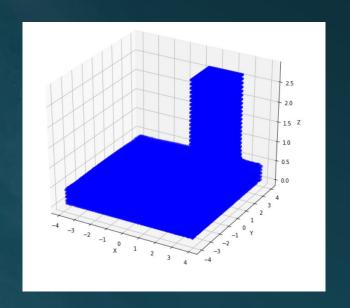
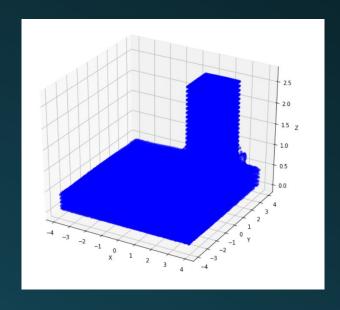
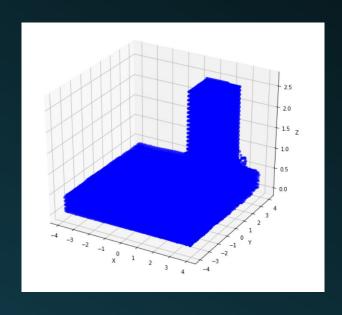


# Deep learning on fluid simulation

## Progress





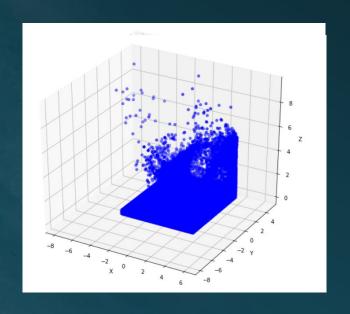


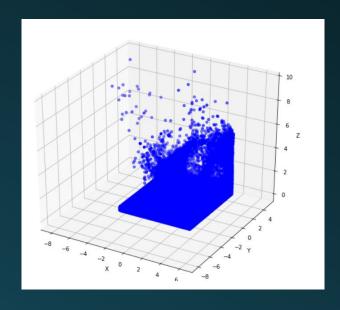
当前帧

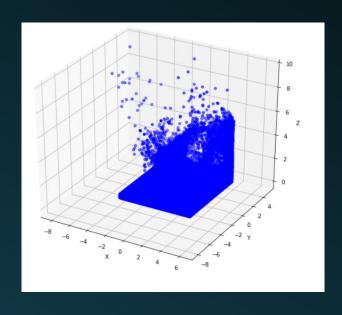
真实输出

预测输出

## Progress





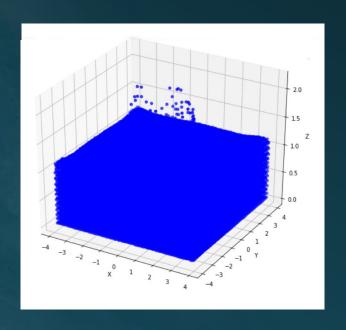


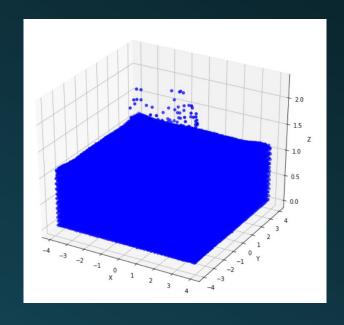
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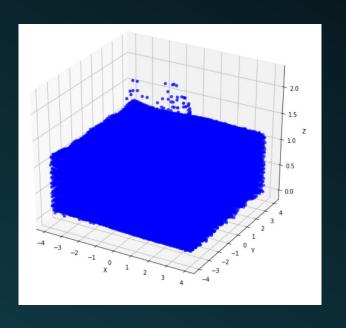
真实输出

预测输出

## Progress







当前帧

真实输出

预测输出

## Challenge

• 1. Make the model understand the input data

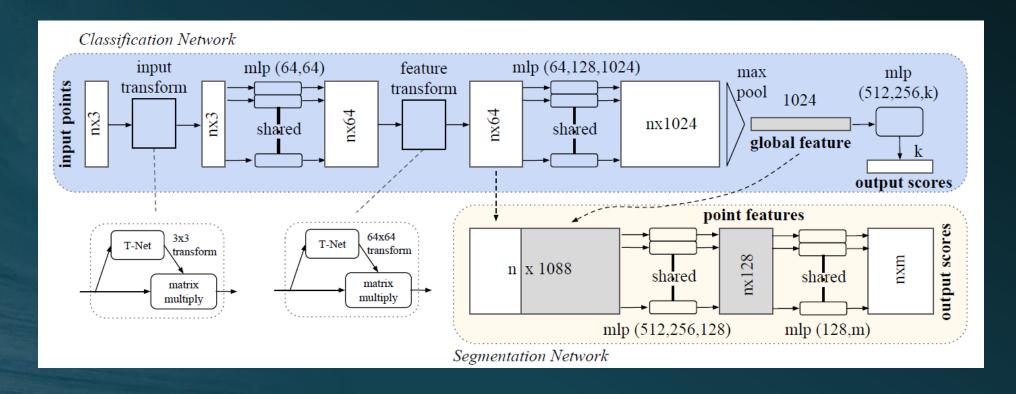
• 2. Make the model learn fluid simulation

## Challenge

- 1. Make the model understand the input data
  - ◆ PointNet
  - ♦ PointNet++
  - ◆ VoxelNet

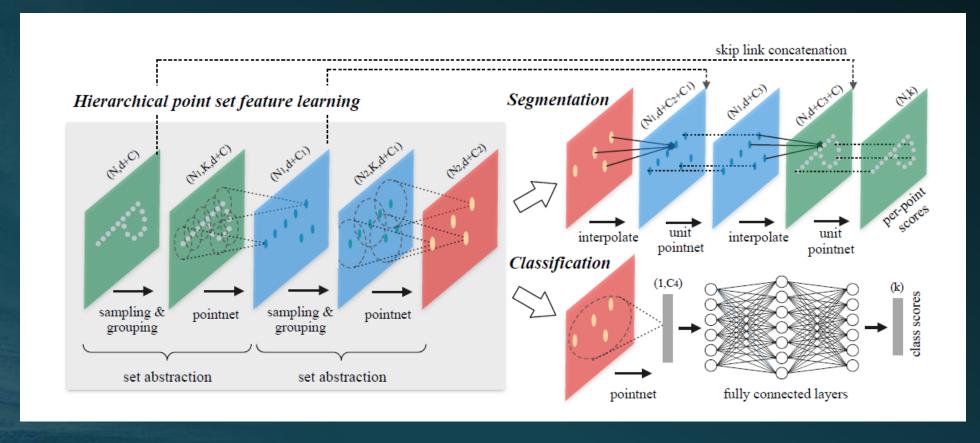
- ◆ PointNet
  - ✓ Unordered
  - ✓ Interaction among points
  - ✓ Invariance under transformations

#### ◆ PointNet



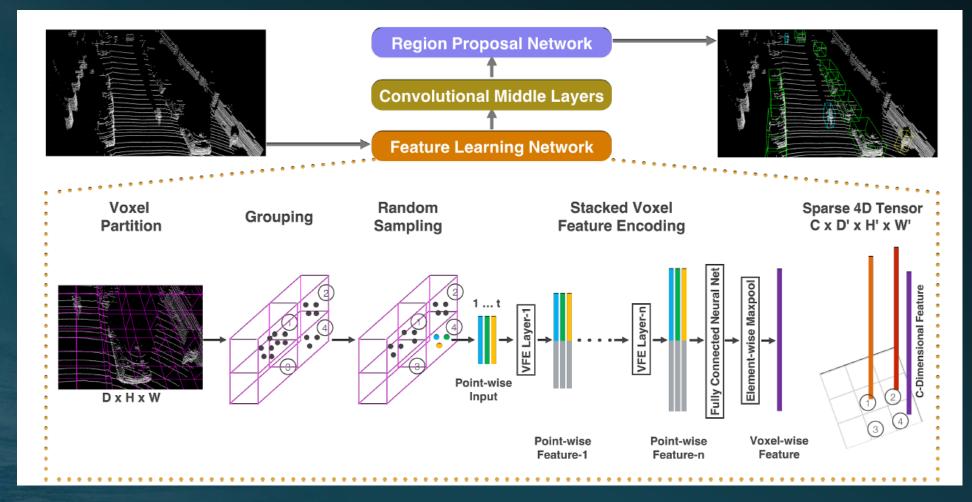
Qi C R, Su H, Mo K, et al. Pointnet: Deep learning on point sets for 3d classification and segmentation[J]. Proc. Computer Vision and Pattern Recognition (CVPR), IEEE, 2017, 1(2): 4.

◆ PointNet++

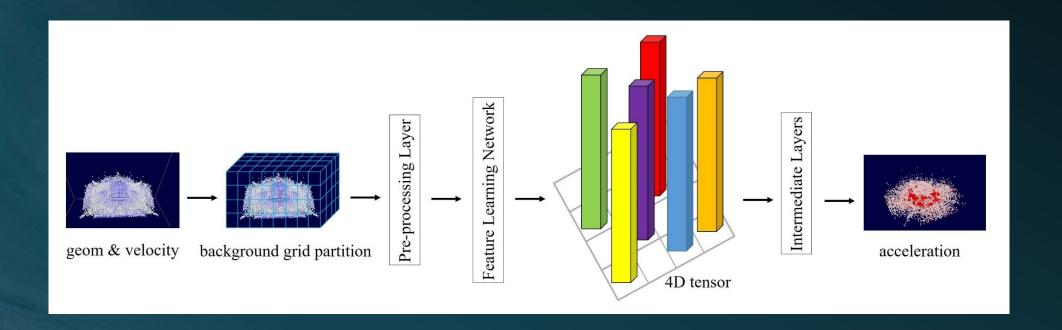


Qi C R, Yi L, Su H, et al. Pointnet++: Deep hierarchical feature learning on point sets in a metric space[C]//Advances in Neural Information Processing Systems. 2017: 5099-5108.

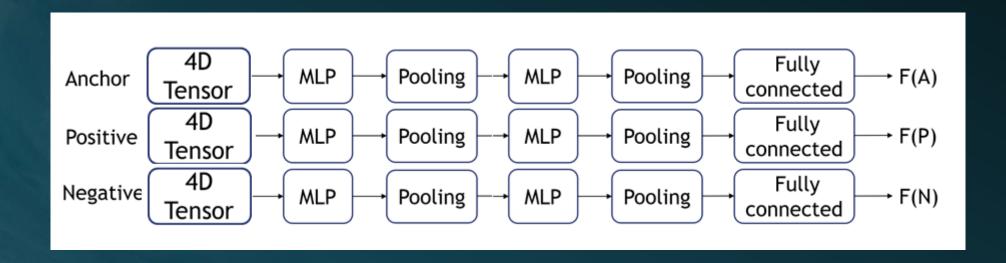
◆ VoxelNet



#### Our model



### Does the neural network working?



$$L(A, P, N) = \max(\|f(A) - f(P)\|^2 - \|f(A) - f(N)\|^2 + \alpha, 0)$$

## Challenge

- 2. Make the model learn fluid simulation
  - Accelerating Eulerian Fluid Simulation With Convolutional Networks
  - Data-Driven Synthesis of Smoke Flows with CNN-based Feature Descriptors
  - Generating Liquid Simulations with Deformation-aware Neural Networks
  - tempoGAN: A Temporally Coherent, Volumetric GAN for Super-resolution
    Fluid Flow
  - Latent-space Physics Towards Learning the Temporal Evolution of Fluid Flow

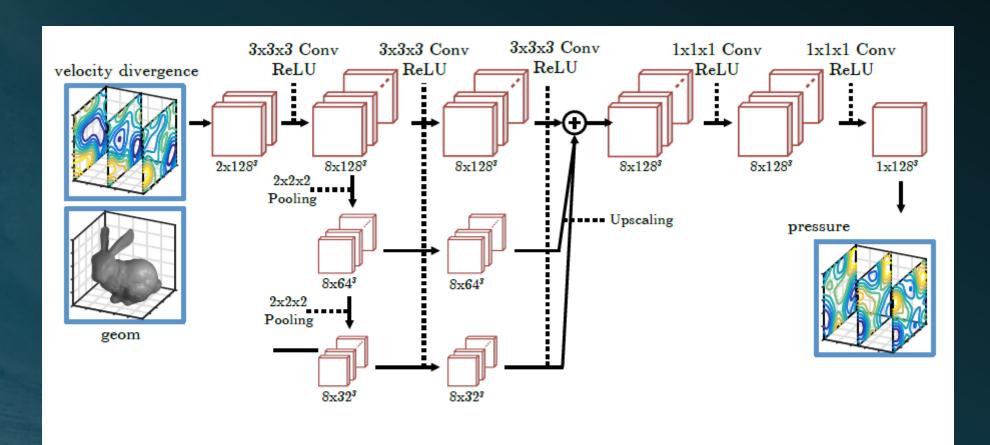
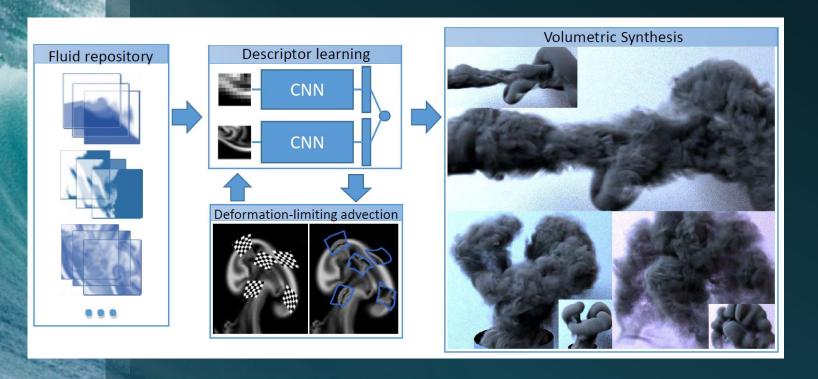
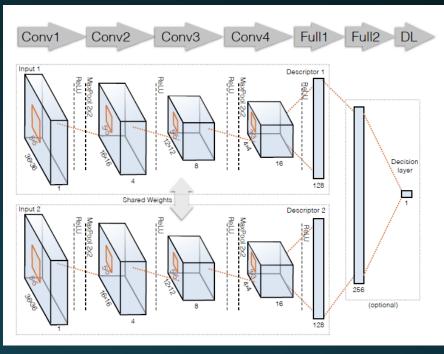


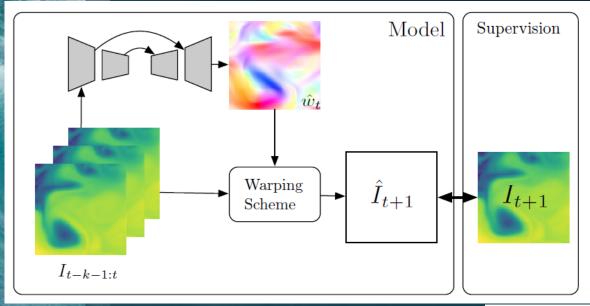
Figure 3. Convolutional Network for Pressure Solve

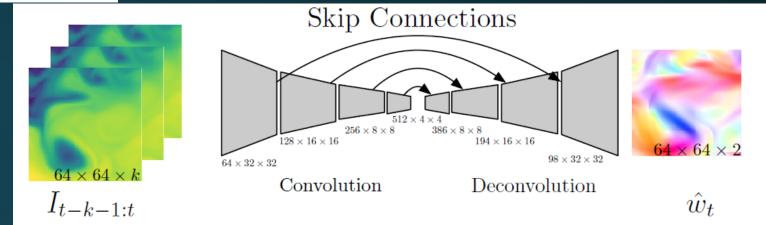




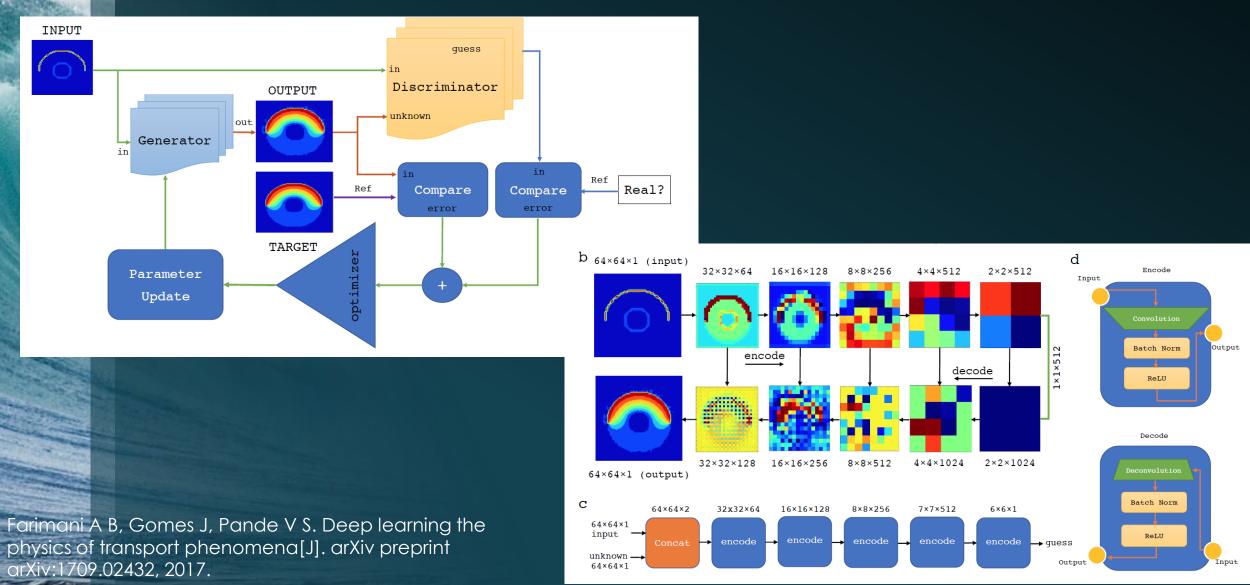
Chu M, Thuerey N. Data-driven synthesis of smoke flows with CNN-based feature descriptors[J]. ACM Transactions on Graphics (TOG), 2017, 36(4): 69.

- Physical learning
  - Deep Learning for Physical Processes: Incorporating Prior Scientific Knowledge
  - Deep learning the physics of transport phenomena

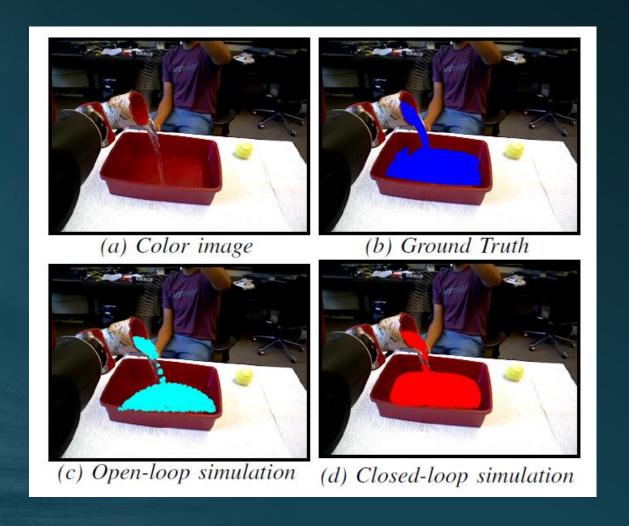




de Bezenac E, Pajot A, Gallinari P. Deep Learning for Physical Processes: Incorporating Prior Scientific Knowledge[J]. arXiv preprint arXiv:1711.07970, 2017



#### Analyze video data



Schenck C, Fox D. Reasoning about liquids via closed-loop simulation[J]. arXiv preprint arXiv:1703.01656, 2017.

#### Plan

- 1. Prove that the network is working and useful
- 2. Combine model with physical prior scientific knowledge
- 3. Learn the temporal evolution of fluid simulation