

plumeAnalyserInversion

May 31, 2019

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In [1]: from bentPlumeAnalyser import *
        from fumarolePlumeModel import *
        from scipy.io.matlab import loadmat
        from itertools import product

        import pandas
        import numpy as np
        import matplotlib.pyplot as plt
        import matplotlib as mpl
        import os, sys
        import json

        # Set numpy options, notably for the printing of floating point numbers
        np.set_printoptions(precision=6)

        # Set matplotlib options
        mpl.rcParams['figure.dpi'] = 300

In [2]: def integrator(V0, p):
        # Initialise an integrator object
        r = ode(derivs).set_integrator('lsoda', nsteps=1e6)
        r.set_initial_value(V0, 0.)
        r.set_f_params(p)

        # Define state vector and axial distance
        V = []      # State vector
        s = []      # Axial distance
        V.append(V0)
        s.append(sexp[0])

        # Define the individual variables - these will be calculated at run time
        Q, M, F, theta = 0., 0., 0., 0.
        Q = np.float64(Q0)
        M = np.float64(M0)
        F = np.float64(F0)
        theta = np.float64(theta0)
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# Integrate, whilst successful, until the domain size is reached
ind = 0
while r.successful() and r.t < t1 and M >= 0.:
    dt = dsexp[ind]
    r.integrate(r.t + dt)
    V.append(r.y)
    s.append(r.t)
    Q, M, F, theta = r.y
    ind += 1
s = np.array(s)
V = np.float64(np.array(V))
return s, V

In [3]: exptNo      = 3
        plotResults = True

        # Read analysed experimental data from file
        fname      = './data/ExpPlumes_for_Dai/GCTA_plumeData.xlsx'
        # exptData = pandas.read_excel(fname, sheet_name='exp%02d' % exptNo)

        # # Display the first 5 lines of the data frame
        # exptData.head()

In [4]: ## DEFINE THE DISTANCE ALONG THE AXIS, THE ANGLE AND THE WIDTH OF THE PLUME
        # sexp = exptData.distAlongAxis
        # thexp = exptData.plumeAngle
        # bexp = exptData.plumeWidth

        path = './data/ExpPlumes_for_Dai/exp%02d/' % exptNo

        skip = 10 # How many points to skip

        xexp = loadmat(path + 'xcenter.mat')['xcenter'][0]
        zexp = loadmat(path + 'zcenter.mat')['zcenter'][0]
        Ox, Oz = (xexp[0], zexp[0])
        pPixels = np.array([xexp, zexp]).T[:, :skip]

        data, xexp, zexp, extent = loadExptData(exptNo)
        if data.mean() < .5:
            data = 1. - data
        p = np.array([xexp, zexp]).T[:, :skip]

        # # # Uncomment if user guesses are to be used for the plume trajectory
        # # with open(path + 'exp%02d_initGuess.json' % exptNo) as f:
        # #     exptData = json.load(f)
        # # p = np.array(exptData['data'])

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# pPixels = p.copy() * scaleFactor
# pPixels[:,0] += 0x
# pPixels[:,1] -= 0z
# pPixels[:,1] *= -1

# Calculate angle, width and distance along plume and errors
thexp, sig_thexp = plumeAngle(p[:,0], p[:,1], errors=[1/scaleFactor]*2)
_, bexp, sig_p, sig_bexp = trueLocationWidth(pPixels, data, errors=[1/scaleFactor])
sexp      = distAlongPath(p[:,0], p[:,1])
bexp      /= scaleFactor
sig_bexp  /= scaleFactor
bexp[0]   = 0.55 / 2
thexp[0]  = np.pi / 2

In [5]: # Import table of experimental conditions
GCTA = pandas.read_excel('./data/ExpPlumes_for_Dai/TableA1.xlsx', sheet_name='CGSdata',
                        names=('exptNo', 'rhoa0', 'sig_rhoa0', 'N', 'sig_N', 'rho0', 'sig_rho0', 'gp', 'sig_gp', 'Q0', 'sig_Q0', 'M0', 'sig_M0', 'F0', 'sig_F0'))

# Extract densities of ambient and plume, and calculate g' at the source
expt = GCTA[GCTA['exptNo'] == 3]
rhoa0 = expt['rhoa0']
rho0 = expt['rho0']
g = 981 #cm/s2
gp0 = (rhoa0 - rho0) / rhoa0 * g

parameters = pandas.read_excel('./data/ExpPlumes_for_Dai/TableA1.xlsx', sheet_name='CGSP')
b0theoretical = parameters[parameters['property'] == 'nozzleSize']['value'].values[0]
u0theoretical = expt['U0'].values[0]

In [13]: # Load initial conditions for a given experiment, run the model for those conditions
# and then compare model and experimental data
# fig, ax = plt.subplots()
# ax.plot(exptData.plumeWidth, exptData.distAlongAxis, 'g.', label='natural')
V0, p = loadICsParameters(pathname, exptNo, alpha=0.05, beta=0, m=2)

#p = (0.05, .5, .012, 2., 4.)
nGrid = 51 # Number of grid points
b0Vec = np.linspace(.05, 2, nGrid) #cm
u0Vec = np.linspace(5, 30, nGrid) #cm/s
Q0Vec = u0Vec * b0Vec**2 #cm3/s
M0Vec = Q0Vec * u0Vec #cm4/s2

theta0 = np.pi / 2

objFn, initialConds = [], []
# Set integration domain and step size
t1 = sexp.max() # Domain of integration

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dsexp = np.diff(sexp) # Discretise the distance along plume axis - use as integration s

sequence = [Q0Vec, M0Vec]

for (Q0, M0) in list(product(*sequence)):
    F0 = Q0 * gp0
    V0 = [Q0, M0, F0, theta0]

    # Call the 'integrator' function (defined above) to solve
    # the model
    s, V = integrator(V0, p)
    Q, M, F, theta = [V[:,i] for i in range(4)]

    #####

    b = Q / np.sqrt(M) / np.sqrt(2) # Factor of sqrt{2} to correspond with top-hat mod
    u = M / Q
    gp = F / Q

    Vexp = np.array([thexp]).ravel(order='C')
    Vsyn = np.array([theta]).ravel(order='C')
    sigV = np.array([sig_thexp]).ravel(order='C')
    CovV = np.diag(sigV**2)

    objFn.append(objectiveFn(Vexp, Vsyn, cov=CovV, p=p))
    initialConds.append(V0)
    # ax.plot(b, s, '-', label='Model %.4f %.4f %.4f' % (Q0, M0, F0))

    #ax.legend(loc=5)
    # ax.set_xlim((0, 25))

    # Transform initialConds and objFn from lists to arrays,
    # reshaping the latter
    initialConds = np.array(initialConds)
    objFn = np.array(objFn).reshape((nGrid, nGrid))

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0.0.1 Plot the optimal model solution and the objective function for the parameter space that we have calculated

In [7]: fig, ax = plt.subplots(1, 2, figsize=(8,4))

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### Plot the model solution over the experimental image and data
# Model solution trajectory
xmod, zmod = [0.], [0.]
ds_ = np.diff(sexp)
for (ds, th) in zip(ds_, theta):
    xmod.append(xmod[-1] + ds * np.cos(th))
    zmod.append(zmod[-1] + ds * np.sin(th))

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# On an image of the experimental plume, show the plume trajectories for
# 1) GCTA, 2) our initial guess, 3) the model solution
ax[0].imshow(data, extent=extent, cmap=plt.cm.gray)
ax[0].invert_yaxis()
ax[0].set_xlabel(r'$x$/[cm]')
ax[0].set_ylabel(r'$z$/[cm]')
# 1) GCTA
ax[0].plot(xexp, zexp, 'r-', label='GCTA', lw=2)
# 2) our initial guess
# ax[0].plot(exptData.axisLocn_x, exptData.axisLocn_y, 'r--', label='Init. guess', lw=2)
# 3) the model solution
ax[0].plot(xmod, zmod, 'g--', label='model', lw=2)
ax[0].set_xlim((extent[:2]))
ax[0].legend(loc=2)

### Plot optimal values
ui, bi = np.where(objFn == objFn.max())
bOpt = b0Vec[bi[0]]
uOpt = u0Vec[ui[0]]

im = ax[1].pcolor(b0Vec, u0Vec, np.log10(objFn / objFn.sum()))
hX = ax[1].set_xlabel(r'$b_0$')
hY = ax[1].set_ylabel(r'$u_0$')
hL = ax[1].set_title(r'$b^*_0$ = %.3f cm, $u^*_0$ = %.2f cm/s' %
                    (bOpt, uOpt))

ax[1].plot(bOpt, uOpt, 'r.', ms=8, label='Values with max. prob.')
ax[1].plot(b0theoretical, u0theoretical, 'k.', ms=8, label='Supposed values')

cbar = fig.colorbar(im)
cbar.set_label(r'$\log_{10}\{\Theta(d,m)\}$')

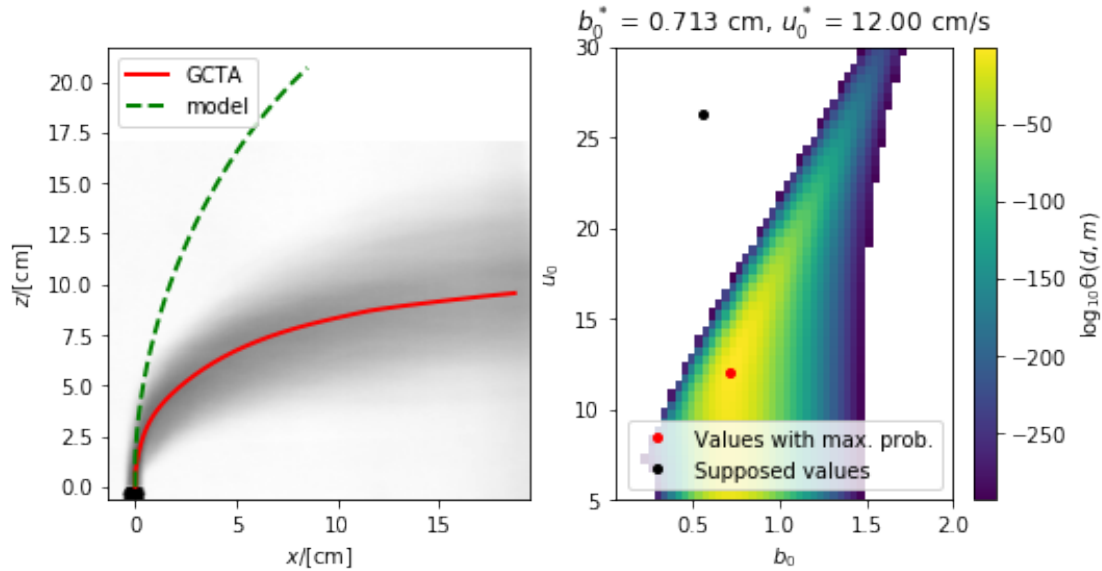
ax[1].legend(loc=4)
fig.savefig('ProbabilityDistribution_b0u0.png', dpi=300)

# ax[1].plot(res[:,0], sexp, '-', label=r'$\frac{b_{\mathrm{exp}} - b_{\mathrm{mod}}}{\sigma_{\mathrm{exp}}}$')
# ax[1].plot((thexp - theta) / sig_thexp, sexp, '-', label=r'$\frac{\theta_{\mathrm{exp}} - \theta_{\mathrm{mod}}}{\sigma_{\mathrm{exp}}}$')
# ax[1].legend(loc=1, fancybox=True, framealpha=.8)
# ax[1].grid()
# ax[1].set_ylabel('Distance along plume axis, s/[cm]')

# ax[1].set_title('Objective fn: %.3f' % objFn[bi][ui])

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/usr/share/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:31: RuntimeWarning: divid



(Left) Visualisation (logarithmic scale) of the probability density of the parameter space. (Right) Image of experimental plume with calculated and estimated trajectories. # (Right) Difference between experimental (guessed) and model solutions as a function of the distance along the plume axis.

1 To do:

- Identify which cases produce minima in the objective function

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In [8]: # Uncomment the following line to transform this notebook into a latex file
!jupyter nbconvert --to latex plumeAnalyser.ipynb
```

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[NbConvertApp] Converting notebook plumeAnalyser.ipynb to latex
[NbConvertApp] Support files will be in plumeAnalyser_files/
[NbConvertApp] Making directory plumeAnalyser_files
[NbConvertApp] Writing 37194 bytes to plumeAnalyser.tex
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In [9]: # Now run pdflatex plumeAnalyser from the command line
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In [11]: CovV.shape
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Out[11]: (143, 143)
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