

# Entropy stable notes

Jesse Chan

## 1 Curved meshes

Non-affine geometric factors

Let  $\mathbf{F}(\mathbf{U})$  denote the flux matrix whose rows are

$$\mathbf{F}(\mathbf{U}) = \begin{pmatrix} \mathbf{F}_x \\ \mathbf{F}_y \end{pmatrix}.$$

The conservation law we're interested in is the following

$$\frac{\partial \mathbf{U}}{\partial t} + \nabla \cdot \mathbf{F}(\mathbf{U}) = 0,$$

where the divergence is taken over each column of  $\mathbf{F}(\mathbf{U})$ .

Let  $\mathbf{G}$  denote the Jacobian of the geometric mapping

$$\mathbf{G}_{ij} = \frac{\partial \hat{\mathbf{x}}_j}{\partial \mathbf{x}_i},$$

and let  $J$  denote the determinant of  $\mathbf{G}$ . One can show (Kopriva?) that

$$\hat{\nabla} \cdot (J\mathbf{G}^T) = 0.$$

We also have that the mapped normals obey the property

$$J\mathbf{G}\hat{\mathbf{n}}\hat{J}^f = \mathbf{n}J^f.$$

At the continuous level, the physical gradient and divergence satisfy

$$(\nabla u, \mathbf{v})_{D^k} = (J\mathbf{G}\hat{\nabla}u, \mathbf{v})_{\hat{D}}, \quad (\nabla \cdot \mathbf{u}, v)_{D^k} = (\hat{\nabla} \cdot (J\mathbf{G}^T \mathbf{u}), v)_{\hat{D}},$$

as well as a corresponding integration by parts property.

## 2 Flux differencing

Replace the flux derivative with

$$(\nabla \cdot \mathbf{F}_S(\mathbf{U}(\mathbf{x}), \mathbf{U}(\mathbf{x}'))|_{\mathbf{x}'=\mathbf{x}})_{D^k}.$$

It is possible to remove the effect of geometric aliasing by evaluating the above term via

$$\left( \hat{\nabla} \cdot \mathbf{F}_S^k(\mathbf{U}(\mathbf{x}), \mathbf{U}(\mathbf{x}')) \Big|_{\mathbf{x}'=\mathbf{x}} \right)_{\hat{D}}, \quad \mathbf{F}_S^k = \{\{J\mathbf{G}\}\} \mathbf{F}_S(\mathbf{U}(\mathbf{x}), \mathbf{U}(\mathbf{x}'))$$

We can evaluate the new geometrically de-aliased flux using the discrete gradient of Chen and Shu

$$(\nabla_h \cdot \mathbf{u}, v)_\Omega = \sum_k \left( \hat{\nabla} \cdot J\mathbf{G}^T \Pi_N \mathbf{u}, v \right)_{\hat{D}} + \frac{1}{2} \langle J\mathbf{G}^T \mathbf{u}^+ - \Pi_N J\mathbf{G}^T \mathbf{u}, v \hat{\mathbf{n}} \rangle_{\partial \hat{D}} + \frac{1}{2} \langle J\mathbf{G}^T \mathbf{u} - \Pi_N J\mathbf{G}^T \mathbf{u}, \Pi_N (v \hat{\mathbf{n}}) \rangle_{\partial \hat{D}}$$

Another alternative is to treat curvilinear grids using a direct skew-symmetric form.

### 3 Limiting

Ignoring high order terms in the Taylor expansion, we have

$$u(\Pi_N v) - u = \frac{\partial u}{\partial v} (\Pi_N v - v).$$

We want to limit  $u$  with a compression towards the mean  $u = \bar{u} + \theta(u - \bar{u})$  such that  $|u(\Pi_N v)| \leq C|u|$ .

#### 3.1 Option 1

One option: ensure  $|u(\Pi_N v) - u(\bar{v})| \approx |u - u(\bar{v})|$  based on Taylor series.

$$u(\Pi_N v) - u(\bar{v})$$

Limit  $\tilde{u} = \bar{u} + \theta(u - \bar{u})$  such that

$$\min \Pi_N v \geq \min v, \quad \max \Pi_N v \leq \max v.$$

Another option: estimate or filter  $\Pi_N v$

#### 3.2 Other options

- Adaptively choose time-step size?
- Expensive option: bisection-type algorithm for finding theta.