1 Project 1

1.1 PDE

We get a parabolic equation because of the gradient descent flow

$$\begin{cases}
\frac{\partial u}{\partial t} = div \left(\frac{\nabla u}{\sqrt{u_x'^2 + u_y'^2 + u_z'^2 + \epsilon}} \right) - \lambda^*(u - f) \\
\frac{\partial u}{\partial \overrightarrow{n}} \Big|_{\partial \Omega} = 0 \\
u(x, 0) = f
\end{cases}$$
(1)

1.2 Numerical format

For convenience of writing, we define

$$D_x^{\pm}(u_{i,j,k}) \triangleq \pm (u_{i\pm 1,j,k} - u_{i,j,k}),$$

We proposed a numerical approximation of $|\nabla u|$

$$\sqrt{u_x'^2 + u_y'^2 + u_z'^2 + \epsilon} = \sqrt{D_x^+(u_{i,j,k})^2 + D_y^+(u_{i,j,k})^2 + D_z^+(u_{i,j,k})^2 + \epsilon}$$

And then, we have

$$div\left(\frac{\nabla u}{\sqrt{u_x'^2 + u_y'^2 + u_z'^2 + \epsilon}}\right) \approx D_x^- \left(\frac{D_x^+(u_{i,j,k}^n)}{|D_h(u_{i,j,k}^n)|}\right) + D_y^- \left(\frac{D_y^+(u_{i,j,k}^n)}{|D_h(u_{i,j,k})^n)|}\right) \triangleq z_{i,j,k}^n$$
(2)

In addition, we discrete time using forward Euler using forward Euler method. Therefore, the numerical format of parabolic equation (1) is obtained as follow

$$u_{i,j,k}^{n+1} = u_{i,j,k}^n + \tau z_{i,j,k}^n - \tau \lambda (u_{i,j,k}^n - u_{i,j,k}^0), \quad \text{where } \tau = t^{n+1} - t^n.$$
(3)

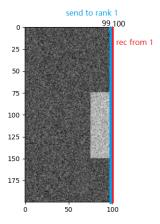
1.3 Distribution and communication

Data distribution:

```
if rank == 0:
    # Generate image
...
    # Adding Gaussian noise
...
    # Data distribution
    nimgsile = nimg[:,:,0:101]
    sendbuf = nimg[:,:,99:200].copy()
    comm.Send(sendbuf, dest=1, tag=11)
    del nimg, sendbuf
else:
    nimgsile = np.empty([200,200,101],dtype=np.float64)
    comm.Recv(nimgsile, source=0, tag=11)
```

Communication:

```
for t in range(T):
 if rank ==0 and not t%5:
   print(t, 'of ', T)
  # In-process computation
 J = worker(nimgsile, J, dt, lam)
 # Blocking communication
 if rank == 0:
   # rank 0: send before recv
   sendbuf = J[:,:,n2-2].copy()
   comm.Send(sendbuf, dest=1, tag=100)
   recbuf = np.empty(J[:,:,n2-1].shape, dtype=J[:,:,n2-1].dtype)
   comm.Recv(recbuf, source=1, tag=110)
   J[:,:,n2-1] = recbuf
  else:
   # rank 1: recv before send
   recbuf = np.empty(J[:,:,n2-1].shape, dtype=J[:,:,n2-1].dtype)
   comm.Recv(recbuf, source=0, tag=100)
   J[:,:,0] = recbuf
   sendbuf = J[:,:,1].copy()
   comm.Send(sendbuf, dest=0, tag=110)
```



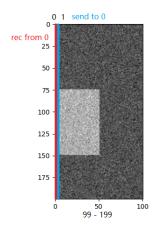


Figure 1: Process 0

Figure 2: Process 1

1.4 In-process computation improvement

Original: Matrix operations (faster than for) but increased 5 times RAM.

```
def worker(In, J, dt, lam):
    ep = 1e-4
    m,n,l = J.shape
# Gradient approximates by forward Euler
DfJx=J[list(range(1,m))+[m-1],:,:]-J
DfJy=J[:,list(range(1,n))+[n-1],:]-J
DfJz=J[:,:,list(range(1,l))+[l-1]]-J
# \varphi(\nabla u)
TempDJ=(ep+DfJx*DfJx+DfJy*DfJy*DfJz*DfJz)**(1/2)
DivJx=DfJx/TempDJ
DivJy=DfJy/TempDJ
DivJz=DfJz/TempDJ
```

```
Div=bdx(DivJx,m)+bdy(DivJy,n)+bdz(DivJz,1)
J += dt * Div -dt*lam*(J-In)
return J
```

Improvement: Divided into 8 parts along the x direction, each parts have 25+1 or 25+2 layers. RAM is shown in Figure 3.

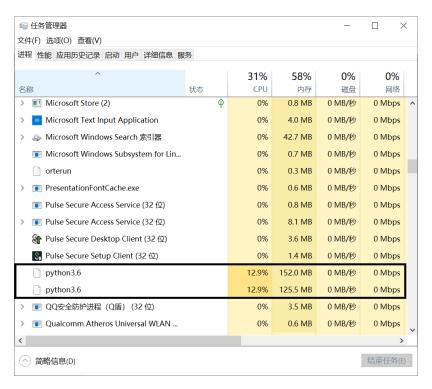


Figure 3: RAM Improvement

1.5 Results

Results collection and presentation:

```
if rank ==0:
    # Process0 collect data and plt
    deimg = np.empty([nx,ny,nz], dtype = np.float64)
    deimg[:,:,0:100] = J[:,:,0:100]
    recbuf = np.empty([200,200,100],np.float64)
    comm.Recv(recbuf, source=1, tag=20)
    deimg[:,:,100:200] = recbuf
    plt.figure()
    plt.imshow(deimg[100,:,:],"gray")
    ...
else:
    sendbuf = J[:,:,1:101].copy()
    comm.Send(sendbuf, dest=0, tag=20)
```

Results is shown Figure 4-6. (Code see **TV3d2.py**)

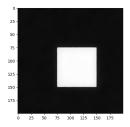


Figure 4: y-z Section

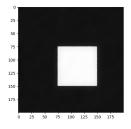


Figure 5: x-z Section

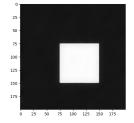


Figure 6: x-y Section