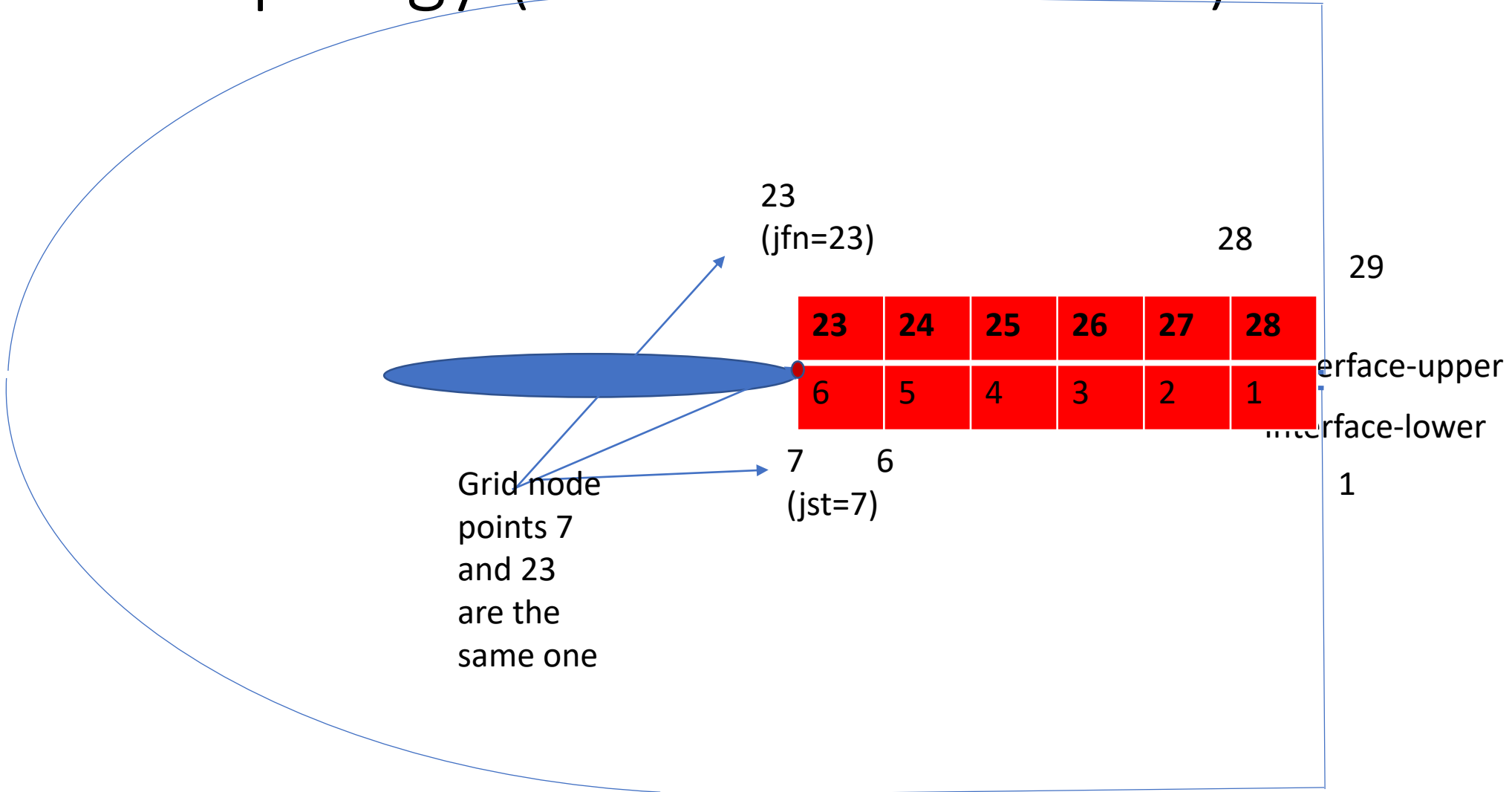


Instruction

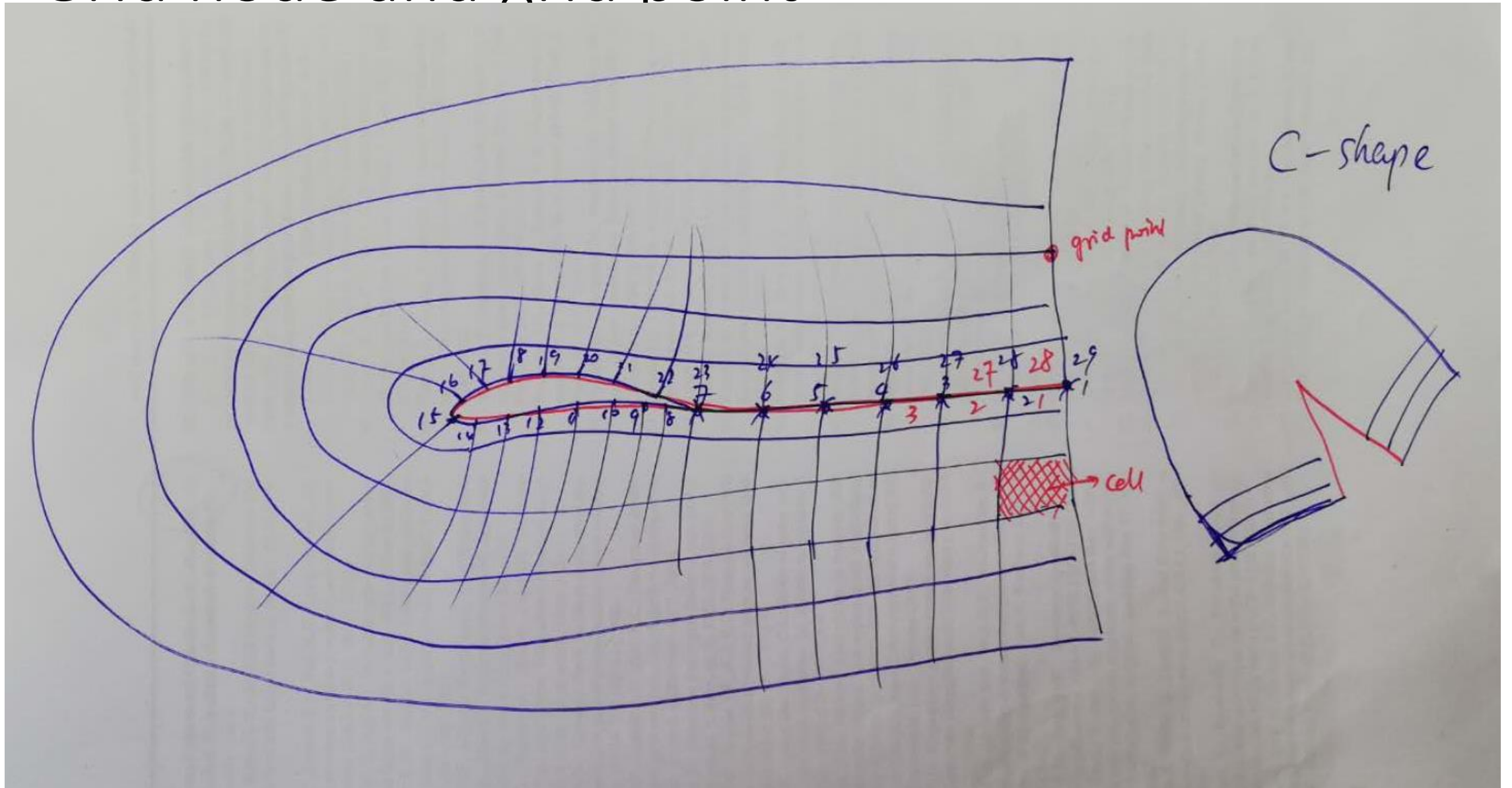
Liwei Chen

2023-02-10

Grid Topology (indices start from 1)



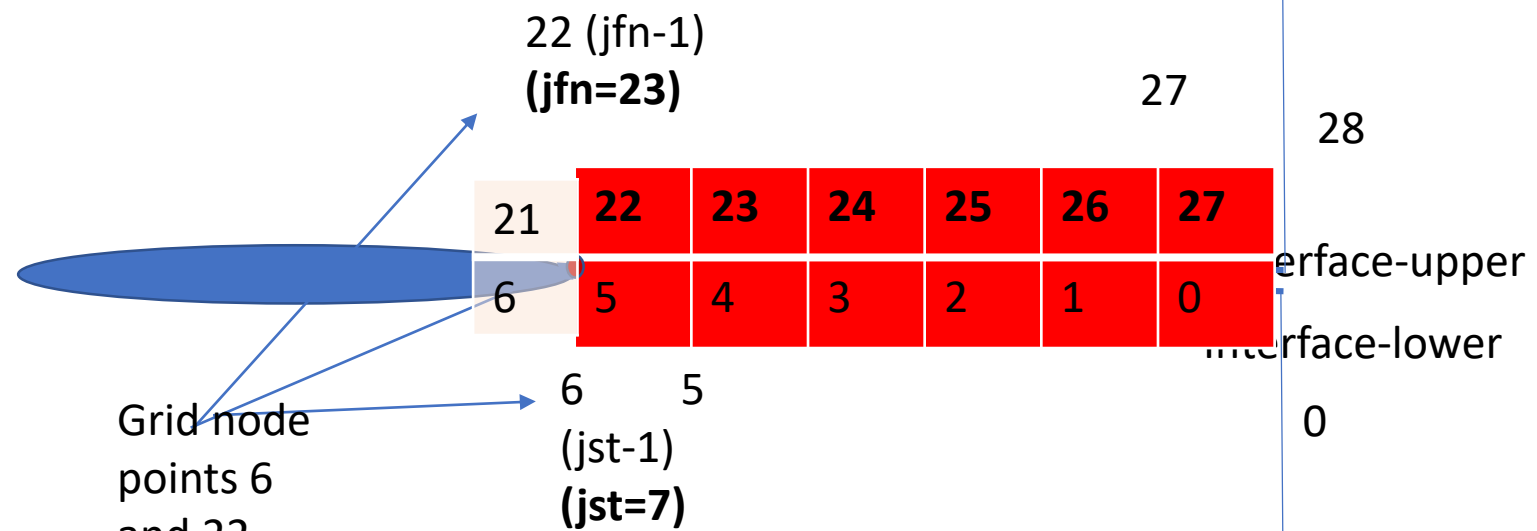
Grid node and grid point



Some notes

- Grid node file is under the folder: train_mesh_point (129 x 129 grid points)
- Grid cell file in the folder: train_mesh (128 x 128 cells)
- All the flowfield variables are defined on grid cell: 128 x 128. Files are under the folder "train"
- At the moment, I only provide you three airfoils

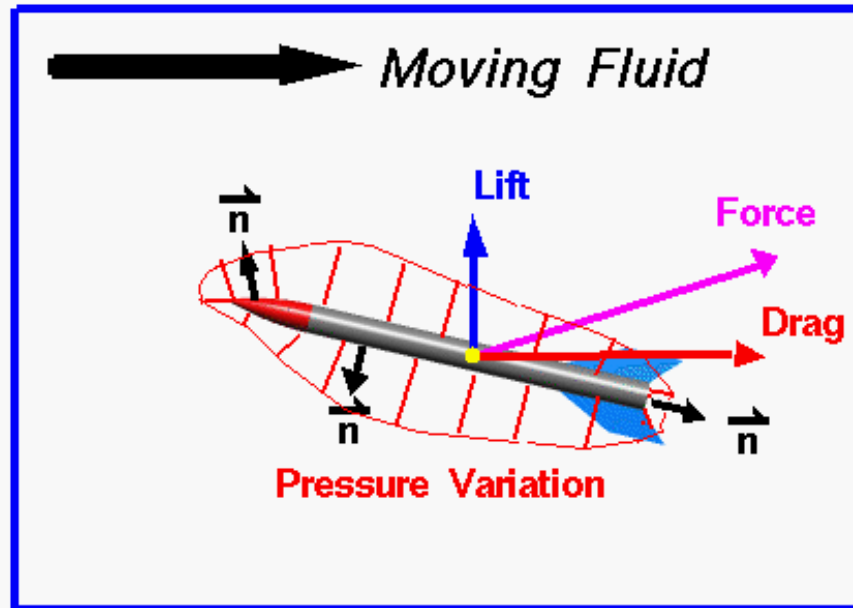
Grid Topology (indices start from 0)



How to calculate Force-X and Force-Y



Aerodynamic Forces



Pressure forces act normal (perpendicular) to surface.
Force on the body is the vector sum of the pressure x area
around the entire solid body.

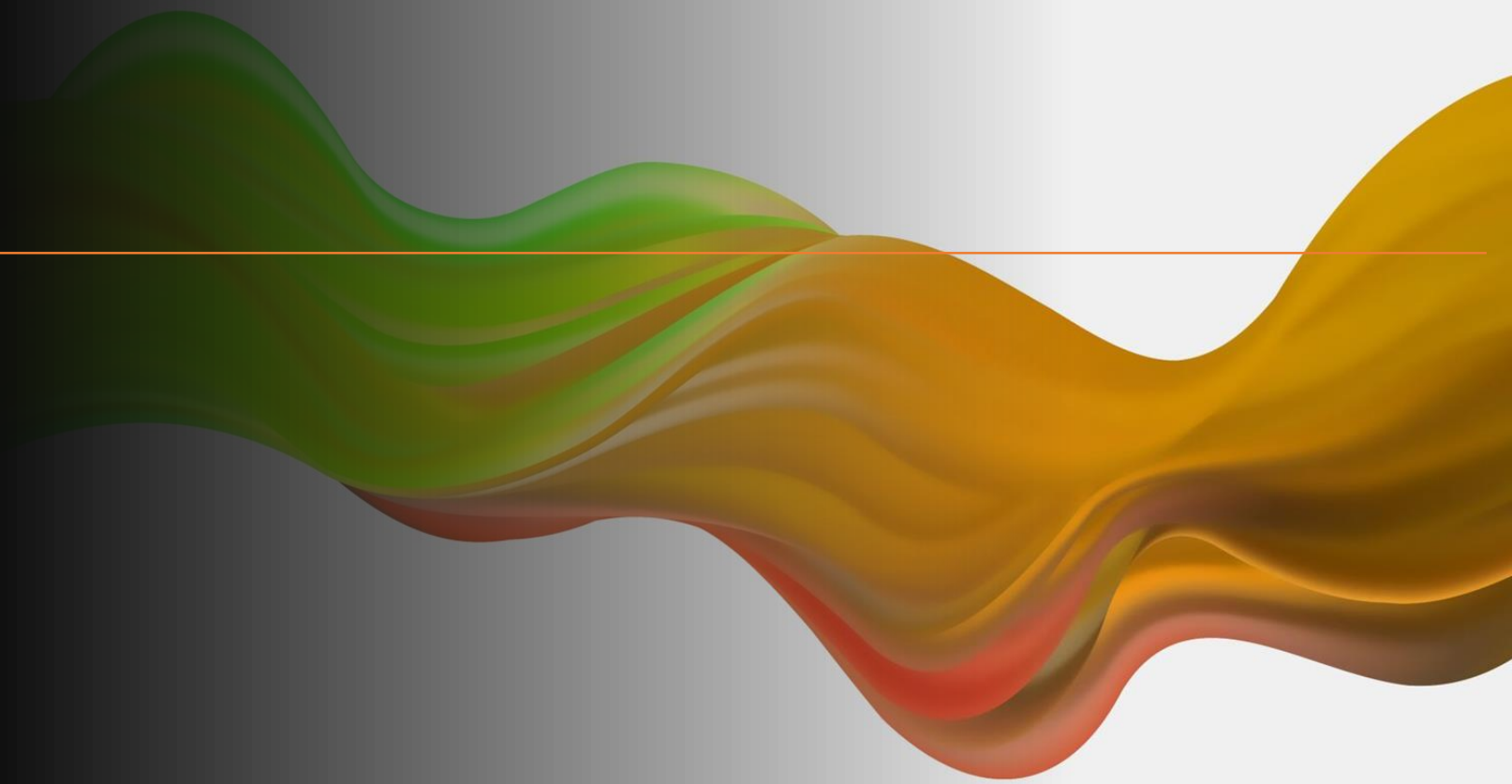
$$\vec{F} = \sum_{\text{surface}} p \vec{n} A = \oint p \vec{n} dA$$

$$\begin{aligned} \text{Lift} &= F_{\text{normal}} \\ \text{Drag} &= F_{\text{stream}} \end{aligned}$$

Calculate "Cl, Cd" w/ Differentiable Functions

Liwei Chen

2023-02-22

An abstract graphic consisting of several overlapping, wavy, ribbon-like shapes. The colors transition from dark green on the left to yellow and orange on the right, with some red and brown tones at the bottom. The shapes are fluid and organic, resembling liquid or smoke. A thin horizontal orange line is positioned above the graphic, intersecting it.

Differentiable Metrics

- Implemented and tested the calculation of C_l and C_d using differentiable functions.
- Here we provide some example codes for the calculation of metrics using differentiable operators (in pytorch).
- We are currently developing it for complex cases, such as „differentiable solver + NN“, learned simulator for unsteady aerodynamics with mesh deformations (we will publish it in the near future).

Example Codes

- readnpz_and_mesh--Torch.py



tfi_torch.py



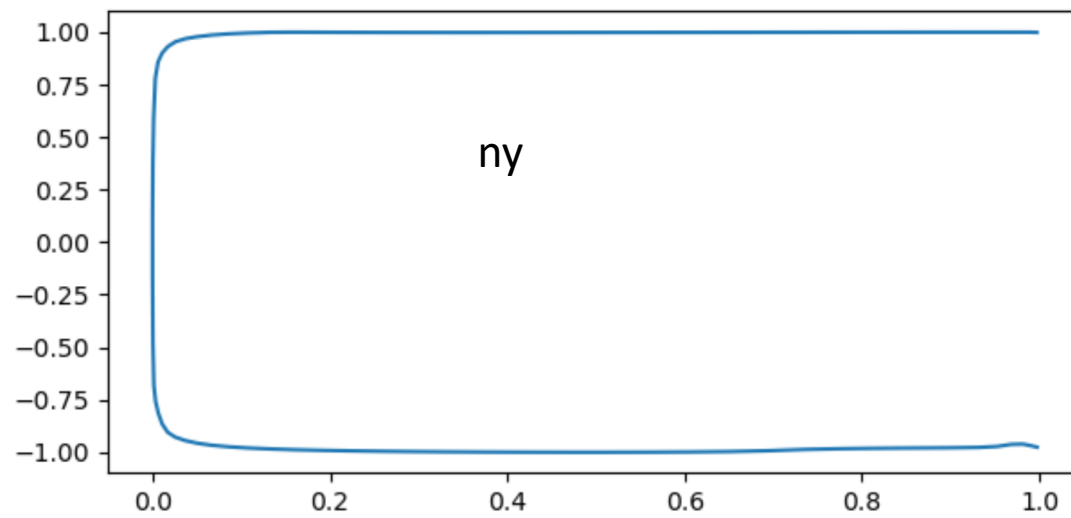
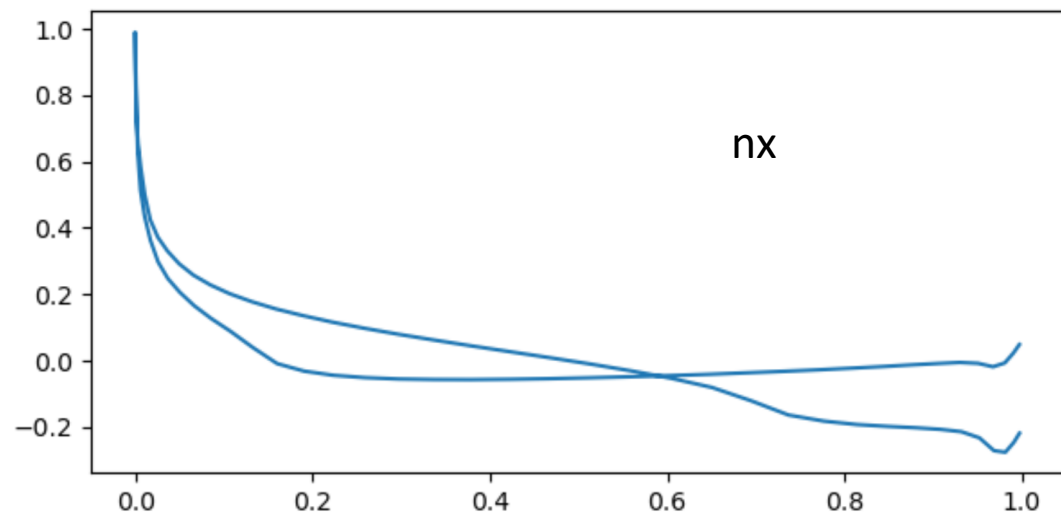
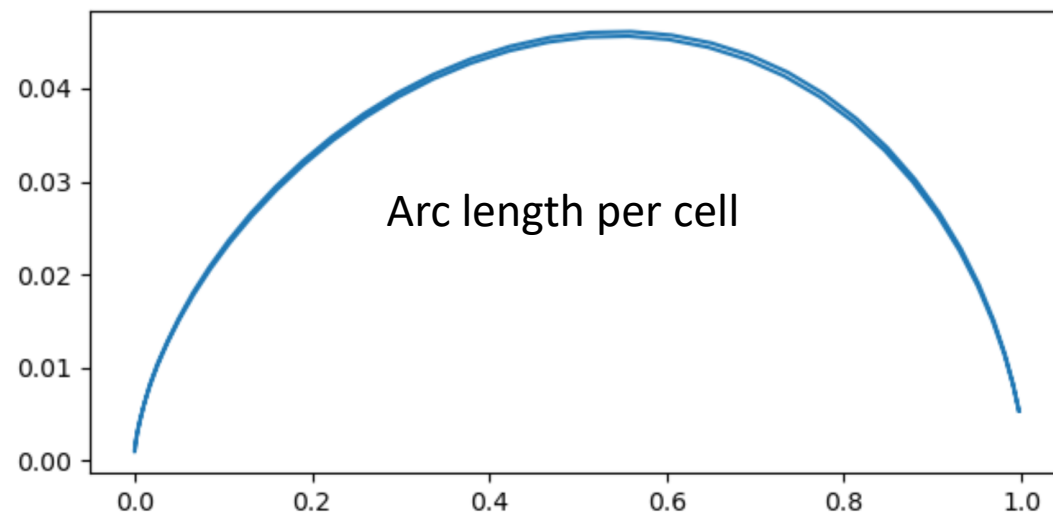
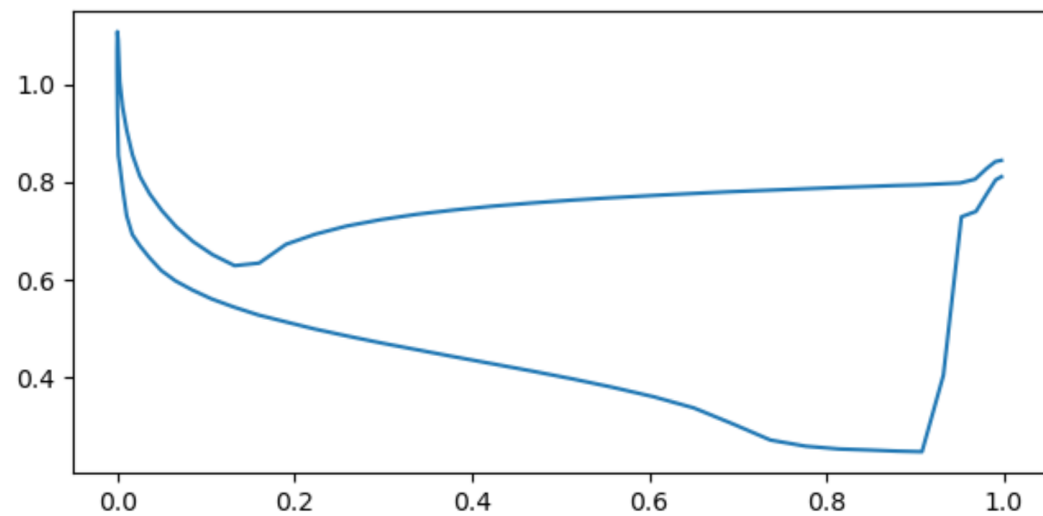
readnpz_and_mesh--Torch.py



global_variables.py

```
(base) liwei@PC05:~/Codes/coord-trans-encoding_inv/BASIC_simulations_inv$ python readnpz_and_mesh--Torch.py
Enter the file name: ah63k127_80_125_1000.npz
ah63k127
0.8 1.25 1.0
Successfully read grid file train_mesh_point/ah63k127.p3d
torch.Size([1, 1, 128, 129])
jst-2 tensor(0.0053, device='cuda:0')
jfn-1 tensor(0.0053, device='cuda:0')
wake_lower_jfn - 1 tensor(3.5774, device='cuda:0')
wake_upper_jfn - 2 tensor(3.5774, device='cuda:0')
jst-1 tensor(0.0053, device='cuda:0')
jfn-2 tensor(0.0053, device='cuda:0')
jst-1 1.0 0.0
jfn-1 1.0 0.0
(128, 128) (128, 128)
torch.Size([128, 128])
tensor([0.9476], device='cuda:0') tensor([0.1344], device='cuda:0') tensor([2.0387], device='cuda:0')
```

Cl, Cd from the ref. CFD: 0.9476330E+00 0.1343771E+00





tfi_torch.py



readnpz_and_mesh--Torch.py



global_variables.py

```

#imax, jmax, kmax, xc, yc, threed = read_grid("train_mesh/"+basename+".p3d")

_, _, _, xp, yp, threed = read_grid("train_mesh_point/"+basename+".p3d")
xc, yc = convert_center(xp, yp)

x_current, y_current = torch.from_numpy(xp).cuda(), torch.from_numpy(yp).cuda()
x_current = x_current.view(batch_size, 1, idim, jdim).type(torch.cuda.DoubleTensor)
y_current = y_current.view(batch_size, 1, idim, jdim).type(torch.cuda.DoubleTensor)

# now calculate the coordinate transformation metrics according to the paper Comput and Fluids
# ideally this should be done with "cell-point-type" grid files.
#dx_i, dy_i, ds_i, dx_j, dy_j, ds_j, area = calcMetrics(xp, yp)
dx_i, dy_i, ds_i, dx_j, dy_j, ds_j = calculateMetrics(x_current, y_current)
#pressure_center = calculateCellCenter(dynamics[:,istep,3], x_current, y_current)

# dx_i: the x-component of the unit normal-wise vector of i-face
# dy_i: the y-component of the unit normal-wise vector of i-face
# ds_i: the arc length of the cell along-i (circumferential)
# note: the face norm vec of i-face points into the airfoil (into the object)

# dx_j: the x-component of the unit normal-wise vector of j-face
# dy_j: the y-component of the unit normal-wise vector of j-face
# ds_j: the arc length of the cell along-j (normal)
# note: the face norm vec of j-face points clock-wise direction

x, y = xc, yc

# data[0] - xmach # in our current case,
# data[1] - aoa # these three numbers are the
# data[2] - re # same in all the samples.
# data[3] - rho ... density
# data[4] - rho*u ... density times X-velocity
# data[5] - rho*v ... density times Y-velocity
# data[6] - rho*E ... density times total energy (it is complex, but below you will see how I calculate pressure)

print(ds_i.shape)


```

```


215
216 def calculateMetrics(x, y):
217     '''
218     Input x, y
219     Outputs:
220     mesh edge-i unit normal vector x-component: dx_i
221     mesh edge-i unit normal vector y-component: dy_i
222     mesh edge-i arc length: ds_i
223     mesh edge-j unit normal vector x-component: dx_j
224     mesh edge-j unit normal vector y-component: dy_j
225     mesh edge-j arc length: ds_j
226     '''
227     #kernel_dx = torch.Tensor( [[-1, 0, 1],
228     #                           [-2, 0, 2],
229     #                           [-1, 0, 1]] ) * (1/8)
230     #kernel_dy = torch.Tensor( [[-1, -2, -1],
231     #                           [0, 0, 0],
232     #                           [1, 2, 1]] ) * (1/8)
233     bs = x.shape[0]
234     kernel_i = torch.Tensor( [[-1.0], [1.0]] ).cuda().type(torch.cuda.DoubleTensor) # 2x1
235     kernel_j = torch.Tensor( [[-1.0, 1.0]] ).cuda().type(torch.cuda.DoubleTensor) # 1x2
236     kernel_i = kernel_i.view((1,1,2,1))
237     kernel_j = kernel_j.view((1,1,1,2))
238     # the first dim: # of input channels
239     # the second dim: # of output channels
240     # the third and fourth dims: kernel size
241     #print(kernel_i, kernel_j)
242
243     # calculate tangential direction
244
245     t1 = F.conv2d(x, kernel_i, padding=0) # dx/di (or dx/dx1)
246     t2 = F.conv2d(y, kernel_i, padding=0) # dy/di (or dy/dx1)
247     #print(t1.shape, t2.shape)
248     #print(t1, t2)
249     ds_i = ((t1)**2 + (t2)**2)**0.5
250     ## normalize
251     #dx_i = torch.div(t1, ds_i)
252     #dy_i = torch.div(t2, ds_i)
253     # rotate -90deg to get normal, i.e. *(0 - 1*j)
254     #dx_i = torch.div(-t2, ds_i)
255     #dy_i = torch.div( t1, ds_i)
256     dx_i = t2/ds_i
257     dy_i = -t1/ds_i
258
259     #print(dx_i.shape, dy_i.shape)
260
261     # calculate tangential direction
262     tj1 = F.conv2d(x, kernel_j, padding=0)
263     tj2 = F.conv2d(y, kernel_j, padding=0)
264     ds_j = ((tj1)**2 + (tj2)**2)**0.5
265     ## normalize
266     #dx_j = torch.div(tj1, ds_j)
267     #dy_j = torch.div(tj2, ds_j)
268     # rotate -90deg to get normal, i.e. *(0 - 1*j)
269     #dx_j = torch.div( tj2, ds_j)
270     #dy_j = torch.div(-tj1, ds_j)
271     dx_j = tj2/ds_j
272     dy_j = -tj1/ds_j
273

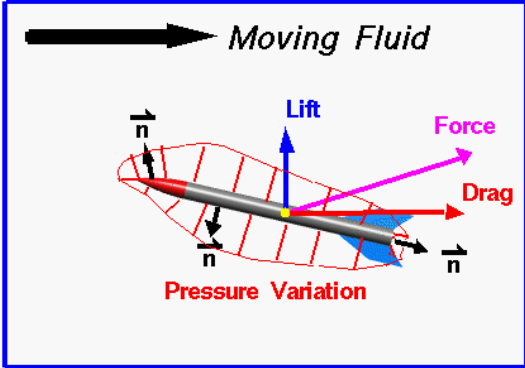
```

How to calculate Force-X and Force-Y



Aerodynamic Forces





Pressure forces act normal (perpendicular) to surface.
Force on the body is the vector sum of the pressure x area around the entire solid body.

$$\vec{F} = \sum_{\text{surface}} p \vec{n} A = \oint p \vec{n} dA$$

Lift = F_{normal}

Drag = F_{stream}

Project f_x , f_y onto the fluid-moving direction, so we need angle of attack.

```

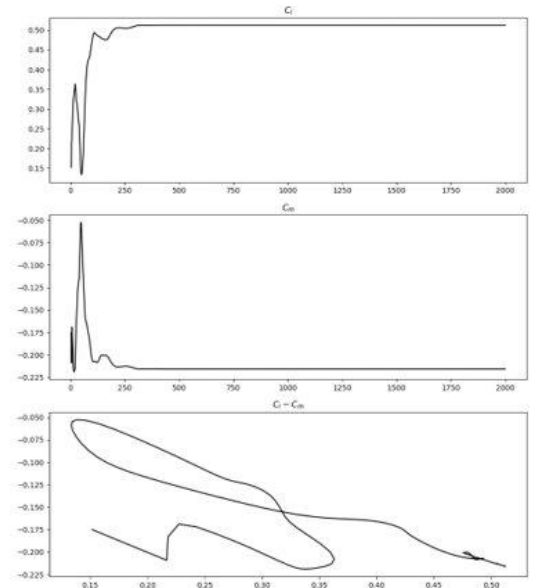
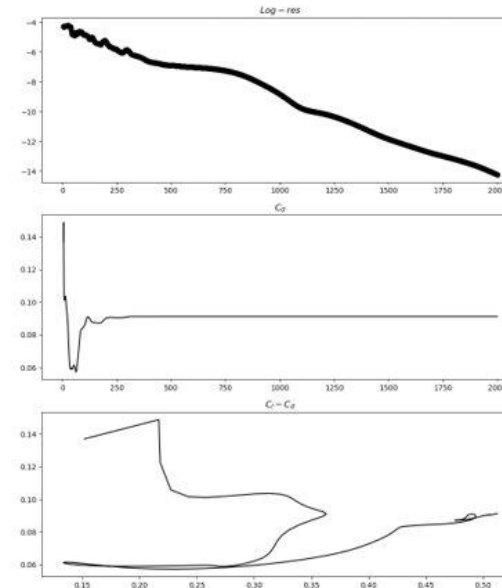
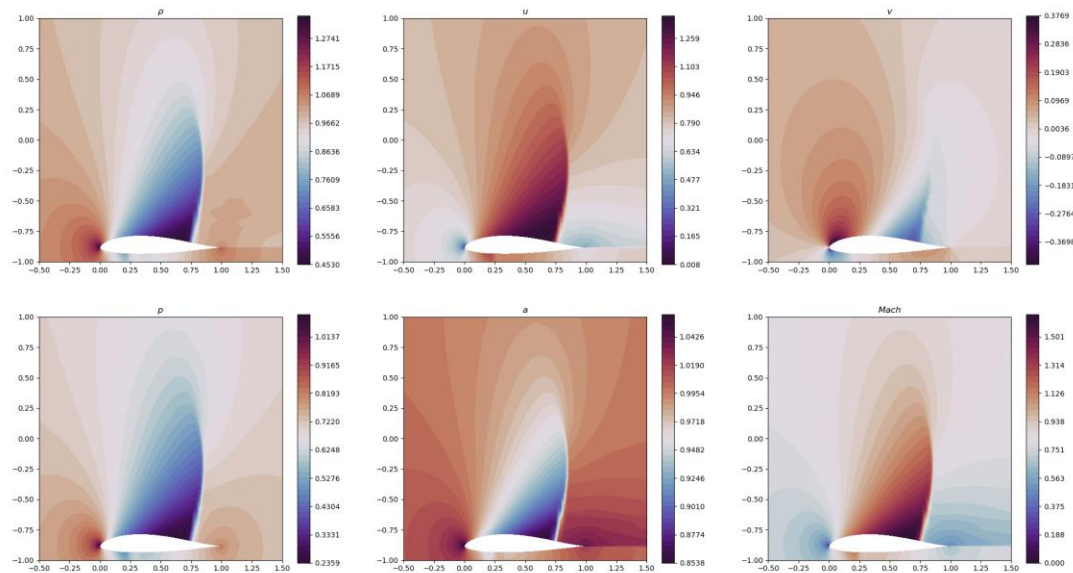
293
294 angle_of_attack = 1.25 #degree
295 angle_of_attack = angle_of_attack / 180. * np.pi
296 cosDir = np.cos(angle_of_attack)
297 sinDir = np.sin(angle_of_attack)
298
299
300 pressure_center = torch.from_numpy(p).cuda().type(torch.cuda.DoubleTensor)
301 #pressure_farfield_boundary = torch.from_numpy(p[:, -1]).cuda().type(torch.cuda.DoubleTensor)
302 #pressure_farfield_boundary = pressure_farfield_boundary.view(imax,1)
303 #print(pressure_farfield_boundary.shape)
304 #pressure_center = torch.cat((pressure_center, pressure_farfield_boundary), dim=1) # just make sure it has the same shape as ds and dx
305 print(pressure_center.shape)
306
307 xc_cuda = torch.from_numpy(xc).cuda().type(torch.cuda.DoubleTensor)
308 #yc_cuda = torch.from_numpy(yc).cuda().type(torch.cuda.DoubleTensor)
309 pressure_surface = torch.masked_select( pressure_center, cellcenterMask ).view(1,-1)
310 xc_surface = torch.masked_select( xc_cuda, cellcenterMask ).view(1,-1)
311 arcLength_surface = torch.masked_select( ds_i, segmentMask ).view(1,-1)
312 nx_surface = torch.masked_select( dx_i, segmentMask ).view(1,-1)
313 ny_surface = torch.masked_select( dy_i, segmentMask ).view(1,-1)
314
315 #print("inner_x:", inner_x.shape) # inner_x: torch.Size([5, 1, 256, 129])
316 #print("segmentMask:", segmentMask.shape) # segmentMask: torch.Size([5, 256, 129])
317 #print("mode_1a:", mode_1a.shape)
318 #print("test mask:", torch.masked_select(inner_x, segmentMask).view(batch_size,-1).shape)
319 #f_pred_1 = torch.sum( torch.masked_select(inner_x, segmentMask).view(batch_size,-1), dim=1 ) ##mode_1a
320 #f_pred_2 = torch.sum( torch.masked_select(inner_y, segmentMask).view(batch_size,-1), dim=1 ) ##mode_2a
321 #print(f_pred_1, f_pred_2)
322
323 #plt.figure()
324 fig, axs = plt.subplots(2, 2)
325 axs[0,0].plot(xc_surface[0].cpu().detach().numpy(), pressure_surface[0].cpu().detach().numpy())
326 axs[0,1].plot(xc_surface[0].cpu().detach().numpy(), arcLength_surface[0].cpu().detach().numpy())
327 axs[1,0].plot(xc_surface[0].cpu().detach().numpy(), nx_surface[0].cpu().detach().numpy())
328 axs[1,1].plot(xc_surface[0].cpu().detach().numpy(), ny_surface[0].cpu().detach().numpy())
329 plt.show()
330
331
332
333 dfx = (1.4*pressure_surface-1)*arcLength_surface*nx_surface/(0.5*0.8**2*1.4)
334 dfy = (1.4*pressure_surface-1)*arcLength_surface*ny_surface/(0.5*0.8**2*1.4)
335
336 fx = torch.sum( dfx, dim=1 )
337 fy = torch.sum( dfy, dim=1 )
338 # for sanity check, calculate the wet surface, theoretically it should 2.0
339 wet_surface = torch.sum( arcLength_surface, dim=1 )
340
341 Cd = fx*cosDir + fy*sinDir
342 Cl = -fx*sinDir + fy*cosDir
343
344
345 print(Cl, Cd, wet_surface)
    
```



In Progress: Surrogate Model

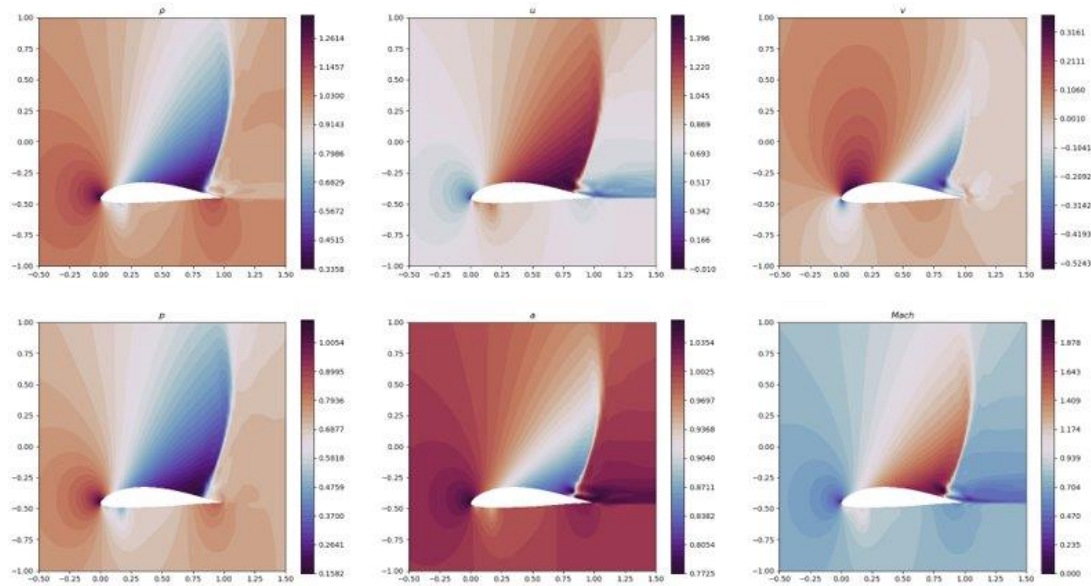
- <https://github.com/tum-pbs/coord-trans-encoding>
- Now we directly calculate metrics information by python script. No need Fortran solver anymore.

0.283, 0.087, -0.062



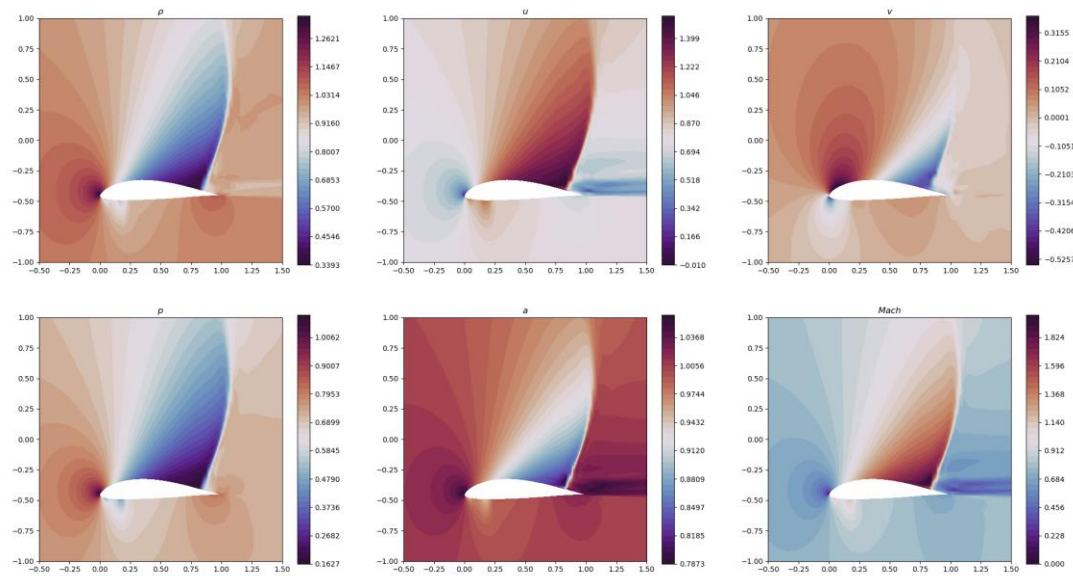
$C_L = 0.5123548E+00$
 $C_D = 0.9131187E-01$
 $C_m = -0.2159179E+00$

0.639, 0.043, -0.127



$C_l = 0.8308828E+00$
 $C_d = 0.1734693E+00$
 $C_m = -0.3984813E+00$

0.692, 0.162, -0.283



$C_l = 0.8255088E+00$

$C_d = 0.1721221E+00$

$C_m = -0.3907791E+00$



256 points



512 points



768 points



1024 points