

# Scalable Riemann Solvers with the Discontinuous Galerkin Method for Hyperbolic Network Simulation

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# Hyperbolic Conservation Laws on a Network

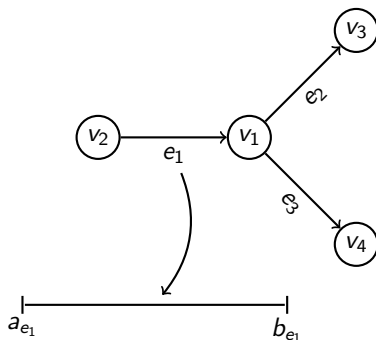
A **network** is a topological graph, i.e a couple  $(V, E)$  where

- ▶  $E$  is a collection of intervals  $[a_e, b_e]$
- ▶  $V$  a collection of vertices connecting the intervals

A system of conservation laws naturally extends to a network edgewise

$$\partial_t u_e + \partial_x f(u_e) = 0,$$

- ▶  $u_e \in \mathbb{R}^m$
- ▶  $f : \mathbb{R}^m \rightarrow \mathbb{R}^m$
- ▶ Only considering  $m = 1, 2$
- ▶ Challenge is coupling the 1D problems at vertices



# Conservation Laws: Examples

## Shallow Water (SWE):

$$\partial_t h + \partial_x(h\nu) = 0,$$

$$\partial_t h\nu + \partial_x\left(h\nu^2 + \frac{g}{2}h^2\right) = 0,$$

- ▶  $h(x, t)$  : water height
- ▶  $\nu(x, t)$  : water velocity

The Jacobian of the flux function has eigenvalues:

$$\lambda_1(u) = \nu - \sqrt{gh}, \quad \lambda_2(u) = \nu + \sqrt{gh}$$

The system is *fluvial* if  $|\nu| < \sqrt{gh}$ ,  
 $\lambda_1 < 0$ ,  $\lambda_2 > 0$ .

## LWR Traffic Flow

$$\partial_t \rho + f(\rho)_x = 0$$

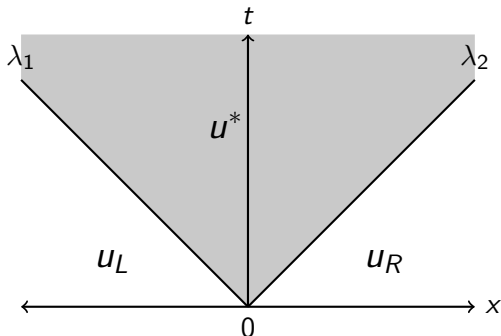
- ▶  $\rho(x, t)$  : density of cars
- ▶  $\rho \in [0, 1]$  by normalization
- ▶  $f$  is  $C^2$  and strictly concave
- ▶  $f(0) = f(1) = 0$
- ▶ for my examples  
 $f(\rho) = 4\rho(1 - \rho)$

## 1D Riemann Problem

A Riemann problem for a 1D hyperbolic conservation law is the PDE with a jump initial condition

$$u_0(x) = \begin{cases} u_L, & x < 0, \\ u_R, & x > 0, \end{cases}$$

For  $2 \times 2$  systems solutions take the form

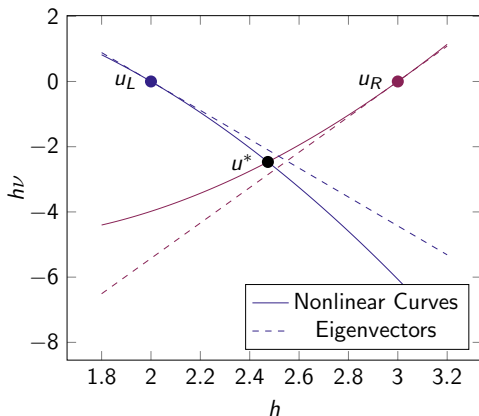


## 1D Riemann Problem II

The star state  $u^*$  is found by:

- ▶ finding the intersection of two nonlinear curves in phase space
- ▶ the curves are points that can be connected to  $u_L$ ,  $u_R$  by shock/rarefaction waves.

For the shallow water equations these solutions look like



# Riemann Problem at a Vertex

## Riemann Problem at a Vertex $v^a$

$$\partial_t u_e + \partial_x f(u_e) = 0, \quad t \in \mathbb{R}^+, e \in E(v)$$

$$u_e(x, t = 0) = \bar{u}_e, \quad x \in \begin{cases} \mathbb{R}^+ & e \text{ outgoing} \\ \mathbb{R}^- & e \text{ incoming} \end{cases}$$

where  $\bar{u}_e$  are constant states.

- ▶ well-posedness requires an additional  $\deg(v)$  coupling conditions
- ▶ these are additional modeling choices
- ▶ For SWE a choice<sup>b</sup>

$$h_e^* = h_{e'}^* \quad \text{height continuity}$$

$$\sum_{\substack{e \in E(v), \\ \text{incoming}}} \nu_e^*(v) = \sum_{\substack{e \in E(v), \\ \text{outgoing}}} \nu_e^*(v) \quad \text{conservation of mass}$$

- ▶ **Fundamental principle:** Connect each  $\bar{u}_e$  to a  $u_e^*$  by a single wave-curve, that is outgoing from the vertex.

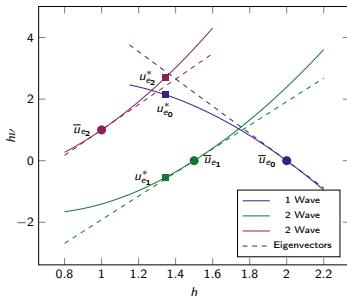


Figure: Example SWE Riemann problem for a 3-branch vertex.

<sup>a</sup>Colombo, Herty, and Sachers, “On  $2 \times 2$  Conservation Laws at a Junction”.

<sup>b</sup>Briani, Benedetto Piccoli, and Qiu, “Notes on RKDG Methods for Shallow-Water Equations in

# LWR Riemman Problem

Model is an optimization problem<sup>1</sup>

- ▶ Consider a traffic distribution matrix  $A \in \mathbb{R}^{\text{outdeg}(v) \times \text{indeg}(v)}$ , where  $A\mathbb{1} = \mathbb{1}$ 
  - ▶ Dictates what proportion of cars moving into a vertex from incoming road  $j$ , go to outgoing road  $i$
- ▶ Let  $\sigma$  be the point that  $f$  attains its max, define function  $\gamma^{max}$   
for incoming roads  $j$  for outgoing roads  $i$

$$\gamma_j^{max}(\rho) = \begin{cases} f(\rho), & \rho \leq \sigma \\ f(\sigma), & \rho > \sigma. \end{cases}$$

$$\gamma_i^{max}(\rho) = \begin{cases} f(\sigma), & \rho \leq \sigma \\ f(\rho), & \rho > \sigma. \end{cases}$$

- ▶ Full problem, maximize flux through the vertex


$$\max f(\rho_j^*)$$

$$f(\rho_j^*) \in [0, \gamma_j^{max}(\rho_j)] \text{ for } j \text{ incoming}$$

$$A(f(\rho_j^*)) \in [0, \gamma^{max}(\rho_0)] \times \dots \times [0, \gamma^{max}(\rho_{\text{outdeg}(v)})]$$

- ▶ Solve in PETSc using TAO ALLM

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<sup>1</sup>Garavello and B. Piccoli, *Traffic Flow on Networks: Conservation Laws Models*. 



# Discretization

1. Discretize along edges using the standard RKDG framework.

- ▶ Per element  $I_j$  evolve

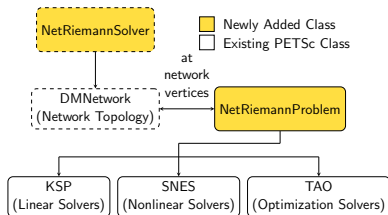
$$\int_{I_j} v \partial_t u_h dx = \int_{I_j} f(u_h) \partial_x v dx - \left( \hat{f}_{j+\frac{1}{2}} v^-|_{x_{j+\frac{1}{2}}} - \hat{f}_{j-\frac{1}{2}} v^+|_{x_{j-\frac{1}{2}}} \right),$$

where  $v, u_h \in P^k(I_j)$ , and  $\hat{f}$  is the numerical flux.

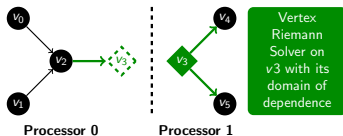
- ▶ Evolve with explicit SSP RK methods (TSSSP)
  - ▶ Use a characteristic-wise TVB slope limiter for stability (TSPostStage)
2. Solve network Riemann problem at vertices to get boundary fluxes.

# Implementation Overview

- ▶ **DMNetwork:**  
manage distributed network
- ▶ **NetRiemannProblem:**  
specify and solve *local* vertex Riemann problems
- ▶ **NetRiemannSolver:**  
scalably solve a collection of NetRiemannProblems on vertices of a DMNetwork.



## Distributed Vertex Riemann Solver



## NetRiemannProblem: Solver Reuse

- ▶ **The problem:** Every vertex Riemann problem requires some sort of PETSc solver, SWE (SNES), Traffic (TAO).
- ▶ Thankfully, Riemann problem structures (Jacobians etc..) only depend on the vertex degree.
- ▶ **Solution** Solver objects can be cached and reused. Have maps:
  - ▶ (SWE)  $deg(v) \rightarrow SNES$
  - ▶ (Traffic)  $(indeg(v), outdeg(v)) \rightarrow TAO$
- ▶ Can reuse symbolic solves in direct LU factorization

# NetRiemannSolver: How it's used

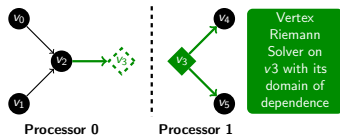
## Setup

1. Given a DMNetwork, it creates a cloned copy
2. NetRiemannProblem objects can be assigned to the vertices of DMNetwork
3. NetRiemannSolver generates a PetscSection and corresponding Vec for the Riemann problems at each assigned vertex.

## Solve

1. The *local* Riemann data for are set into the Vec (NetRiemannSolver handles indexing, not PetscSection)
2. NetRiemannSolver then solves all Riemann problems
  - ▶ basic interlacing, solve local Riemann problems while waiting for communication of nonlocal Riemann data
  - ▶ optional SolveBegin; ... SolveEnd; interface for further interlacing

Distributed Vertex Riemann Solver



# RHS Evaluation Algorithm

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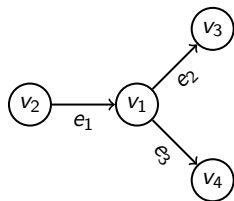
## Algorithm 1 DG RHS Evaluation for a Hyperbolic Network

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```
for vertex  $v$  in DMNetwork do
  for  $e \in E(v)$  do
    Evaluate  $u_e(v)$ 
    Place  $u_e(v)$  in NetRiemannSolver
  end for
end for
NetRiemannSolverSolveBegin()                                ▷ Communication here
for Edge  $e$  in DMNetwork do
  for Cell  $c$  in edge  $e$  do
    DG update (not vertex fluxes)
  end for
end for
NetRiemannSolverSolveEnd()                                  ▷ Communication here
▷ All vertex fluxes  $\hat{f}_v$  are now available
for Edge  $e$  in DMNetwork do
  Update boundary fluxes using  $\hat{f}_v$ 
end for
```

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# SWE Convergence Test



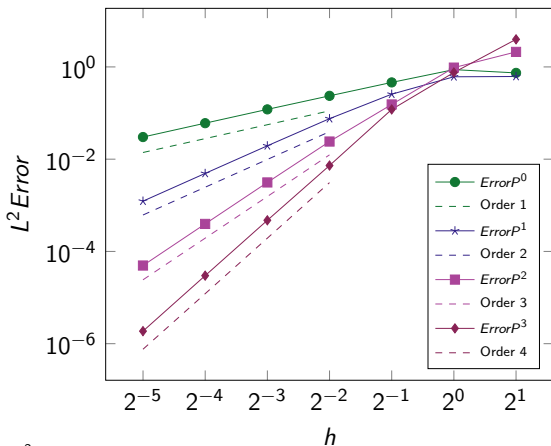
## Initial Data

$$h_1(x, 0) = 1 + e^{-5(x-9)^2},$$

$$h_1 v_1(x, 0) = h_1(x, 0)/2,$$

$$h_2(x, 0) = h_3(x, 0) = 1 + e^{-5(10-9)^2},$$

$$h_2 v_2(x, 0) = h_3 v_3(x, 0) = h_2(x, 0)/4.$$



# Dam Break on Shallow Water Network

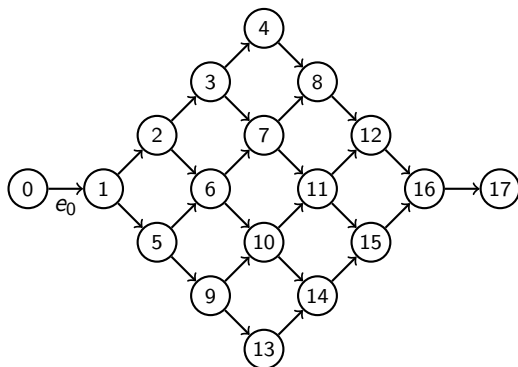


Figure: Grid graph variant used in the dam break test

# Traffic Network Example

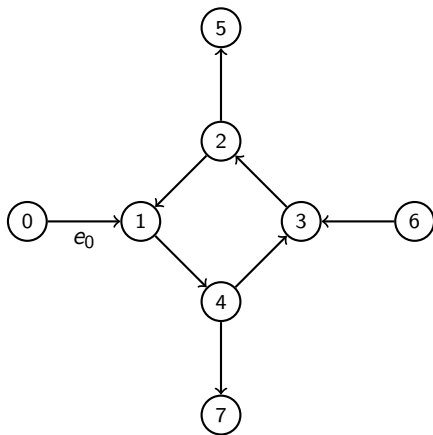


Figure: Traffic Circle Network

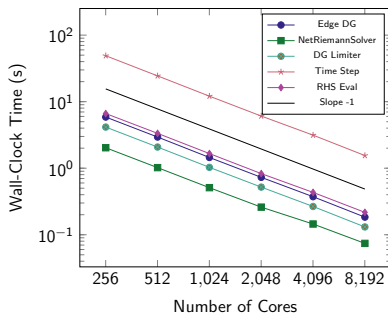


# Scalable Simulation of Mississippi River Network

- ▶ Simulate SWE on the Mississippi river system, with data obtained from the NHDPlus dataset<sup>a</sup>
- ▶ contains 892,740 edges and 872,300 vertices
- ▶ Discretize with elements of length 10m (the resolution of the dataset),  $P^2$  basis and a 5-stage SSPRK2 method<sup>b</sup>, with  $\approx 10^9$  degrees of freedom.
- ▶ Simulation conducted on the Theta supercomputer at ALCF. Theta has 4,392 nodes, each with 64 1.3GHz Intel Xeon Phi 7230 cores with 16 GiB of MCDRAM per node.

<sup>a</sup>McKay et al., "NHDPlus Version 2: user guide".






<sup>b</sup>Ketcheson, "Highly Efficient Strong Stability-Preserving Runge-Kutta Methods with Low-Storage Implementations".



# Future Work

1. Implement Blood Flow Networks
2. Large scale simulation for traffic flow
3. General code improvements. Move out of a PETSc branch into its own separate thing.
4. Move various pieces of my work into PETSc, particular DMNetwork improvements.

# References

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