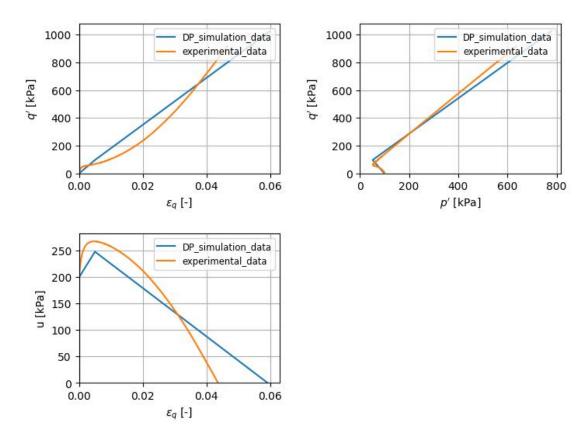
Compare with the drained triaxial simulation results, the MC simulation in undrained triaxial test looks not at that good, the reason is Mohr-Coulomb model is based on the effective stress, the pore water pressure obviously will have an effect on the simulation results. But the overall trend from the simulation results and experimental data are similar.

f) DP model TXU

```
import numpy, sys
# import additional functions
from functions elementTest import *
from functions matModels import *
from plotting routines import *
from dataExperiment import *
# Input
# initial state
epor = 0.828 # void ratio
p = 100
              # [kPa]
q = 0.559
            # [kPa]
eps 1 = 0
            # [-]
            #[-]
eps 2 = 0
# material models for the elastic and the plastic part
# "linelast" ... linear elasticity
model el = "linelast"
# "mc" ... mohr-coulomb
# "dp" ... drucker-prager
# model pl = "mc"
model pl = "dp"
model={"model el":model el, "model pl":model pl}
# material parameters
E = 25647.6
              #[kPa] ... Young's modulus
nu = 0.325
              # [-] ... Poisson's ratio
              # [°] ... friction angle
phi = 35
psi = 16.2
              # [°] ... dilatancy angle
modelParam={"E":E, "nu":nu, "phi":phi, "psi":psi}
# test control
# "oed" ... oedometer test
# "txd" ... drained triaxial test
# "txu" ... undrained triaxial test
testType = 'txu'
# stop criteria for calculation
p max = 1000 # [kPa]
epsq max = 0.06
                    # [%]
i max = 1000 # max iterations
```

```
# numerical parameter
dt = 0.001
              # time step (numerical integration)
# Calculation
# inital stress and strain in volumetric and deviatoric invariants
stress = numpy.array([p, q])
strain = eps2pq(eps_1,eps_2)
stateVar = numpy.array([epor,0])
sigma 1 = pq2sigma(stress[0], stress[1])[0]
sigma 2 = pq2sigma(stress[0], stress[1])[1]
# record the state variables in a numpy array in the following order
# column: eps_p, eps_q, p, q, eps_1, eps_2, sigma_1, sigma_2, epor
# every row represents a time step
data = numpy.array([[strain[0], strain[1], stress[0], stress[1], eps 1, eps 2, sigma 1,
sigma_2, epor, stateVar[1] ]])
# initialise the loop
step = 1
while (stress[0] < p max) and (strain[1] < epsq max) and (step < i max):
  dstress, dstrain, dStateVar =
rates(stress, strain, stateVar, testType, model, modelParam)
  stress, strain, stateVar =
integration(stress, strain, dstress, dstrain, stateVar, dStateVar, dt)
  data = recordData(data,stress,strain,stateVar)
  step += 1
# open output file for saving the data and the plot
filename = 'output ' + testType + 'expDataV.S.DP'+' TMU1'
# save results to a text file
printResults(filename,data)
data2 = expData()[6]
# plot the results
plotting testData(data, testType='txu', outputName=filename, data2=data2,
data3=None)
```

Undrained triaxial test



Results analysis:

The simulation results of DP model looks very similar to the MC model, the reason is that we only simulate the compression test.

3. Original source code:

Remark: see details in python scripts

functions_elementTest.py: for define the txd and txu element test

fnctions_matModels.py: for extend the Mohr-Coulomb model and the Drucker-Prager

model

plotting_routines.py: for plot task B

taskB_TXD.py:for simulate the two models under drained triaxial test taskB_TXU.py: for simulate the two models under undrained triaxial test