

Learning to Simulate Water Wave Packet

Jinning Li, 515030910592

Lianmin Zheng, 515030910117

Content

- Water wave packet
- Learning-based methods to do the simulation
 - Packet wise
 - Area wise
- Data-driven method to improve the simulation

How to model water surface?

Build a height filed by integration

$$\eta(x, t) = \int_{-\infty}^{\infty} a(k) \cos(kx - \omega(k, h)t) dk$$

↑
Height
of point x
at time t ↑
Amplitude ↑
Cosine Wave ↑
Wave Number, $k = \frac{2\pi}{\lambda}$

Approximate it by discretizing around representative wavenumber

$$\eta(x, t) \approx \sum_{j=1}^N a_j \phi(x - c_g t) \cos(k_j(x - c_p t))$$

Water wave packet

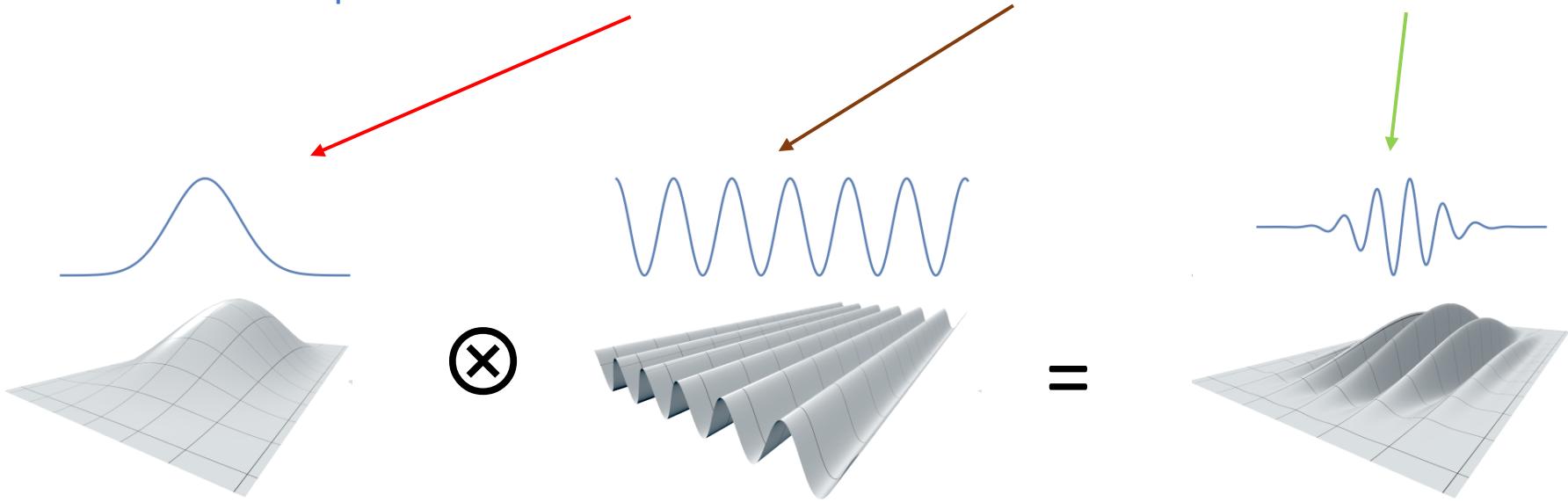
$$\eta(x, t) \approx \sum_{j=1}^N a_j \phi(x - c_g t) \cos(k_j(x - c_p t))$$

Amplitude

Kernel Function

Cosine Wave

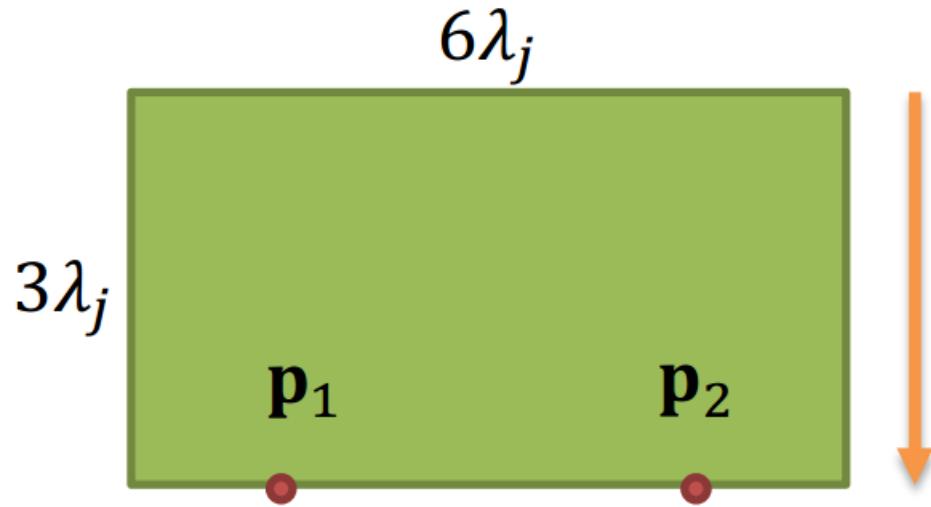
Representative k



Gaussian Kernel

Wave Packet

How to store a wave packet?



- 1. Tracing Two Vertices**
positions, velocity of $\mathbf{p}_1, \mathbf{p}_2$
- 2. Group Attributes**
group velocity, phase, amplitude...

Equations for simulating the packets

$$\mathbf{X}_{pg}(t + \Delta t) := \mathbf{X}_{pg}(t) + \Delta t(\mathbf{c}_p(k_j) - \mathbf{c}_g(k_j))$$

$$\mathbf{p}(t + \Delta t) := \mathbf{p}(t) + \Delta t \mathbf{c}_g(\mathbf{p}(t), k_j)$$

$$l(t + \Delta t) := l(t) + \Delta t(c_g(k_{\text{fast}}) - c_g(k_{\text{slow}}))$$

$$A_j(t_{n+1}) = ||\mathbf{p}_1(t_{n+1}) - \mathbf{p}_2(t_{n+1})||l(t_{n+1})$$

$$A_j(t_n) = ||\mathbf{p}_1(t_n) - \mathbf{p}_2(t_n)||l(t_n)$$

$$a_j(t_{n+1}) := a_j(t_n) \sqrt{\frac{(\rho g + \sigma k_n^2) A_j(t_n)}{(\rho g + \sigma k_{n+1}^2) A_j(t_{n+1})}}$$

...

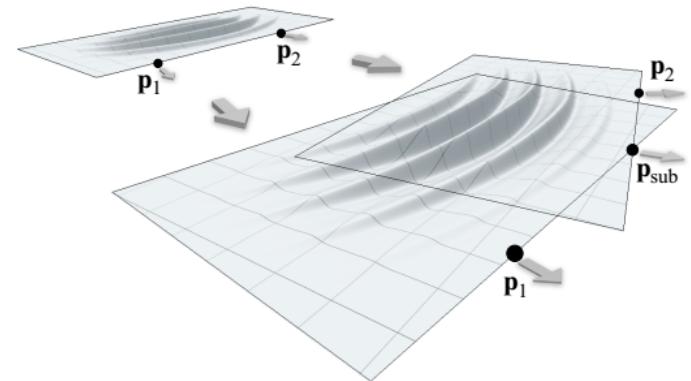
Should also consider
energy conservation,
reflection,
diffraction,
...

Learning comes to help

- **1. Generalizable**
 - one model can be applied to many kinds of fluid
- **2. End-to-end**
 - The pipeline can be simpler or faster
- **3. Data-driven**
 - can solve problems by improving training data

How to learn the simulation

- **Difficulty**
 - Overlap of packets
 - Variable number of packets



- **Two methods:**

1. **Packet wise**

Run inference for every packet

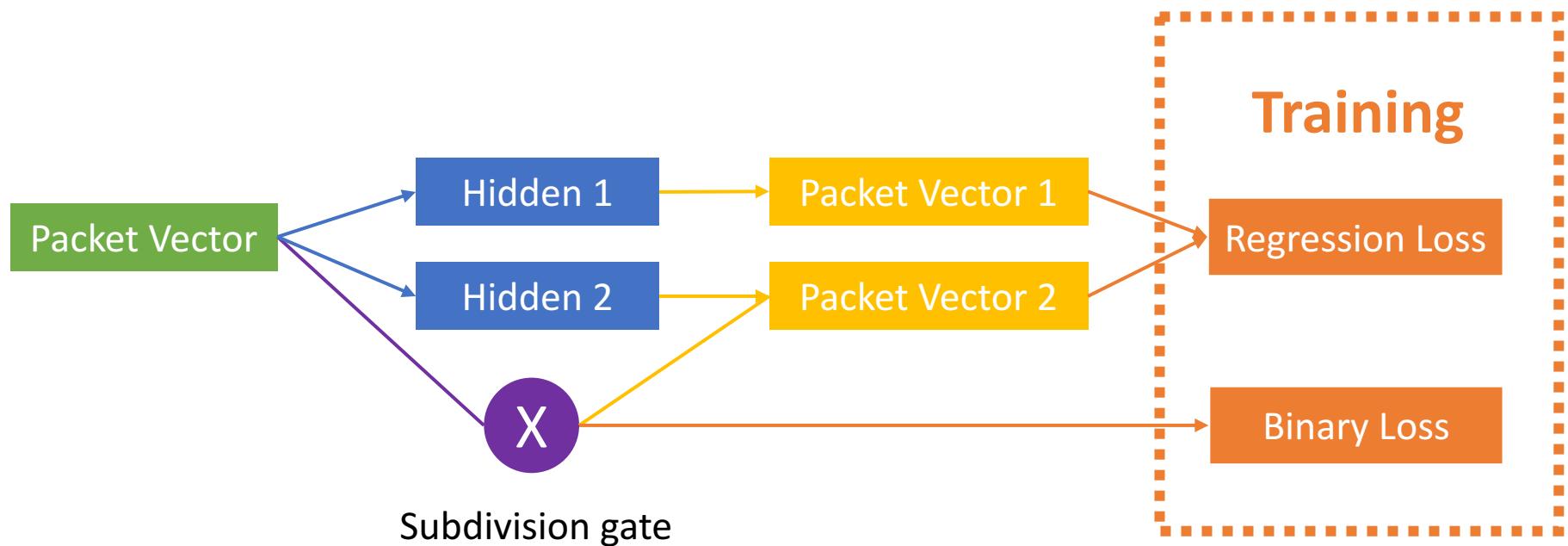
division of packet

2. **Area wise**

Run inference for every area patch

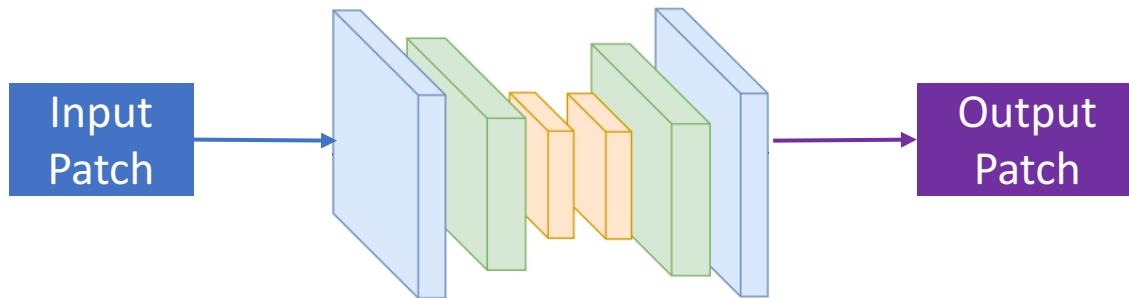
Packet wise

- Run inference for every packet

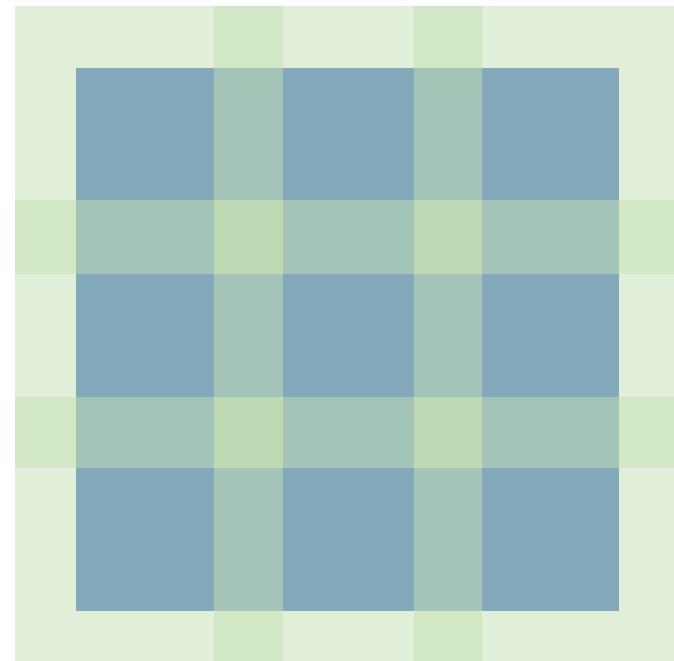


Area wise

- Run inference for every sub area



Use overlapping patches to improve the performance on boundary



Comparison

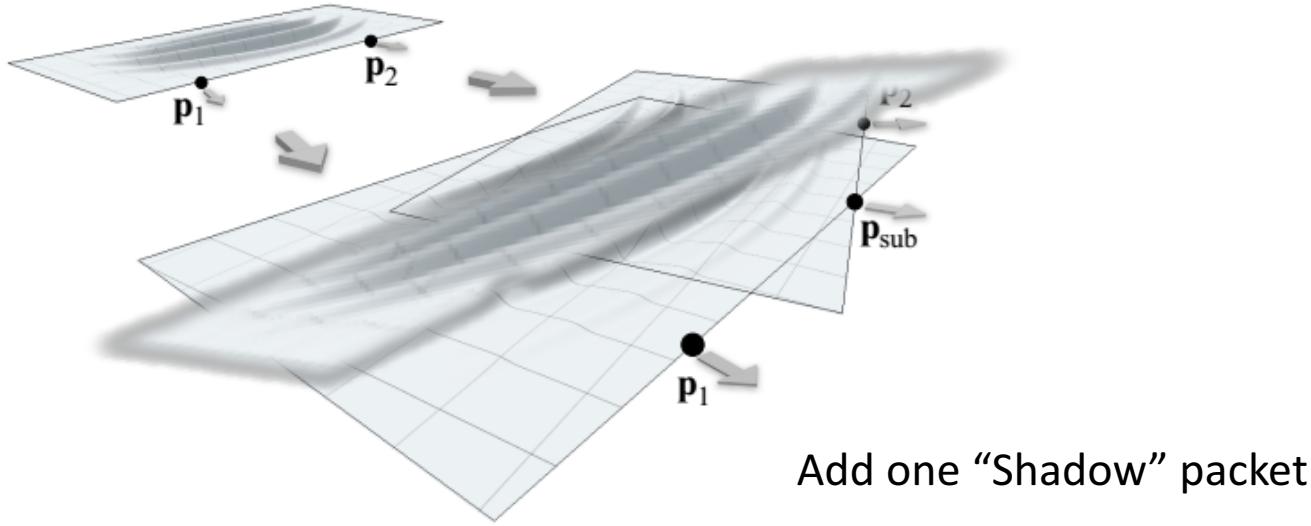
| | Packet wise | Area wise |
|---------------------------|---------------------------|------------------|
| Complexity | $O(\text{packet number})$ | $O(\text{area})$ |
| Handle overlap | ✓ | ✗ |
| Easy to handle reflection | ✗ | ✓ |
| More end-to-end | ✗ | ✓ |

Data-driven Improvement – Adding Shadow Packet

- Send a “Shadow” packet when a packet is dividing

Original: 1 packet -> **2** packets

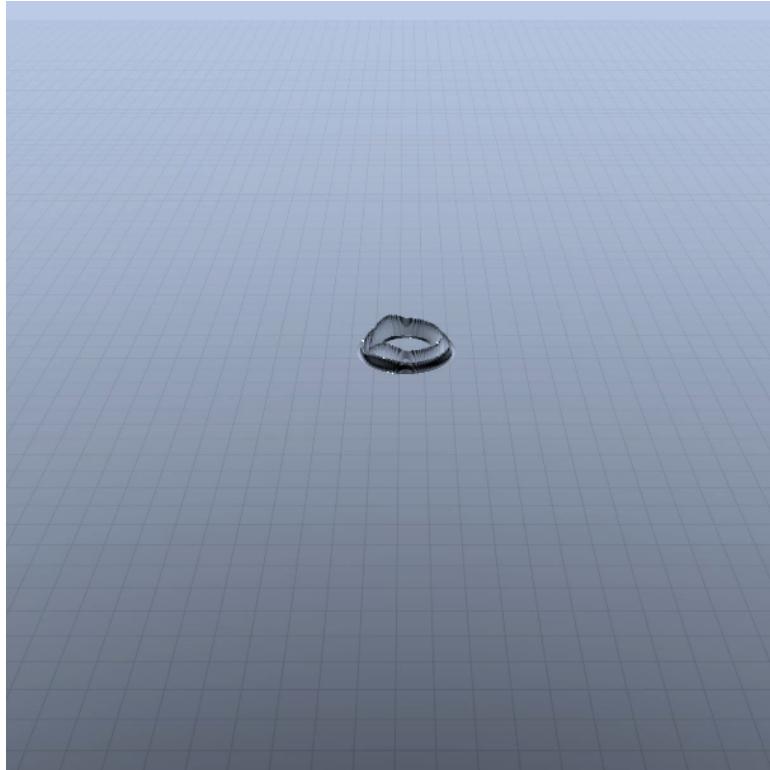
Now: 1 packet -> **3** packet



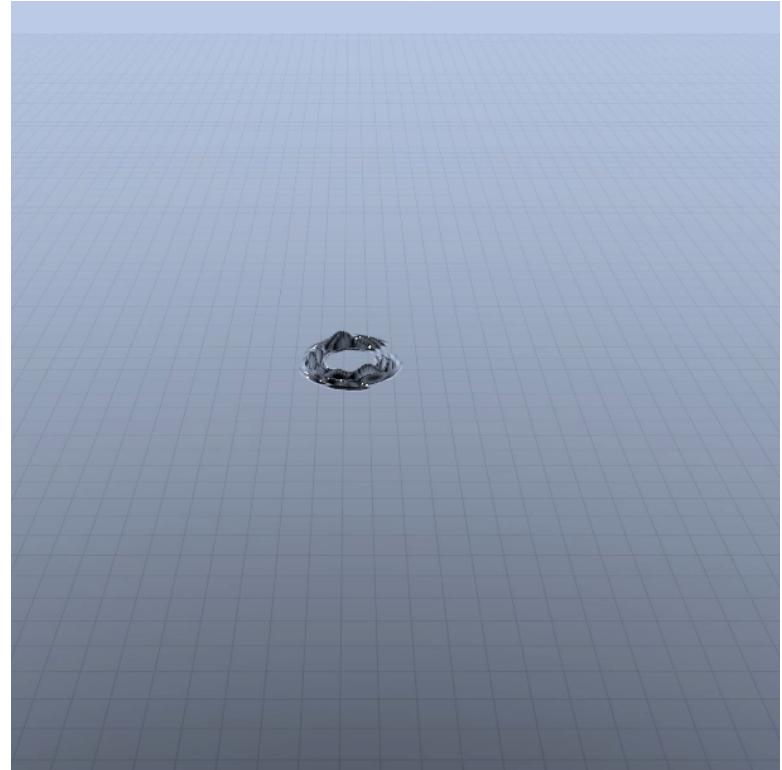
Experiment

- Learning-based method for wave propagation
- Data-driven improvement

Exp1: Learned Simulation

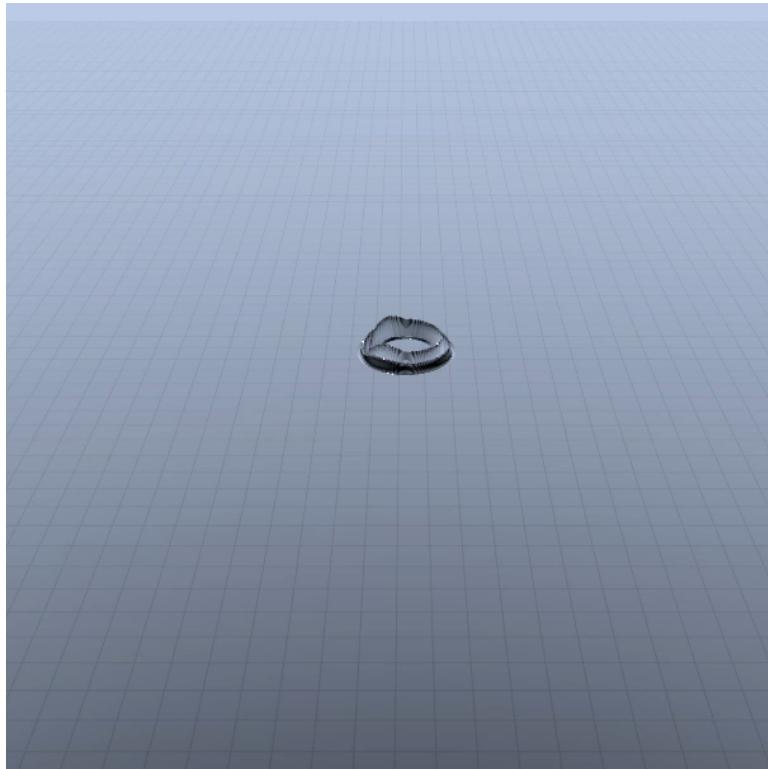


Simulated by Equations



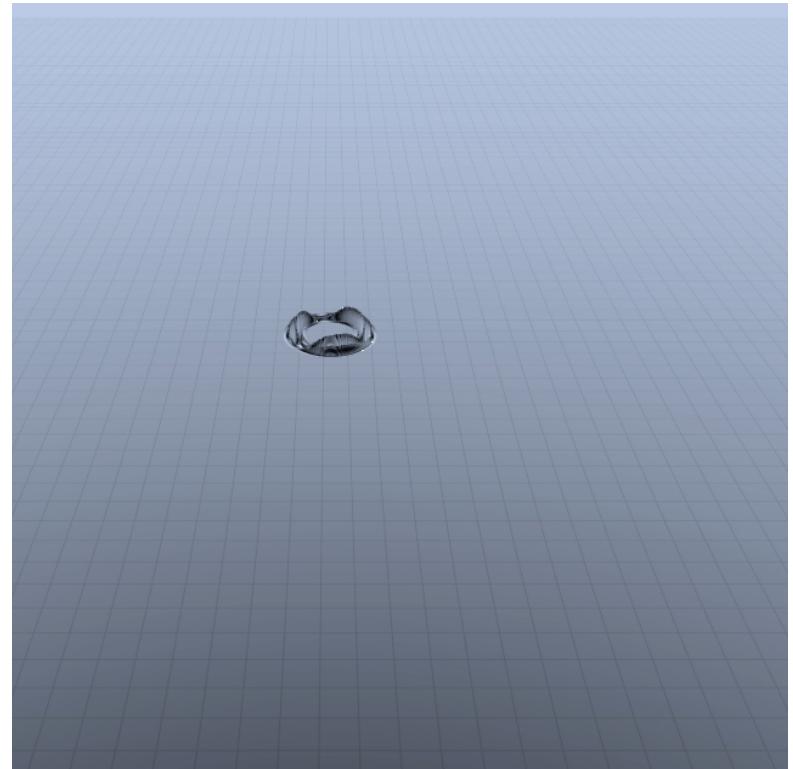
Simulated by Learning
(packet wise)

Exp2: Data Driven Improvement



No Shadow Packet

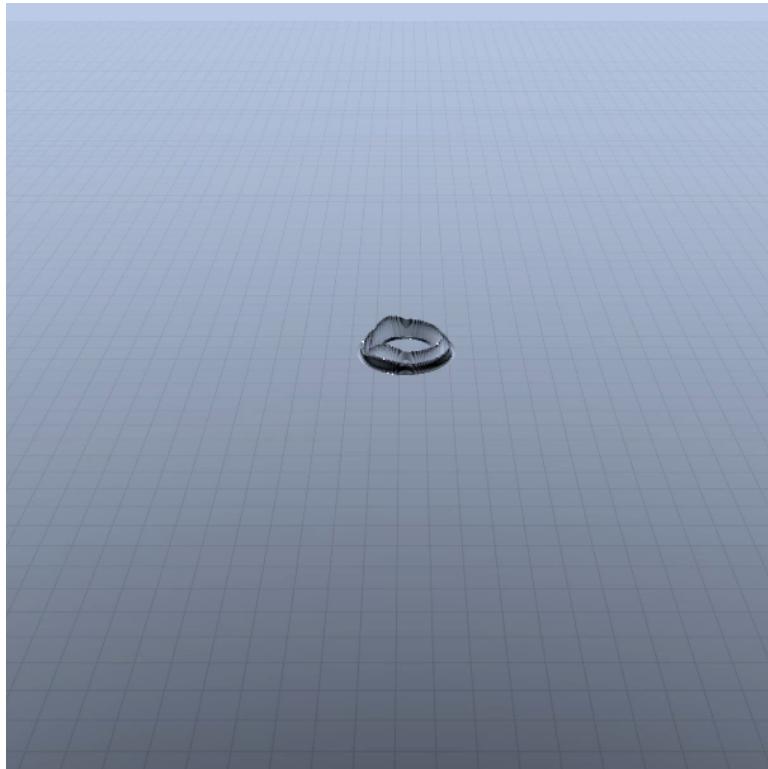
Unrealistic when doing division
You can observe some “dim frames”
when subdivision



Manually Add Shadow Packet

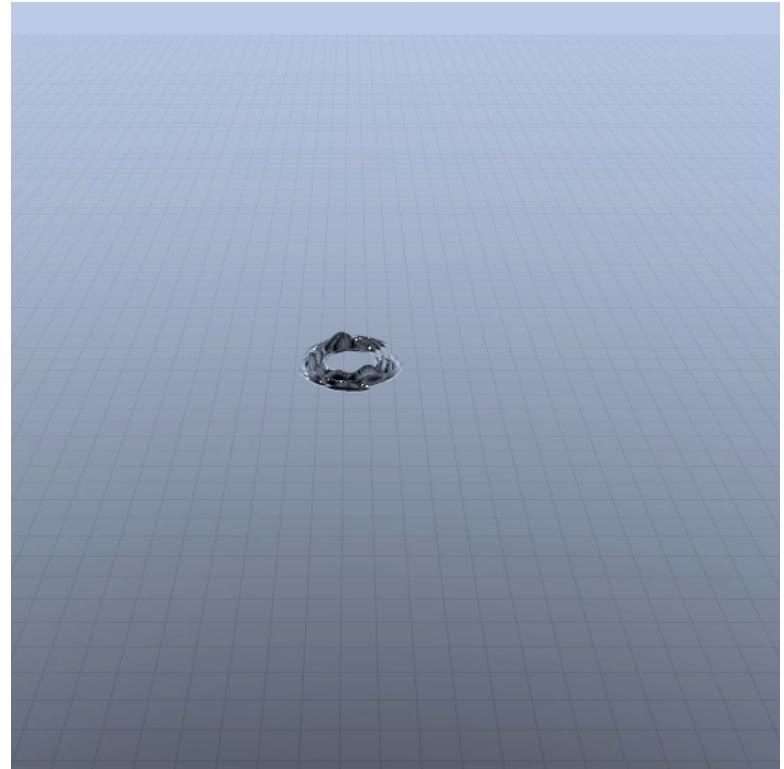
Smooth division

Exp2: Data Driven Improvement



No Shadow Packet

Unrealistic when doing division
You can observe some “dim frames”
when subdivision



Learned Shadow Packet

Smooth division

Future work

- (work in progress) Reflection in complicated environment
- More accurate architecture & loss function