



# PhysHand: A Hand Simulation Model with Physiological Geometry, Physical Deformation, and Accurate Contact Handling

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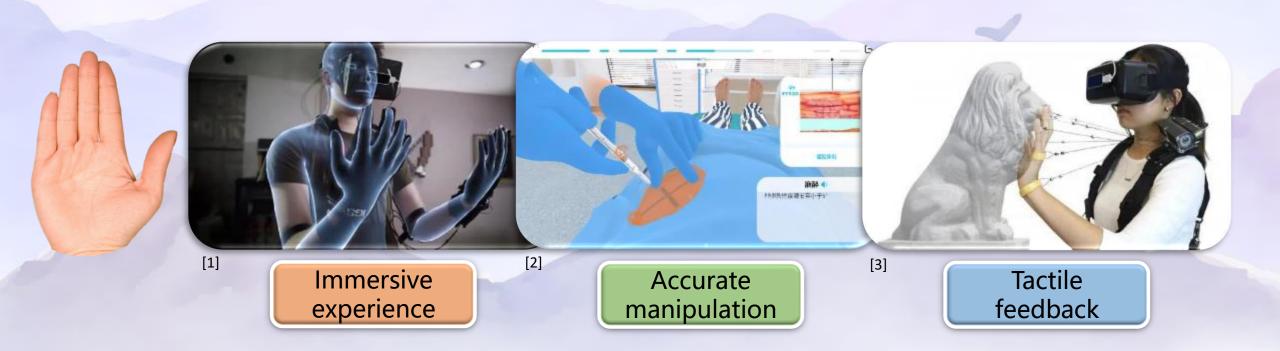
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In virtual Hand-Object Interaction (HOI), the authenticity of the hand's **deformation** is important:



<sup>[1]</sup> https://baijiahao.baidu.com/s?id=1763406672043028291&wfr=spider&for=pc

<sup>[2]</sup> https://xsj.699pic.com/sou/nagongjudeshou.html

<sup>[3]</sup> https://www.sohu.com/a/392493099 427506





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Challenges



Geometry

**Physics** 

**Contact** 

The human hand possesses a layered structure comprising bones, muscles, fat, and skin.





Challenges



**Geometry** 

**Physics** 

**Contact** 

The skeleton is rigid, but the skin and flesh exhibit distinct non-rigid behaviors.









**Geometry** 

**Physics** 

**Contact** 

Rich contacts always imply overlap in HOI, posing challenges for accurate and fast collision detection and correction.



## 2. Related works

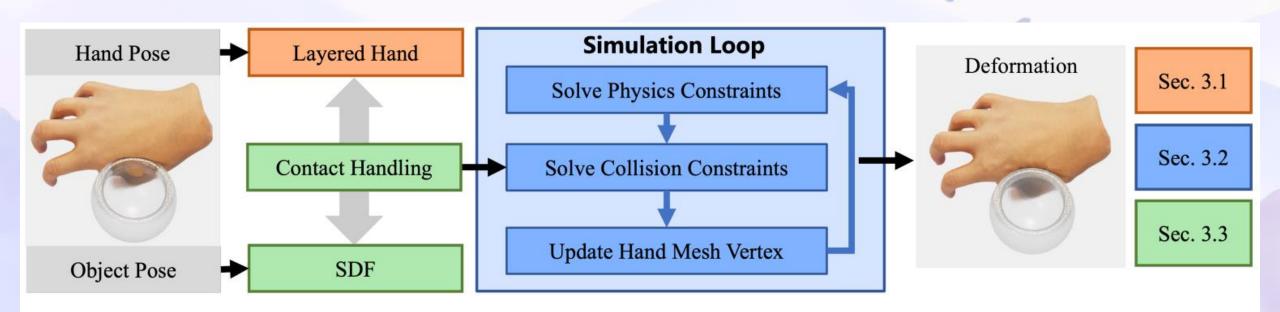


Research	Hand Layer			Deformation	Contact Handling	
	Skin	Flesh	Skeleton	Deformation	Contact Handing	
Ott et al. [OVT10]	✓	-	-	Skinning weights	Mesh	
Jacobs et al. [JSF12]	$\checkmark$	-	-	Skinning weights	Mesh	
Wang et al. [WML24]	$\checkmark$	-	-	Skinning weights	SDF	
Jacobs et al. [JF11]	$\checkmark$	$\checkmark$	-	Linear elasticity	Mesh	
Hirota et al. [HT16]	$\checkmark$	$\checkmark$	-	Linear elasticity	Mesh	
Garre et al. [GHGO11]	$\checkmark$	$\checkmark$	Box	Linear elasticity	Mesh	
Verschoor et al. [VLO18]	$\checkmark$	$\checkmark$	Capsule	Linear elasticity	SDF	
Sorli <i>et al</i> . [SCV*21]	$\checkmark$	$\checkmark$	Capsule	Linear elasticity	SDF	
Akihiko et al. [MYHYH17]	✓	✓	Mesh	Linear elasticity		
PhysHand (ours)	<b>√</b>	✓	Detailed mesh	Layer-corresponding constraints	SDF with multi-resolution querying	

Unrealistic deformation arises from simplified hand **geometry**, neglect of the different **physics** attributes of the hand, and penetration due to imprecise **contact** handling.







**Figure 2:** Schematic overview of the PhysHand. The pose of hand and object can be obtained by manual setup or from a generative model [LZY\*23]. The layered hand serves as a simulation entity, governed by the simulation loop. The simulation loop updates the vertices of the hand model by solving constraints, where the collision constraints are generated by the contact handling module when the hand comes into contact with the object.

To address these problems, we propose *PhysHand*, a novel hand simulation model, which enhances the realism of deformation in HOI.





# TENCENT

## **Geometry**

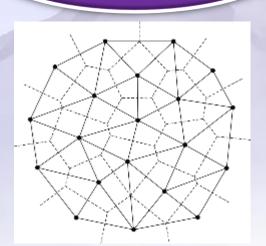
Generate hands with layered structure.

## NIMBLE[1]

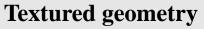
 $\mathcal{G}(\theta, \beta) = LBS(\mathcal{W}, J_p, \theta, \bar{T}).$ 



Delaunay triangulation









[1] Nimble: a non-rigid hand model with bones and muscles. ACM Transactions on Graphics (TOG), 41(4), 1-16.

Skin





## **Physics**

## Linear blend skinning for the skeleton

For the **skeleton** layer, we adopt skinning weights of the skeletal vertices derived from NIMBLE to calculate rigid body motions.

#### XPBD for the flesh and skin

For the **flesh** and the **skin**, we adopt XPBD[1] as our dynamics framework with carefully designed **layer-corresponding constraints** to maintain flesh attached and skin smooth.

[1] XPBD: position-based simulation of compliant constrained dynamics. In Proceedings of the 9th International Conference on Motion in Games (pp. 49-54).

#### Algorithm 1 XPBD simulation for a single time step.

1:  $h \leftarrow \Delta t / \text{numSubSteps}$   $\triangleright$  Subst

2: **for** i = 1, ..., numSubSteps do  $\triangleright$  Iteration

3: initialize solution  $\mathbf{x} \leftarrow \mathbf{x}^i + h\mathbf{v}^i + h^2\mathbf{M}^{-1}\mathbf{f}_{\text{ext}}(\mathbf{x}^i)$ 

4: initialize multipliers  $\lambda \leftarrow 0$ 

5:  $\tilde{\alpha} \leftarrow \frac{1}{h^2} \alpha$ 

6: collision detection and correction

7: **for all** constraints **do** 

8:  $\mathbf{A} \leftarrow \nabla C(\mathbf{x}) \mathbf{M}^{-1} \nabla C^T(\mathbf{x}) + \tilde{\alpha}$ 

9:  $\Delta \lambda \leftarrow -\mathbf{A}^{-1}(C(\mathbf{x}) + \tilde{\alpha}\lambda)$ 

10:  $\Delta \mathbf{x} \leftarrow \mathbf{M}^{-1} \nabla C^T(\mathbf{x}) \Delta \lambda$ 

11:  $\lambda \leftarrow \lambda + \Delta \lambda$ 

12:  $\mathbf{x} \leftarrow \mathbf{x} + \Delta \mathbf{x}$ 

13: end for

14: update positions  $\mathbf{x}^{i+1} \leftarrow \mathbf{x}$ 

15: update velocities  $\mathbf{v}^{i+1} \leftarrow \frac{\mathbf{x}^{i+1} - \mathbf{x}^i}{h}$ 

16:  $i \leftarrow i + 1$ 

17: **end for** 

$$▷ Substep size ▷ Iteration f_{ext}(\mathbf{x}^i)$$
 
$$▷ Sec. 3.3$$

▶ Projection



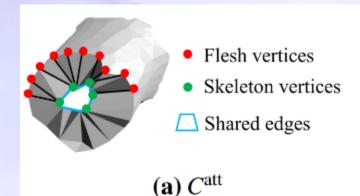


## **Physics**

## > Layer-corresponding constraints

$$C^{\text{edge}}(\mathbf{x}_1, \mathbf{x}_2) = \parallel \mathbf{x}_{1,2} \parallel -L,$$
 Edge length preserving

Basic Constraints 
$$C^{\text{edge}}(\mathbf{x}_1,\mathbf{x}_2) = \parallel \mathbf{x}_{1,2} \parallel -L, \qquad \qquad \text{Edge length preserving}$$
 
$$C^{\text{tet}}(\mathbf{x}_1,\mathbf{x}_2,\mathbf{x}_3,\mathbf{x}_4) = \mathbf{x}_{2,1} \times \mathbf{x}_{3,1} \cdot \mathbf{x}_{4,1} - 6V. \longrightarrow \text{Tetrahedron volume preserving}$$



$$C^{\text{att}}(\mathbf{x}_{\text{flesh}}) = C^{\text{edge}}(\mathbf{x}_{\text{flesh}}, \mathbf{x}_{\text{skeleton}}),$$
$$\mathbf{x}_{\text{skeleton}} = \arg \min_{\mathbf{x}_i \in S_{\text{skeleton}}} ||\mathbf{x}_{\text{flesh},i}||,$$

where  $S_{\text{skeleton}}$  is the set of vertices belong to the skeleton.

#### (a) Flesh Attachment Constraint

Each vertex of the flesh layer establishes an edge constraint with the nearest skeleton vertex to prevent excessive deformation.

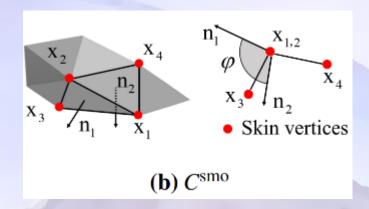




## **Physics**

## > Layer-corresponding constraints

$$C^{\text{edge}}(\mathbf{x}_1, \mathbf{x}_2) = \parallel \mathbf{x}_{1,2} \parallel -L,$$
 Edge length preserving



$$C^{\text{smo}}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4) = \arccos(\mathbf{n}_1 \cdot \mathbf{n}_2) - \mathbf{\varphi},$$

where  $\mathbf{x}_1$ ,  $\mathbf{x}_2$ ,  $\mathbf{x}_3$ , and  $\mathbf{x}_4$  are vertices that belong to two triangles with a shared edge,  $\mathbf{n}_1$  and  $\mathbf{n}_2$  are triangle normals and  $\boldsymbol{\varphi}$  is the initial dihedral angle. We apply  $C^{\text{smo}}$  to the skin.

#### (b) Skin Smoothness Constraint

For the skin layer, we keep the dihedral angle of adjacent triangles through normal vectors to provide smoothness.





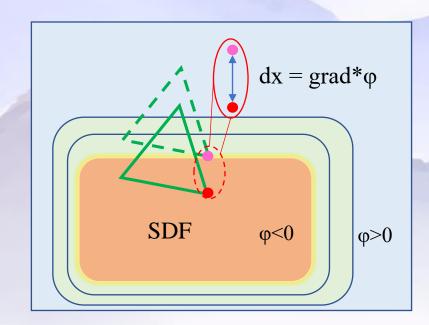
### **Contact**

## > Analysis for SDF-based collision detection

#### SDF Representation

 The zero-level-set of an SDF can implicitly describe the surface of an object:

$$\mathcal{S} = \{ \phi(\mathbf{x}) \} = 0,$$



#### SDF for collision detection

 The collision constraint can be defined directly by the SDF.

$$C^{\text{coll}}(\mathbf{x}) = \phi(\mathbf{x}).$$

#### **Collision constraint**

When  $C^{\text{coll}}(\mathbf{x}) \leq 0$ , penetration occurs, then  $\mathbf{x}$  needs to be projected to  $\mathbf{x}'$  that satisfies the collision constraint:

$$\mathbf{x}' = \mathbf{x} - \frac{\nabla \phi(\mathbf{x})}{||\nabla \phi(\mathbf{x})||} \phi(\mathbf{x}),$$

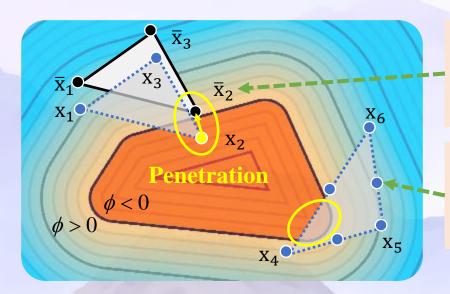
**Correction projection** 







### > Analysis for SDF-based collision detection



• Given a triangle, the TARGET is to find a point on the triangle to represent its minimum signed distance value, so that the penetrated triangle can be projected to the penetration-free position.

#### (a) Vanilla method: vertex-SDF [1]

For a triangle, the signed distances of its vertices and centers of edges fail to precisely depict its penetration state.

[1] Fuhrmann, A., Sobotka, G., & Groß, C. (2003, September). Distance fields for rapid collision detection in physically based modeling. In Proceedings of GraphiCon (Vol. 2003, pp. 58-65).

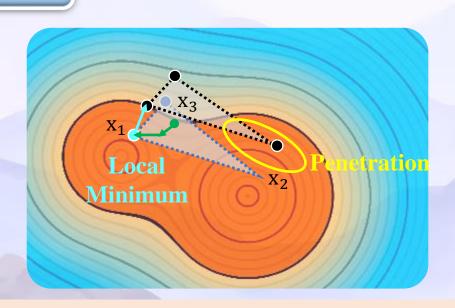


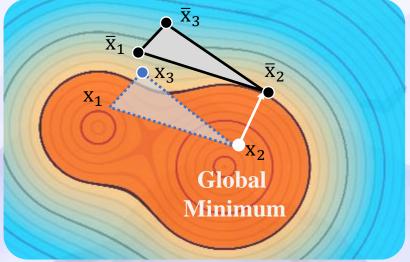


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#### **Contact**

> Analysis for SDF-based collision detection





#### (b) Optimization-based method: opt-SDF [1]

- It iteratively solves for an optimal point (green points and arrows) with the lowest signed distance.
- However, SDF is usually nonlinear, which makes opt-SDF sensitive to initial guesses and susceptible to converging towards local optima (bright blue point), leading to an insufficient correction.
- Instead, the global optimum (white point) can help eliminate penetration.

[1] Macklin, M., Erleben, K., Müller, M., Chentanez, N., Jeschke, S., & Corse, Z. (2020). Local optimization for robust signed distance field collision. Proceedings of the ACM on Computer Graphics and Interactive Techniques, 3(1), 1-17.







### > Analysis for SDF-based collision detection



Algorithm 2 Contact handling with multi-resolution querying.

```
1: for all triangles of the skin do
```

2: 
$$T = \{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$$

3: **for** i = 1, ..., sampling times **do** 

4: query the sampled points  $\mathcal{X}$  on triangle T

obtain the triangle  $T' = \{\mathbf{x}'_1, \mathbf{x}'_2, \mathbf{x}'_3\}$  with the lowest three signed distances.

6: 
$$T \leftarrow T'$$

7: end for

3: obtain the *decision point*  $\mathbf{x}^* \leftarrow \operatorname{argmin} \phi(\mathcal{X})$ 

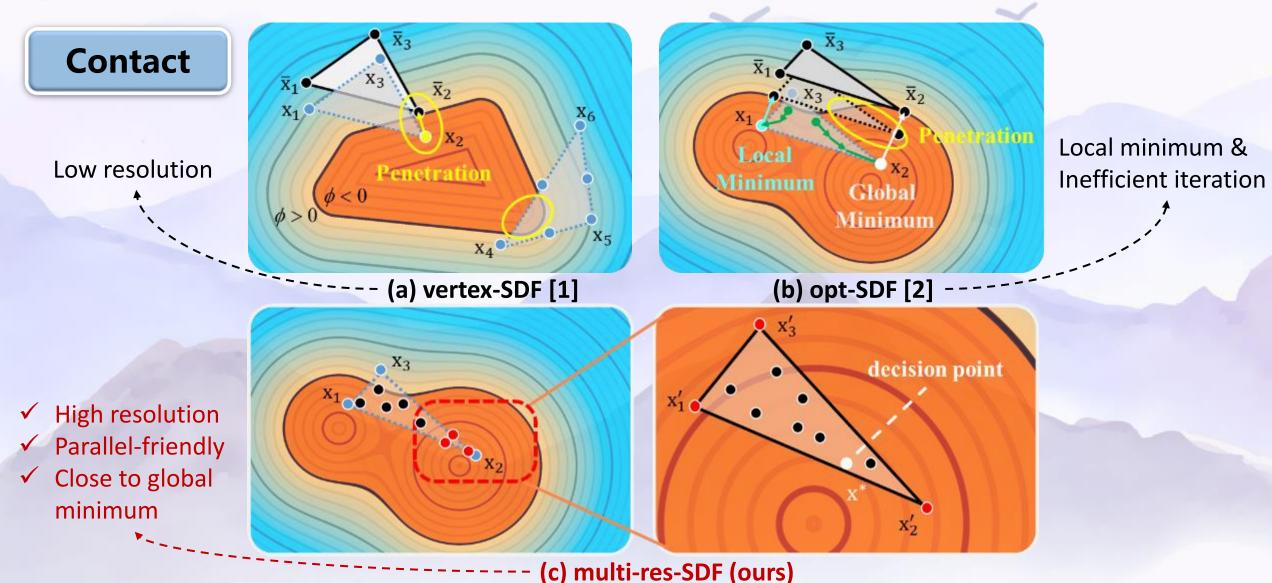
#### (c) Multi-resolution method (ours): multi-res-SDF

- We perform global sampling on the triangle (left), then select three points (red points) with the lowest signed distance to form a new triangle (right), iteratively increasing the resolution.
- After the final sampling iteration, we select the point with the smallest signed distance as the decision point x\*, which is more likely to approach the global optimum.
- Typically 2 sampling iteration is enough.





### TENCENT ROBOTICS X



- [1] Distance fields for rapid collision detection in physically based modeling. In Proceedings of GraphiCon (Vol. 2003, pp. 58-65).
- [2] Local optimization for robust signed distance field collision. Proceedings of the ACM on Computer Graphics and Interactive Techniques, 3(1), 1-17.





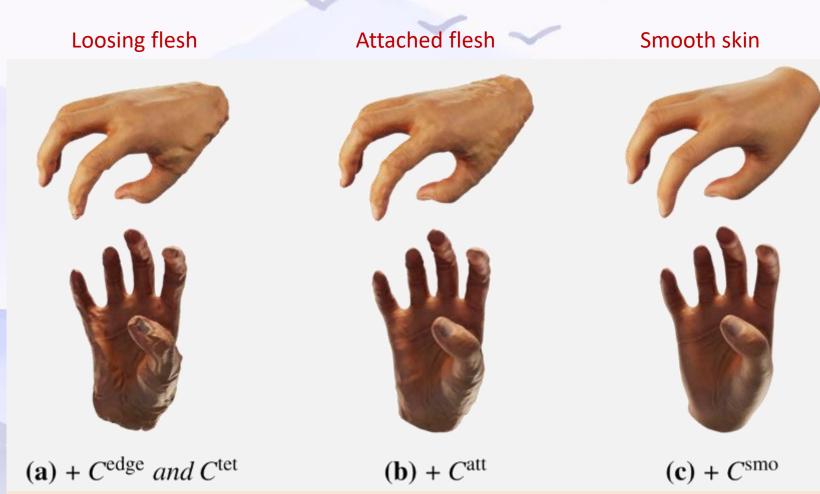
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# PhysHand Sec4.1 Constraints



## **Deformation**

Experiment on physical constraints



**Figure 6:** Improvements in deformation from layer-corresponding constraints. We gradually add constraints. (a) The hand with  $C^{\text{edge}}$  and  $C^{\text{tet}}$ . (b) The activation of  $C^{\text{att}}$ . (c) Enabling  $C^{\text{smo}}$ .

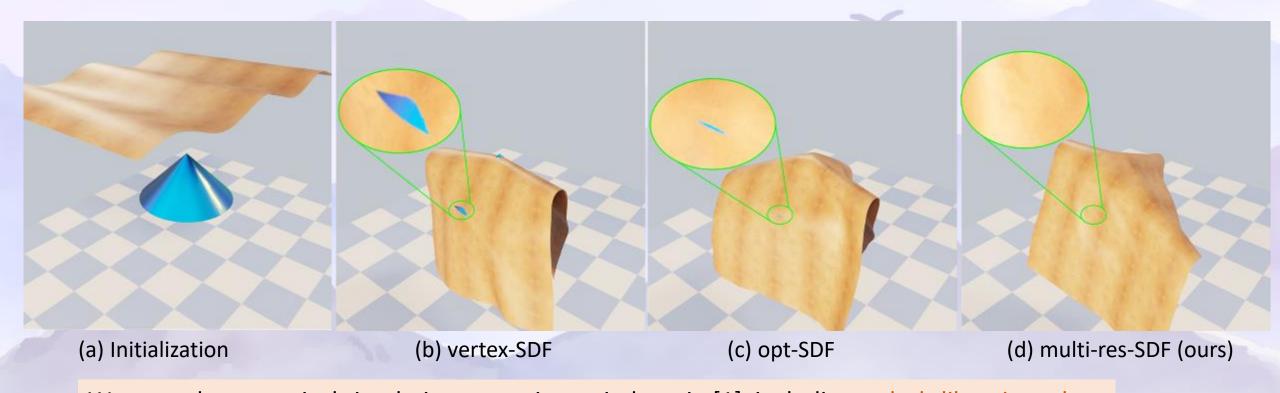




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**Contact** 

#### Qualitative comparison



We reproduce a typical simulation scenario carried out in [1], including a cloth-like triangular mesh and a cone represented as the SDF.

[1] Macklin, M., Erleben, K., Müller, M., Chentanez, N., Jeschke, S., & Corse, Z. (2020). Local optimization for robust signed distance field collision. Proceedings of the ACM on Computer Graphics and Interactive Techniques, 3(1), 1-17.

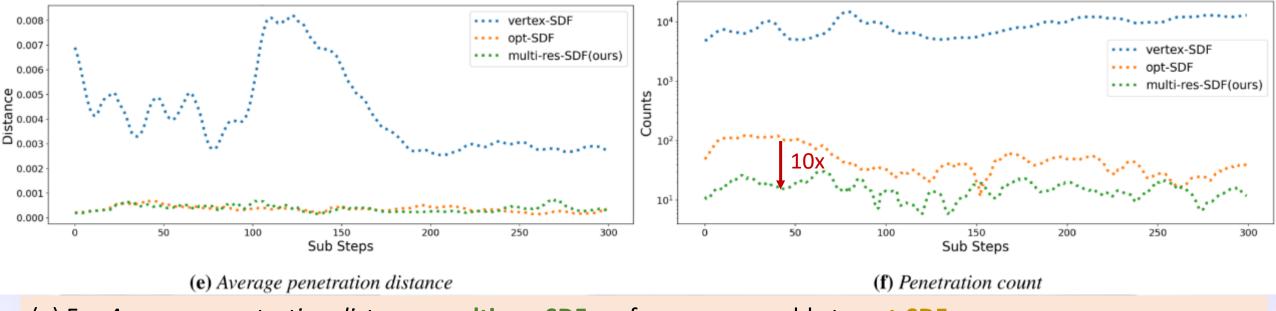




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#### **Contact**

#### Quantitative comparison



- (e) For Average penetration distance, multi-res-SDF performs comparably to opt-SDF.
- (f) For Penetration count, multi-res-SDF achieves nearly tenfold lower penetration count than opt-SDF.
- > It is exactly because opt-SDF has a higher number of penetrations that it reduces the average penetration distance.
- In our approach, since each iteration of querying is performed by global sampling on the corresponding triangle, our approach can approximate the global optima.





Perspective 3

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## HOI

**Experiments on HOI** data generation



ContactGen















- We replicate ContactGen [1], a SOTA model in generating hand-object grasps.
- Despite ContactGen creating excellent grasps, there are obvious penetrations.
- > PhysHand demonstrates its unique value by providing more authentic deformations and overcoming visible penetrations.

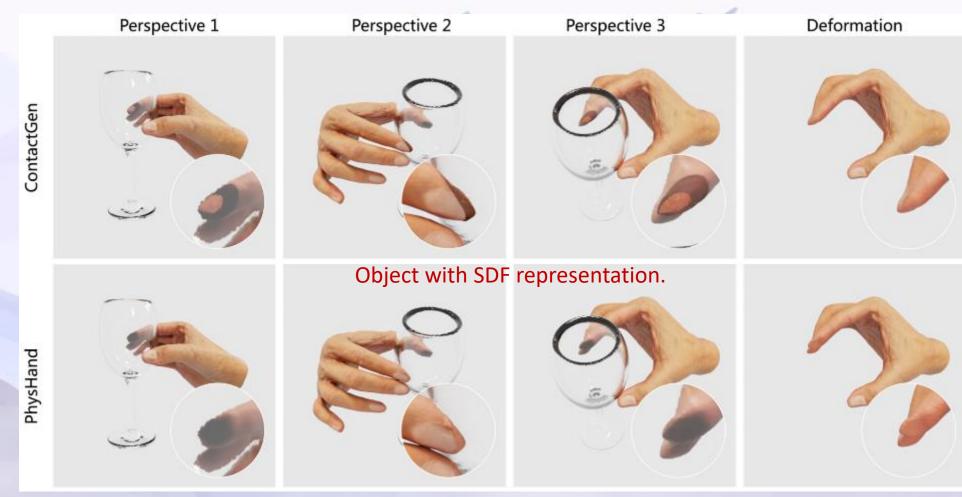




# TENCENT ROBOTICS X

### HOI

Experiments on HOI data generation



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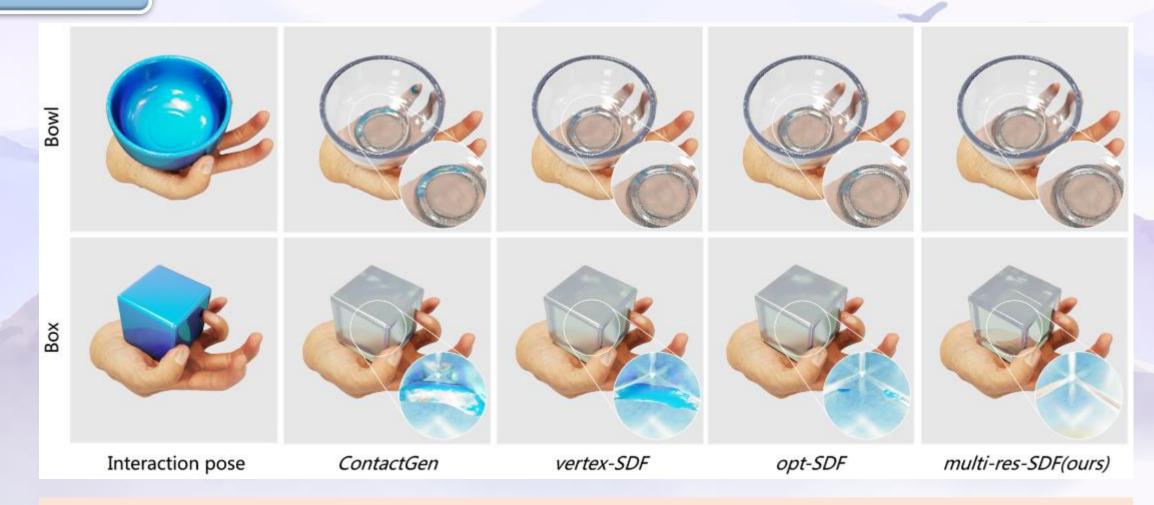




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HOI

Quantitative comparison of different SDF-based contact handling strategy



<sup>&</sup>gt; The dark blue regions indicate penetrations in all graphs except for the first column.





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### HOI

### Quantitative comparison of different SDF-based contact handling strategy



<sup>&</sup>gt; The dark blue regions indicate penetrations in all graphs except for the first column.





### TENCENT ROBOTICS X

#### HOI

Quantitative comparison of contact methods in HOI experiments.

- Across all metrics of contacts, our method significantly surpasses other approaches with values approaching zero.
- opt-SDF does not support parallelization due to the sequential querying.
- Querying in multi-res-SDF is independent of each point, allowing for parallel implementation.

	Object	ContactCon [L7V*22]	PhysHand			
Object		ContactGen [LZY*23]	vertex-SDF [FSG03] opt-SDF [MEN		A CONTRACTOR OF THE PROPERTY O	
	count.	143105	46285	156	3	
glass-1	max (μm)↓	2503.9162	459.1479	6.2744	0.2956	
	avg. (μm)↓	1305.9646	114.9909	1.9033	0.2289	
	CPU/GPU time (ms)↓		22/-	246/-	2,517/ <b>27</b>	
	count.	119067	63013	158	1	
glass-2	max (μm)↓	2503.8602	442.6264	3.5655	0.0500	
	avg. (μm)↓	1150.6618	206.9566	1.9978	0.0500	
	CPU/GPU time (ms)↓	-	21/-	243/-	2,502/ <b>27</b>	
toothpaste	count.	71033	27435	30	1	
	max (μm)↓	3106.4365	2313.6602	7.4098	0.0139	
	avg. (μm)↓	1068.7383	270.8561	1.7885	0.0139	
	CPU/GPU time (ms) ↓	5	21/-	169/-	2,462/26	
	count.	287730	164933	2	1	
phone	max (μm)↓	3323.5322	1916.7412	1.3571	0.0103	
	avg. (μm)↓	1411.4840	70.8230	0.7542	0.0103	
	CPU/GPU time (ms) ↓		15/-	227/-	1,095/12	
	count↓	79827	37927	268	0	
apple	max (μm)↓	3294.8354	189.9071	3.7722	0	
	avg. (μm)↓	1219.7463	78.8565	1.0504	0	
	CPU/GPU time (ms) ↓		21/-	220/-	1,995/24	
	count\_	86841	32120	190	11	
bowl	max (μm)↓	5718.0323	3129.6544	260.9675	0.0427	
	avg. (μm)↓	1788.3155	845.0303	35.4007	0.0093	
	CPU/GPU time (ms) ↓	2)	18/-	145/-	1,480/16	
	count↓	163436	58827	820	7	
box	max (μm)↓	5217.0626	989.9247	0.4664	0.0050	
	avg. (μm)↓	2331.5708	133.4961	0.0110	0.0016	
	CPU/GPU time (ms) ↓	-	17/-	201/-	1,073/12	

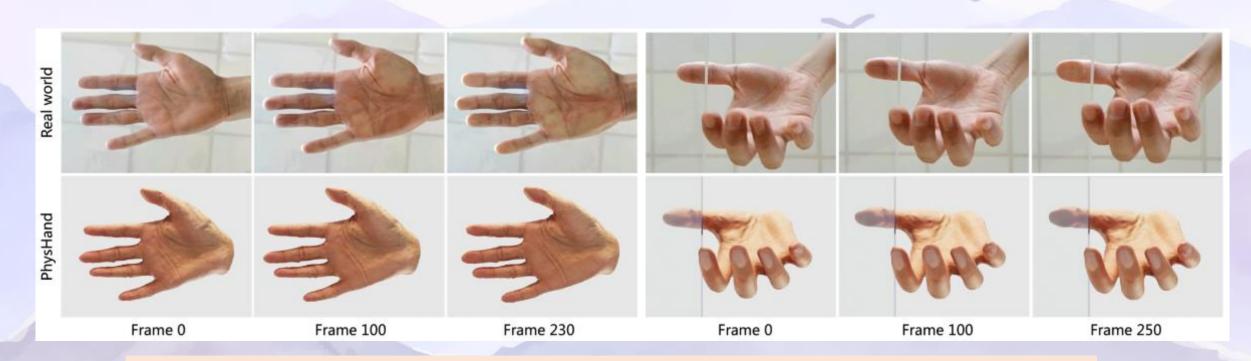




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HOI

Real world comparison



PhysHand is able to achieve deformations remarkably close to those observed in the real world.



## 5. Conclusion

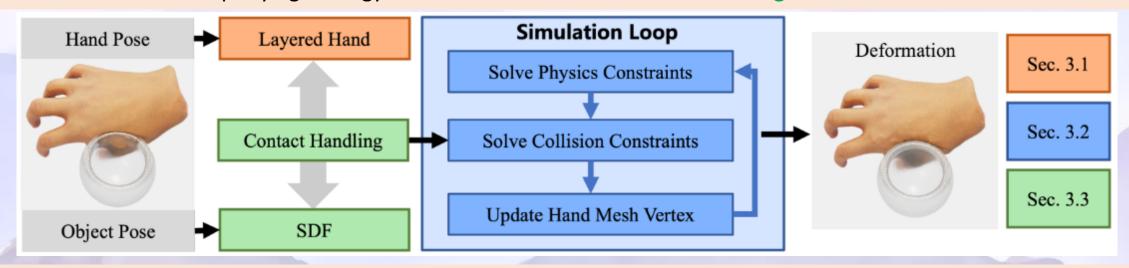


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We present a novel hand simulation model, *PhysHand*, capable of faithfully reproducing the deformations in HOI.

#### **Contributions:**

- A layered hand geometry containing the skin, flesh, and skeleton.
- Layer-corresponding physics formulas within a constraint-based dynamics framework.
- A multi-resolution querying strategy for accurate SDF-based contact handling.



#### **Limitations:**

- 1. Since self-collision is not considered in this work, self-penetration between triangles might occur after solving the collision constraint.
- 2. PhysHand primarily focuses on deformation rather than actuation, thus we treat muscles and fat as the same soft tissue.
- 3. The absence of friction prevents PhysHand from actively grasping objects.





# Thanks for attention!

PhysHand: A Hand Simulation Model with Physiological Geometry, Physical Deformation, and Accurate Contact Handling

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