

# Dynamic 3D Gaussian Splatting

Part of a Tutorial on 3D Gaussian Splatting  
at 3DV 2024

Jonathon Luiten

# Dynamic 3D Gaussian Splatting

Dynamic Gaussian Splatting has Exploded!

First paper on ArXiv 18<sup>th</sup> Aug 2023.

7 Months later ~50 papers.

## Dynamic 3D Gaussians:

### Tracking by Persistent Dynamic View Synthesis

Jonathon Luiten<sup>1,2</sup> Georgios Kopanas<sup>3</sup> Bastian Leibe<sup>2</sup> Deva Ramanan<sup>1</sup>

<sup>1</sup>Carnegie Mellon University, USA <sup>2</sup>RWTH Aachen University, Germany <sup>3</sup>Inria & Université Côte d'Azur, France  
luiten@vision.rwth-aachen.de

## 4D Gaussian Splatting:

### Towards Efficient Novel View Synthesis for Dynamic Scenes

Yuanxing Duan<sup>1\*</sup> Fangyin Wei<sup>2\*</sup> Qiyu Dai<sup>1,4</sup> Yuhang He<sup>1</sup> Wenzheng Chen<sup>3†</sup> Baoquan Chen<sup>1,4†</sup>

<sup>1</sup>Peking University <sup>2</sup>Princeton University <sup>3</sup>NVIDIA <sup>4</sup>National Key Lab of General AI, China

## REAL-TIME PHOTOREALISTIC DYNAMIC SCENE REPRESENTATION AND RENDERING WITH 4D GAUSSIAN SPLATTING

Zeyu Yang, Hongye Yang, Zijie Pan, Li Zhang\*  
Fudan University

<https://fudan-zvg.github.io/4d-gaussian-splatting>

## Deformable 3D Gaussians for High-Fidelity Monocular Dynamic Scene Reconstruction

Ziyi Yang<sup>1,2</sup> Xinyu Gao<sup>1</sup> Wen Zhou<sup>2</sup> Shaohui Jiao<sup>2</sup> Yuqing Zhang<sup>1</sup> Xiaogang Jin<sup>1†</sup>

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## 4D Gaussian Splatting for Real-Time Dynamic Scene Rendering

Guanjun Wu<sup>1\*</sup>, Taoran Yi<sup>2\*</sup>, Jiemin Fang<sup>3†</sup>, Lingxi Xie<sup>3</sup>, Xiaopeng Zhang<sup>3</sup>,  
Wei Wei<sup>1</sup>, Wenyu Liu<sup>2</sup>, Qi Tian<sup>3</sup>, Xinggang Wang<sup>2†‡</sup>

<sup>1</sup>School of CS, Huazhong University of Science and Technology

<sup>2</sup>School of EIC, Huazhong University of Science and Technology <sup>3</sup>Huawei Inc.

{guajuwu, taoranyi, weiw, liuw, xgwang}@hust.edu.cn

{jaminfong, 198808xc, zxphistory}@gmail.com tian.qil@huawei.com

## GauFR<sup>e</sup> 🍷: Gaussian Deformation Fields for Real-time Dynamic Novel View Synthesis

Yiqing Liang<sup>‡</sup>, Numair Khan, Zhengqin Li, Thu Nguyen-Phuoc,  
Douglas Lanman, James Tompkin<sup>‡</sup>, Lei Xiao  
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## MD-Splatting: Learning Metric Deformation from 4D Gaussians in Highly Deformable Scenes

Bardienus P. Duisterhof  
Carnegie Mellon University  
bduister@cmu.edu

Zhao Mandi  
Stanford University  
mandi@stanford.edu

Yunchao Yao  
Carnegie Mellon University  
yunchao@andrew.cmu.edu

Jia-Wei Liu  
National University of Singapore  
jiawei.liu@u.nus.edu

Mike Zheng Shou  
National University of Singapore  
mike.zheng.shou@gmail.com

Shuran Song  
Stanford University  
shuran@stanford.edu

Jeffrey Ichnowski  
Carnegie Mellon University  
jeffi@cmu.edu

## An Efficient 3D Gaussian Representation for Monocular/Multi-view Dynamic Scenes

Kai Katsumata Duc Minh Vo Hideki Nakayama  
The University of Tokyo, Japan

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## DynMF: Neural Motion Factorization for Real-time Dynamic View Synthesis with 3D Gaussian Splatting

Agelos Kratimenos

Jiahui Lei  
University of Pennsylvania

Kostas Daniilidis

Project Page: <https://agelosk.github.io/dynmf/>

## Gaussian-Flow: 4D Reconstruction with Dynamic 3D Gaussian Particle

Youtian Lin<sup>1</sup> Zuozhuo Dai<sup>2</sup> Siyu Zhu<sup>3</sup> Yao Yao<sup>1✉</sup>  
<sup>1</sup>Nanjing University <sup>2</sup>Alibaba Group <sup>3</sup>Fudan University

## Spacetime Gaussian Feature Splatting for Real-Time Dynamic View Synthesis

Zhan Li<sup>1,2\*</sup> Zhang Chen<sup>1†</sup> Zhong Li<sup>1†</sup> Yi Xu<sup>1</sup>  
<sup>1</sup>OPPO US Research Center <sup>2</sup>Portland State University  
lizhan@pdx.edu zhang.chen@oppo.com zhong.li@oppo.com yi.xu@oppo.com  
<https://oppo-us-research.github.io/SpacetimeGaussians-website/>

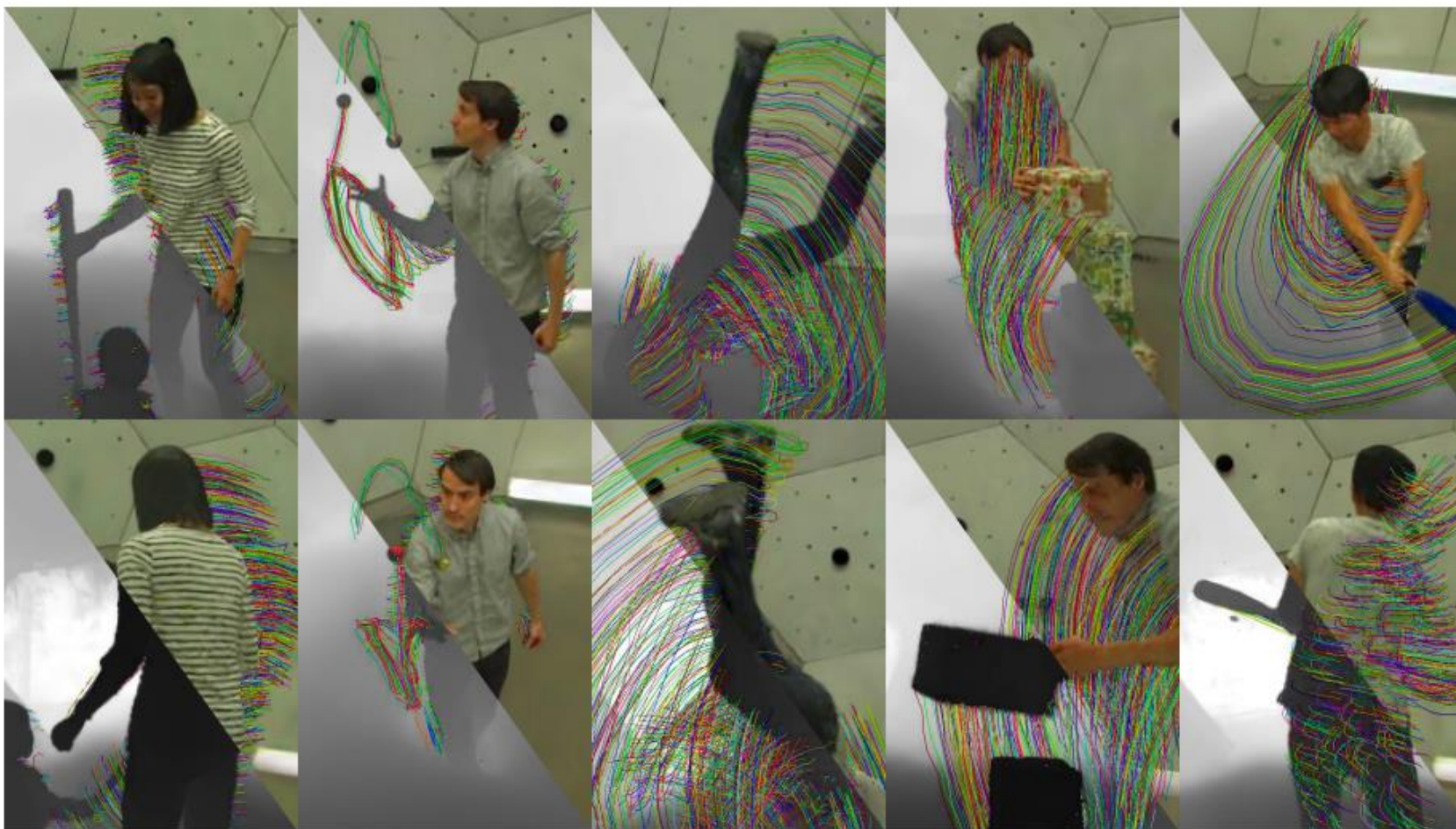
# Dynamic 3D Gaussians

## Dynamic 3D Gaussians: Tracking by Persistent Dynamic View Synthesis

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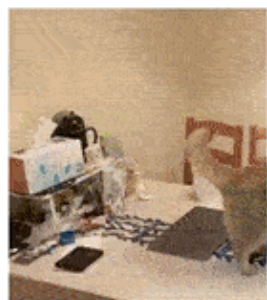
# Analysis-by-Synthesis for Tracking and Dynamic 3D?



# A Good Representation

Gaussians?

BANMo



Iteration: 0  
Epoch: 0.0



Target Mask



Loss

Estimated  
Color

Estimated  
Depth

Fuzzy Metaballs

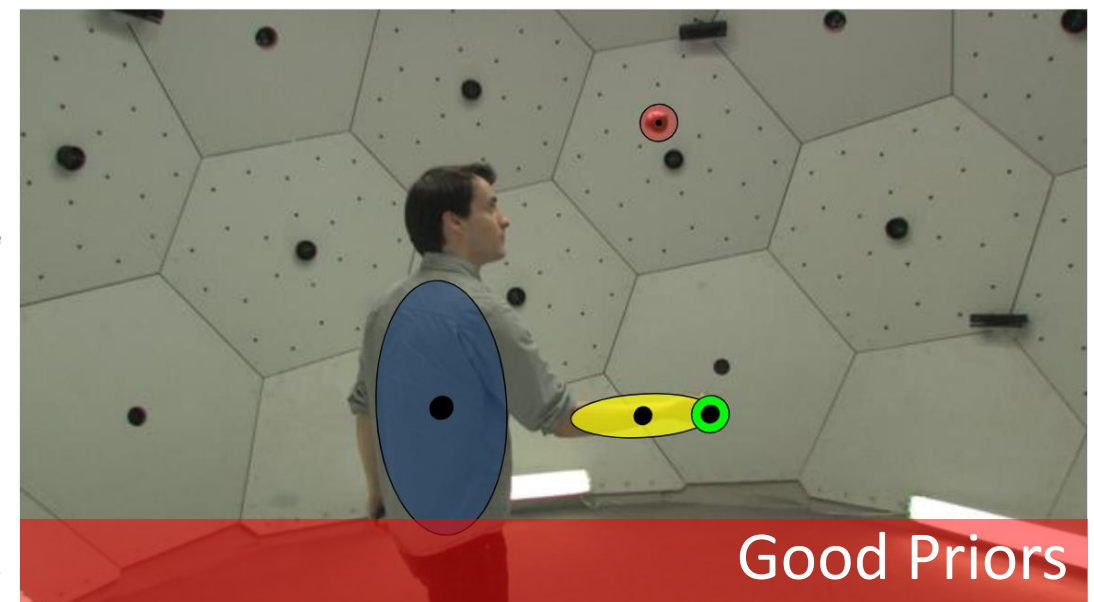
# 3D Gaussian Splatting



# Dynamic 3D Gaussian Splatting?



# What is needed?

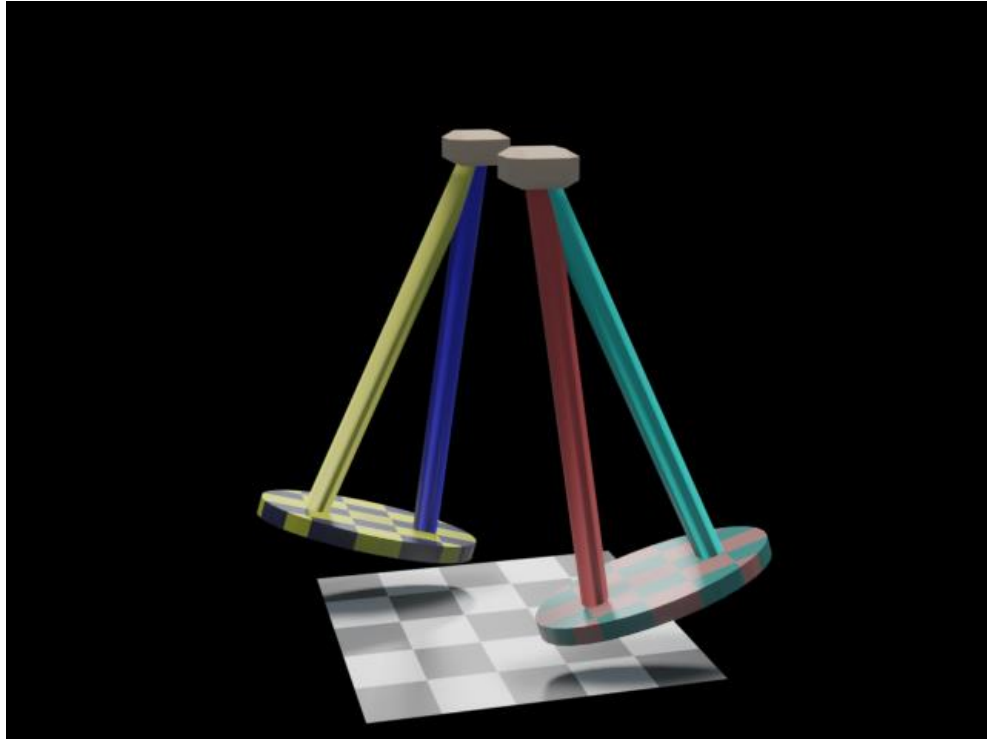




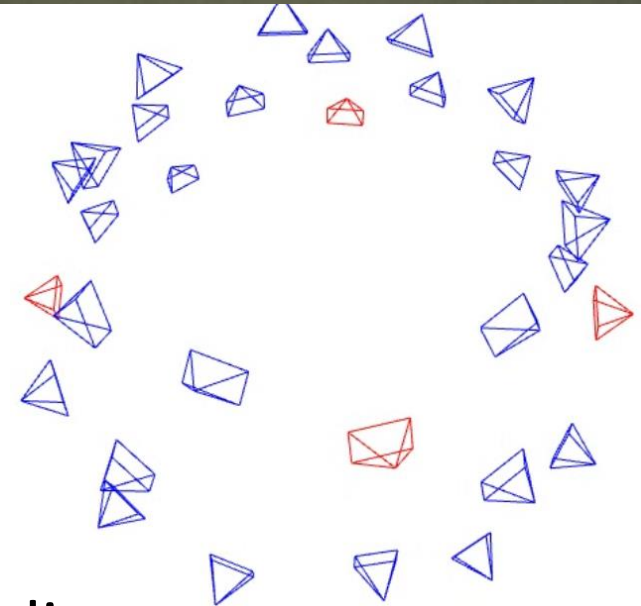
# What is needed?



# Good Data



Pixel NeRF



# Good Data



D-Nerf



Neural 3D Video

# What is needed?



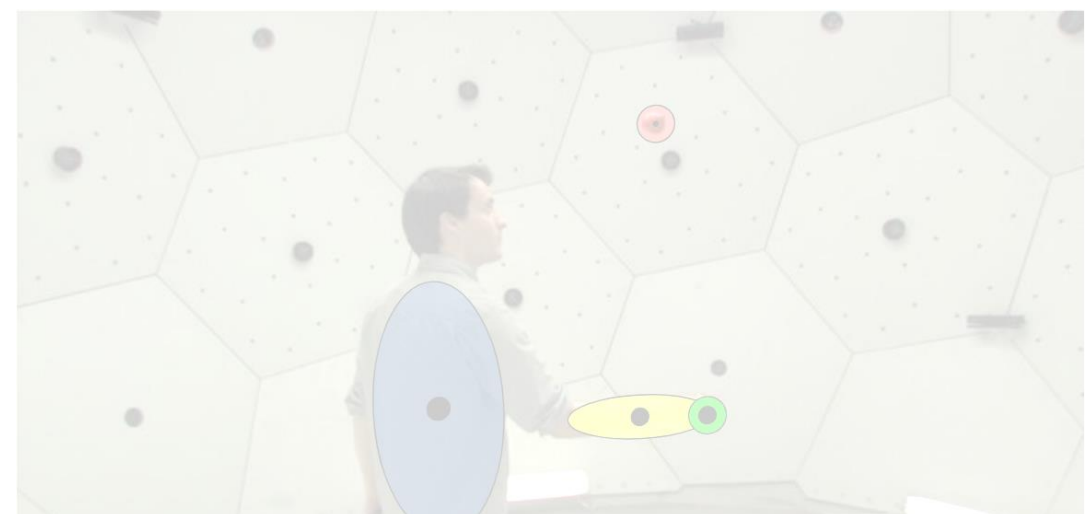
Good Data



A Good Representation



A Good Optimization Procedure



Good Priors

# 3D Gaussians

$$f_{i,t}(p) = \text{sigm}(o_i) \exp \left( -\frac{1}{2} (p - \mu_{i,t})^T \Sigma_{i,t}^{-1} (p - \mu_{i,t}) \right)$$

$$\Sigma_{i,t} = R_{i,t} S_i S_i^T R_{i,t}^T$$

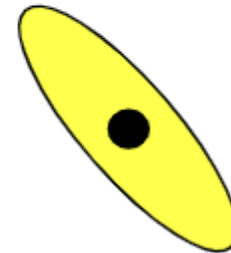
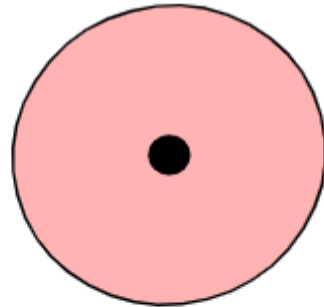
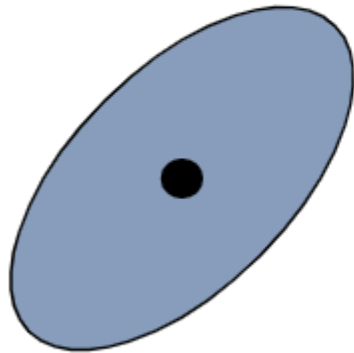
3D Center

3D Size

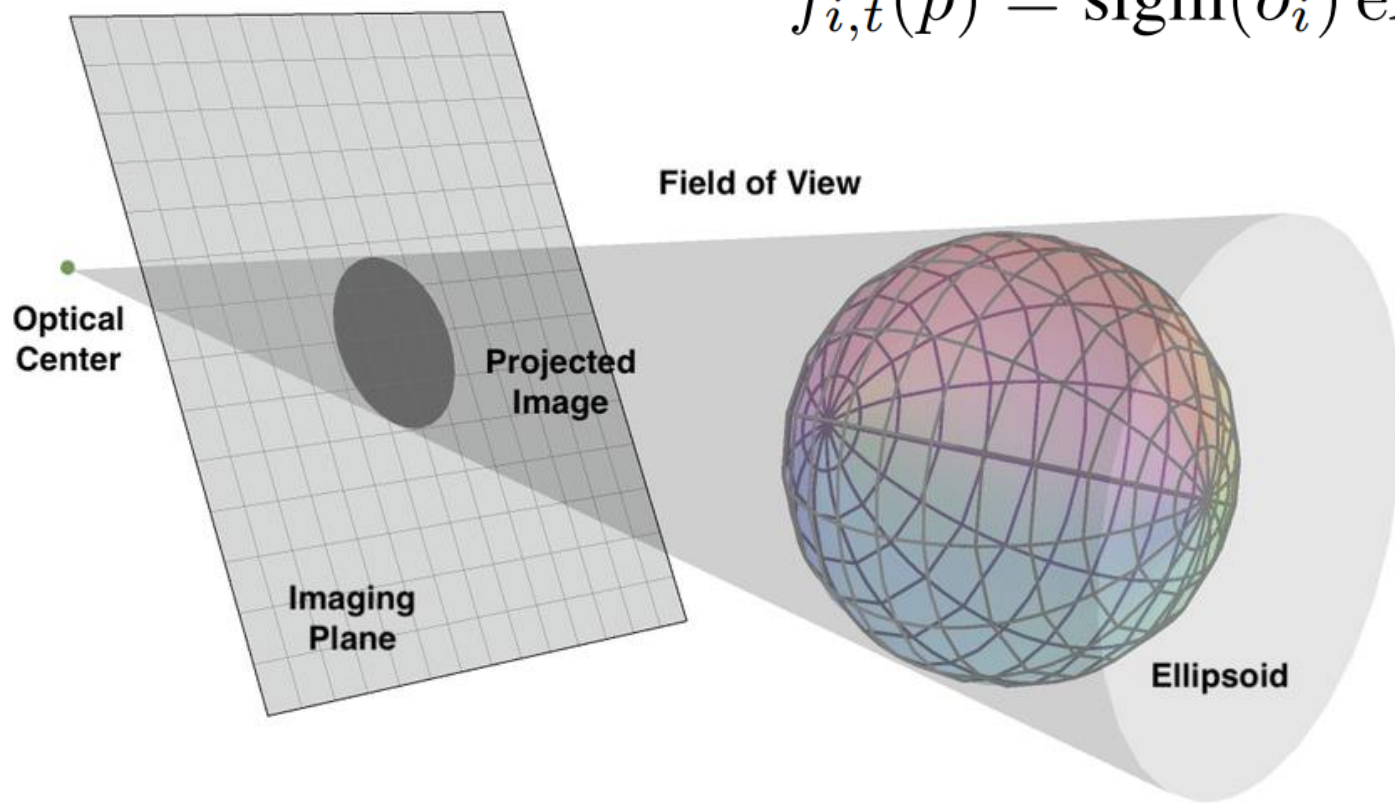
3D Rotation

Color

Opacity



# Rendering 3D Gaussians



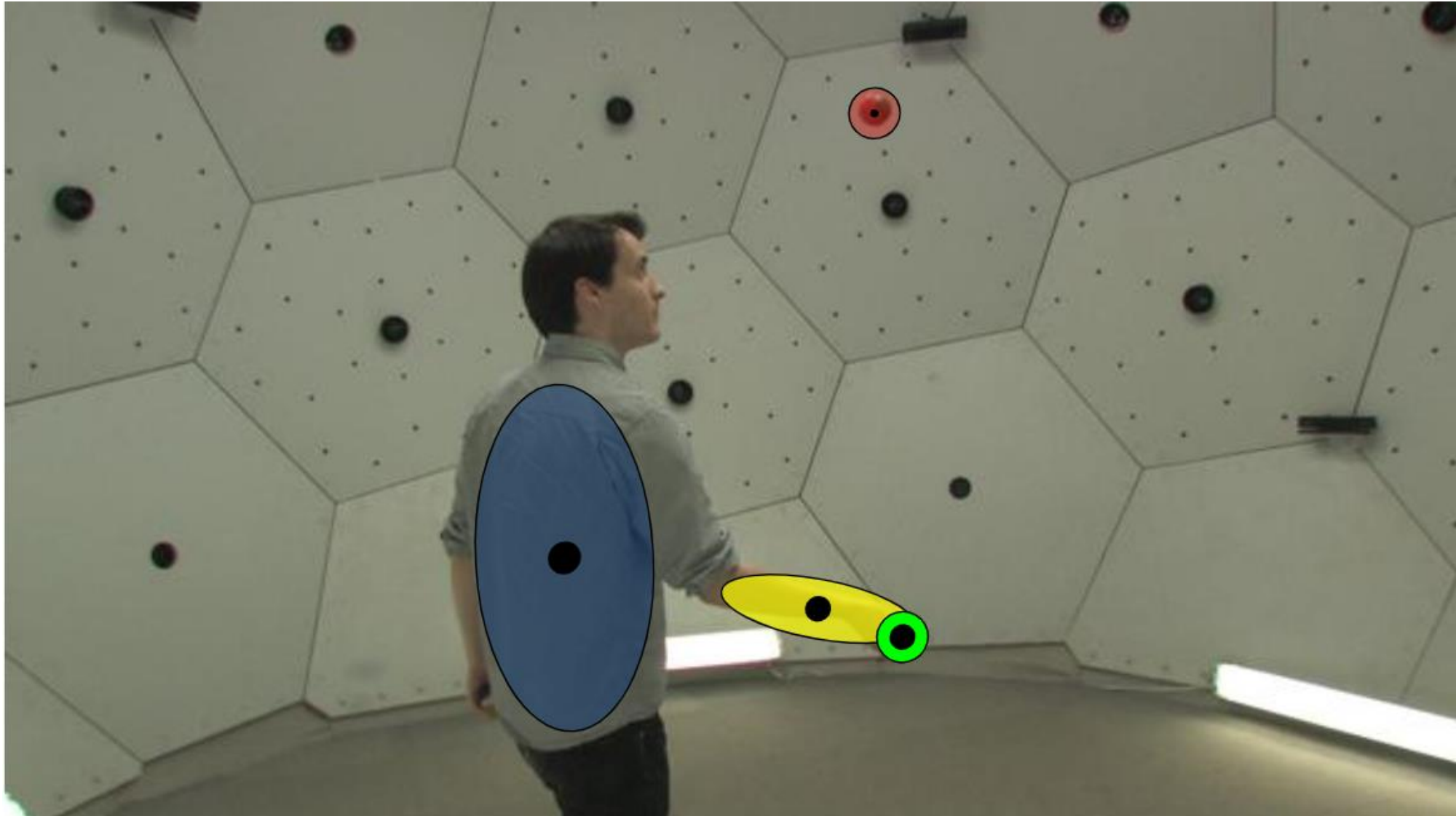
$$f_{i,t}(p) = \text{sigm}(o_i) \exp \left( -\frac{1}{2} (p - \mu_{i,t})^T \Sigma_{i,t}^{-1} (p - \mu_{i,t}) \right)$$

$$\mu^{2D} = K ((E\mu) / (E\mu)_z)$$

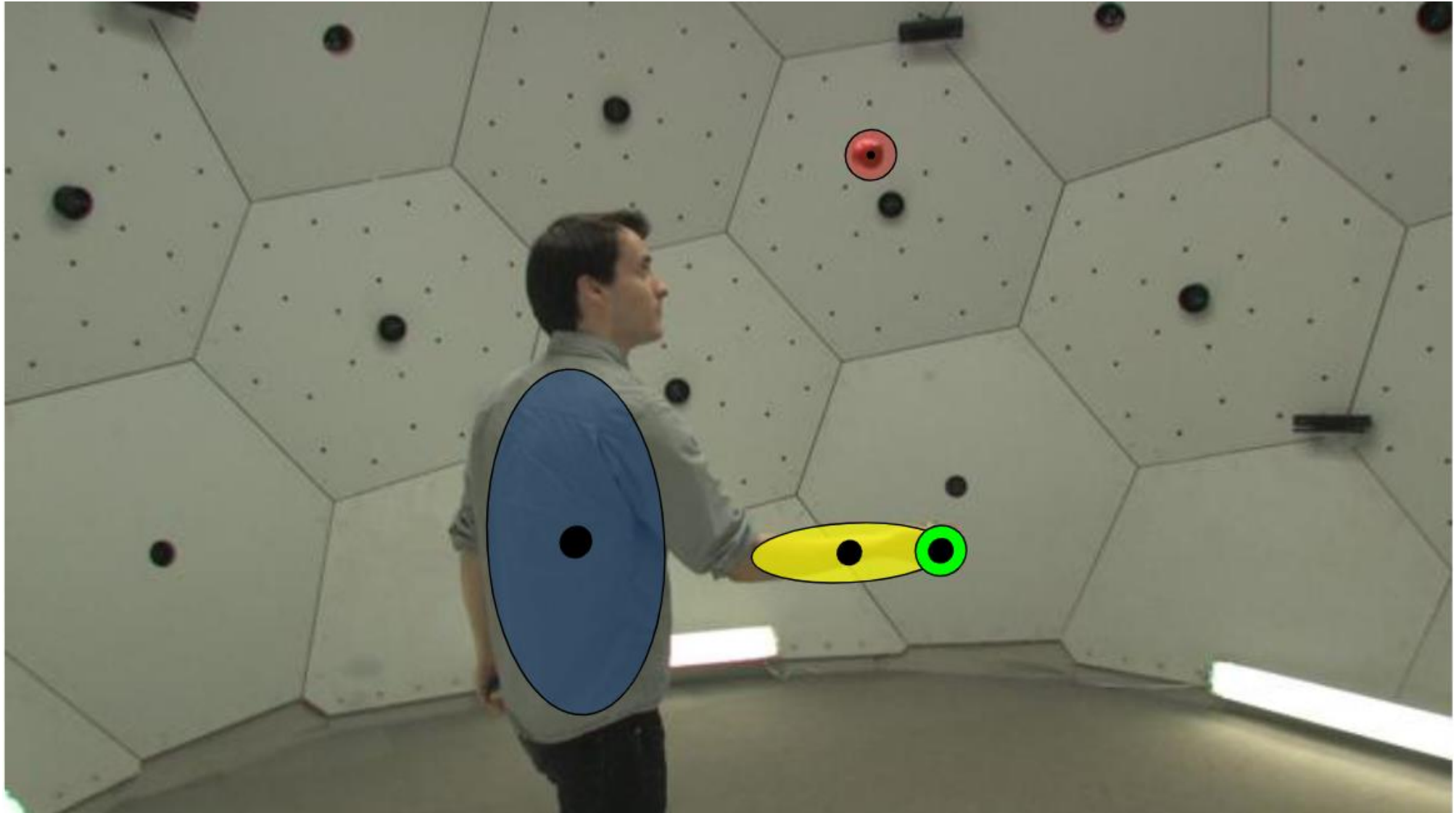
$$\Sigma^{2D} = J E \Sigma E^T J^T$$

$$C_{\text{pix}} = \sum_{i \in \mathcal{S}} c_i f_{i,\text{pix}}^{2D} \prod_{j=1}^{i-1} (1 - f_{j,\text{pix}}^{2D})$$

# Making 3D Gaussians Move



# Making 3D Gaussians Move





# Making 3D Gaussians Move

Fixed / Consistent over time:

3D Size

Color

Opacity

Changing over time (per timestep):

3D Center

3D Rotation

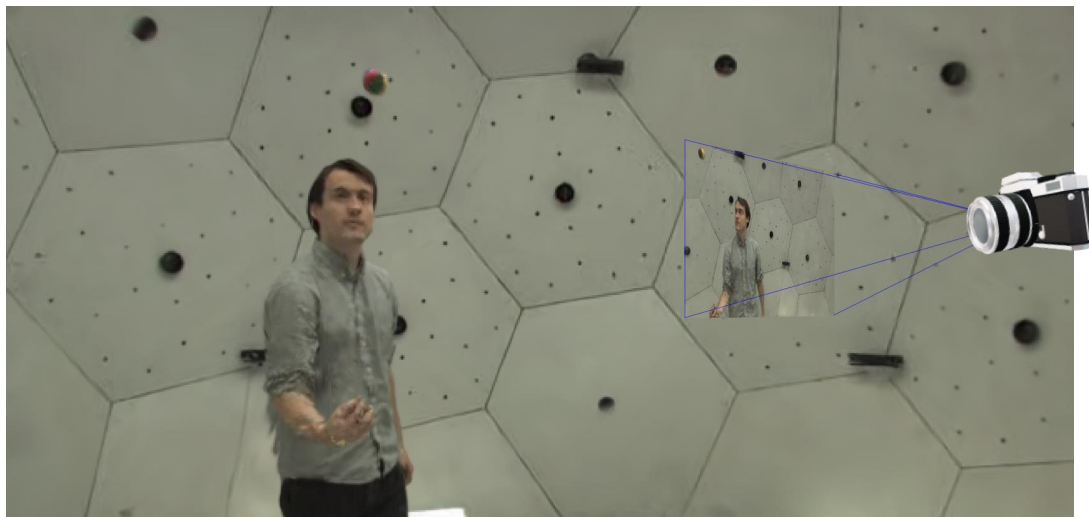
# What is needed?



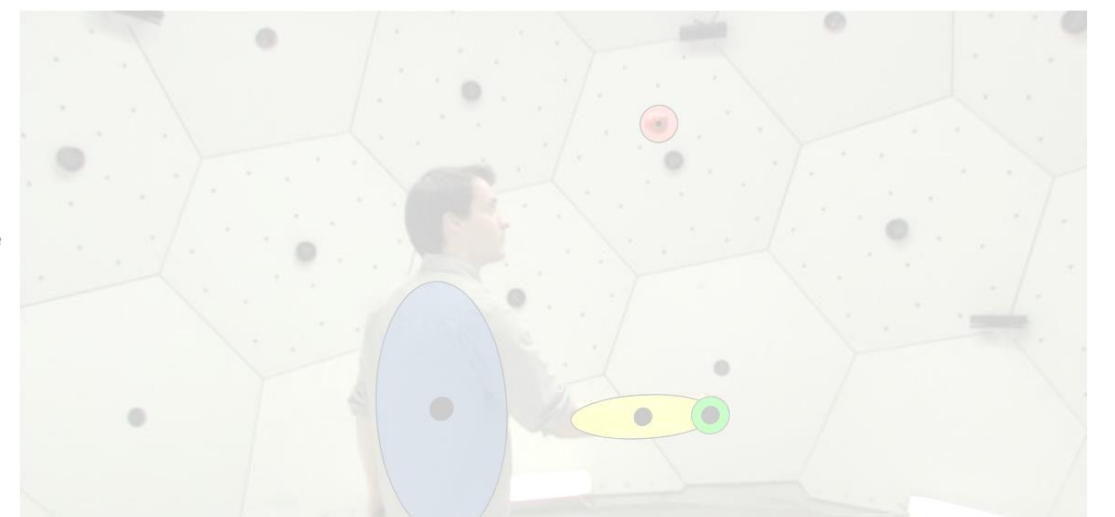
Good Data



A Good Representation

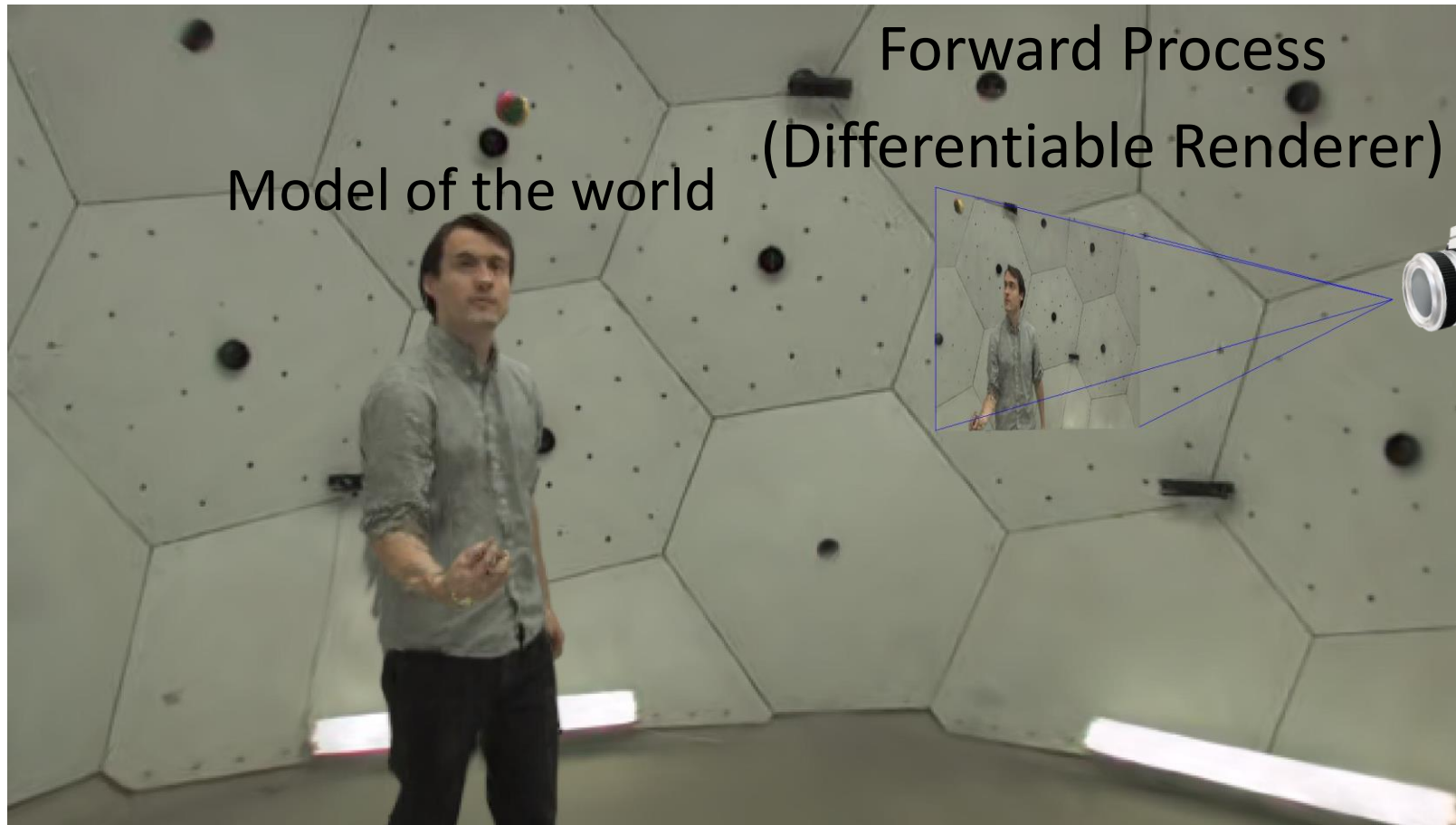


A Good Optimization Procedure

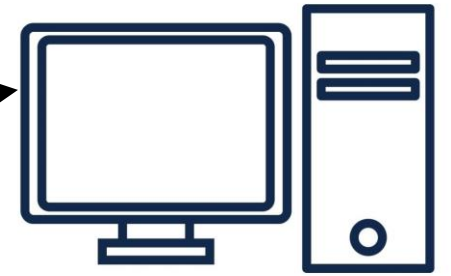


Good Priors

# A Good Optimization Procedure



Gradient-based  
Optimization  
(+ Physics-based Priors)



Analysis-by-Synthesis

# A Good Optimization Procedure



# What is needed?



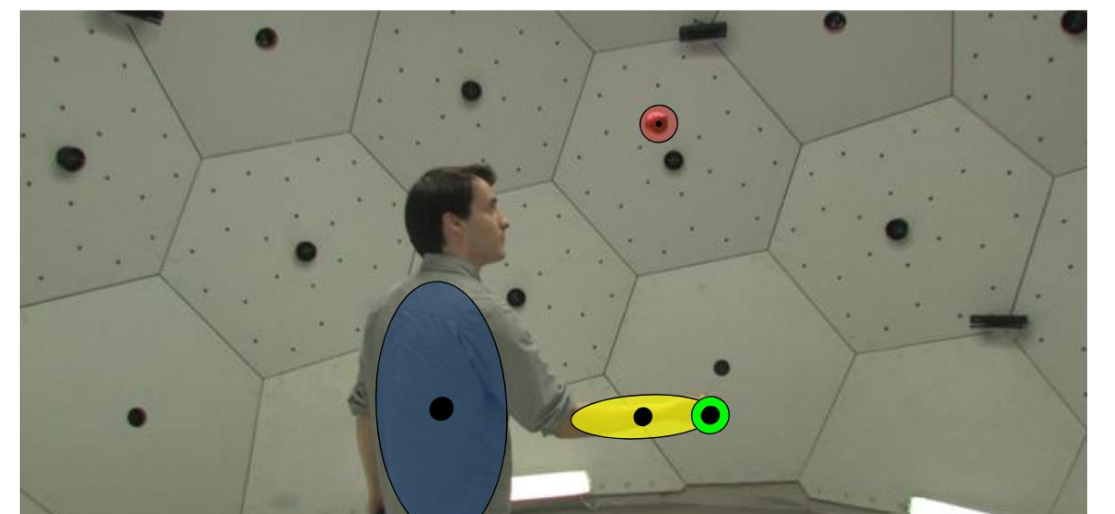
Good Data



A Good Representation



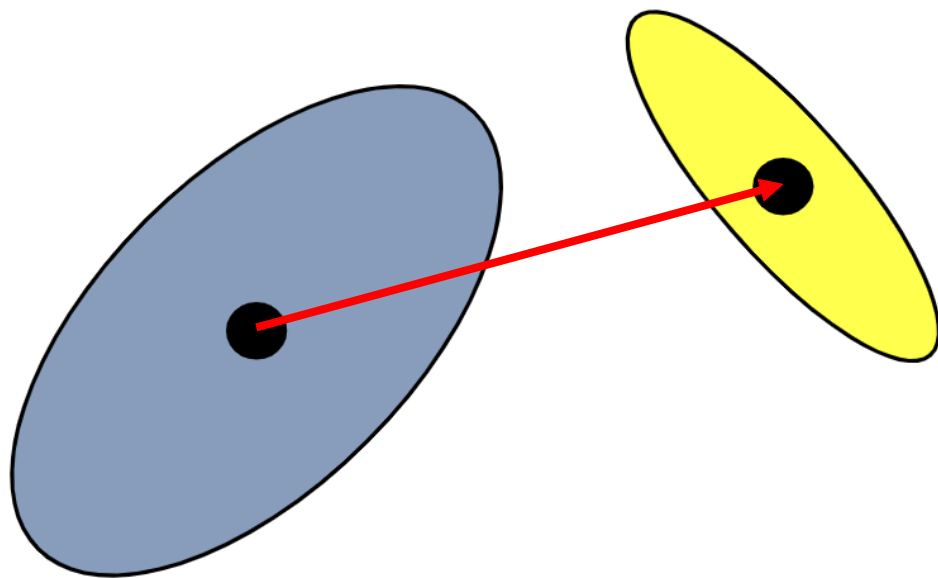
A Good Optimization Procedure



Good Priors

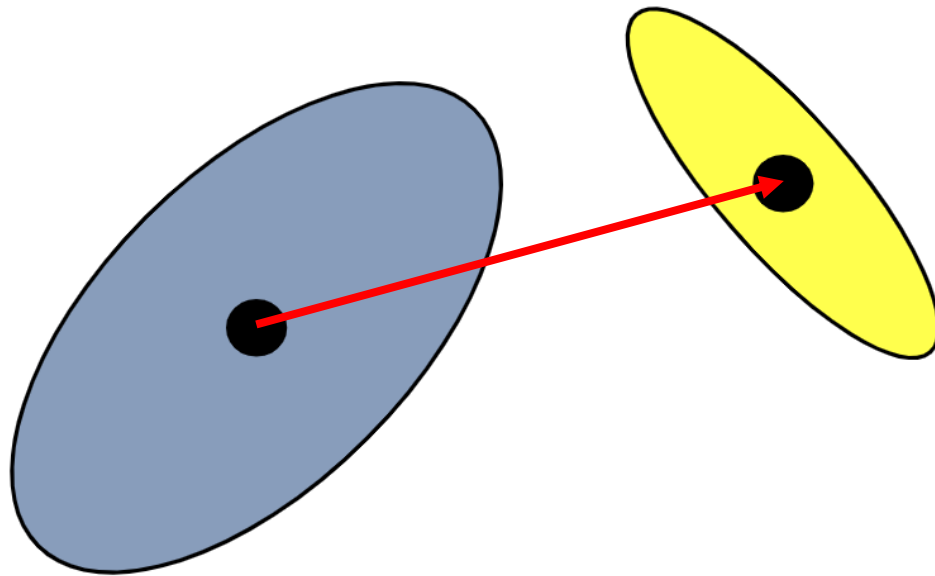
# Good Priors

$$w_{i,j} = \exp\left(-\lambda_w \|\mu_{j,0} - \mu_{i,0}\|_2^2\right)$$



# Good Priors

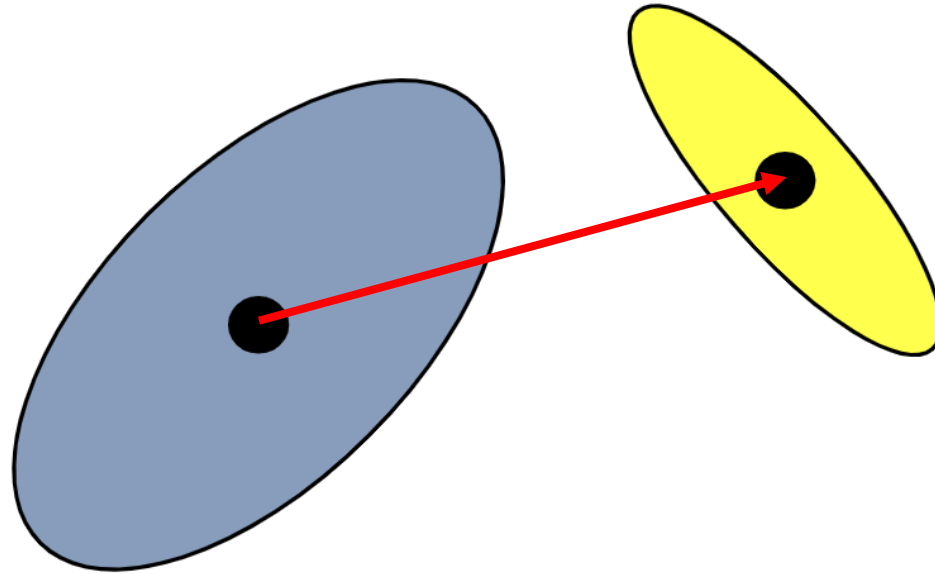
$t-1$



$$(\mu_{j,t-1} - \mu_{i,t-1})$$

# Good Priors

$t$

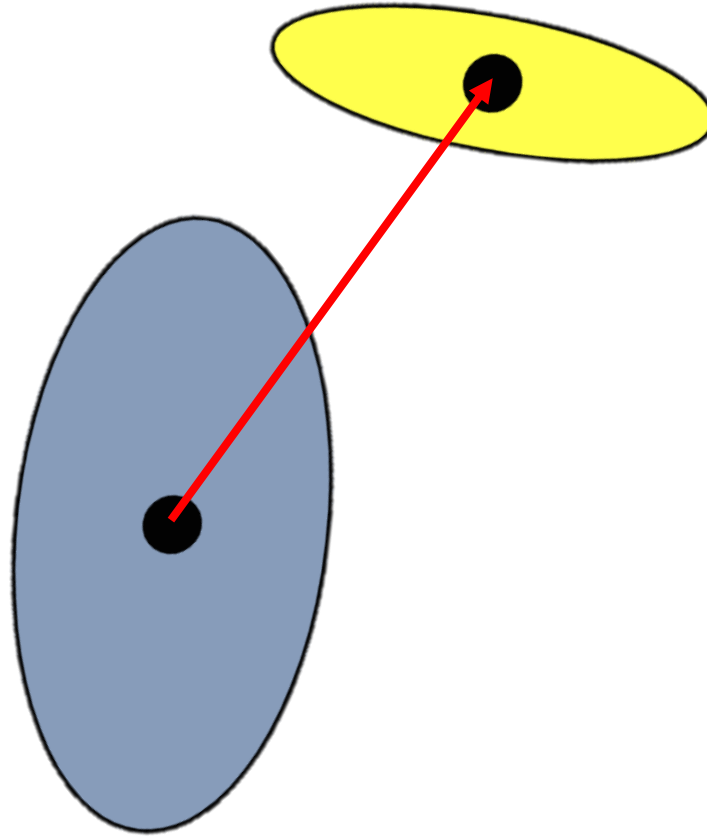


$$(\mu_{j,t} - \mu_{i,t})$$



# Good Priors

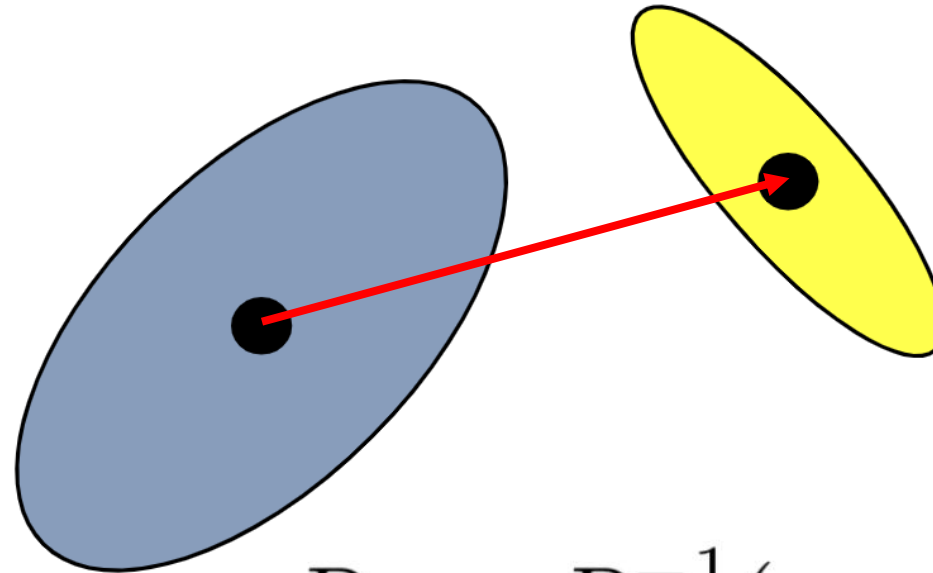
$t$



$$(\mu_{j,t} - \mu_{i,t})$$

# Good Priors

$t$



$$\|R_{i,t-1}R_{i,t}^{-1}(\mu_{j,t} - \mu_{i,t})\|_2$$

# Good Priors

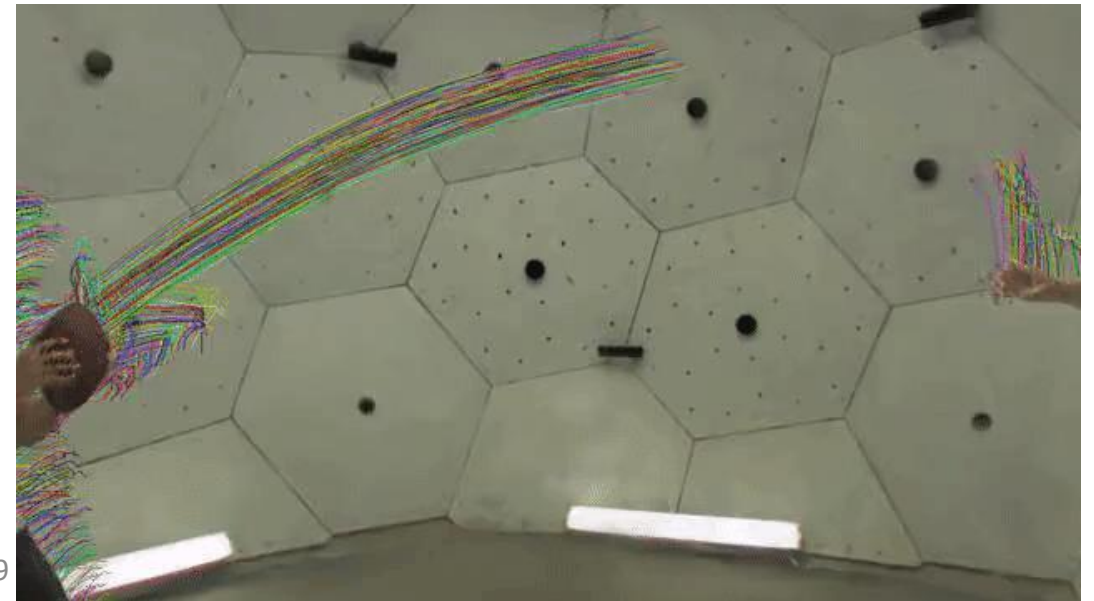
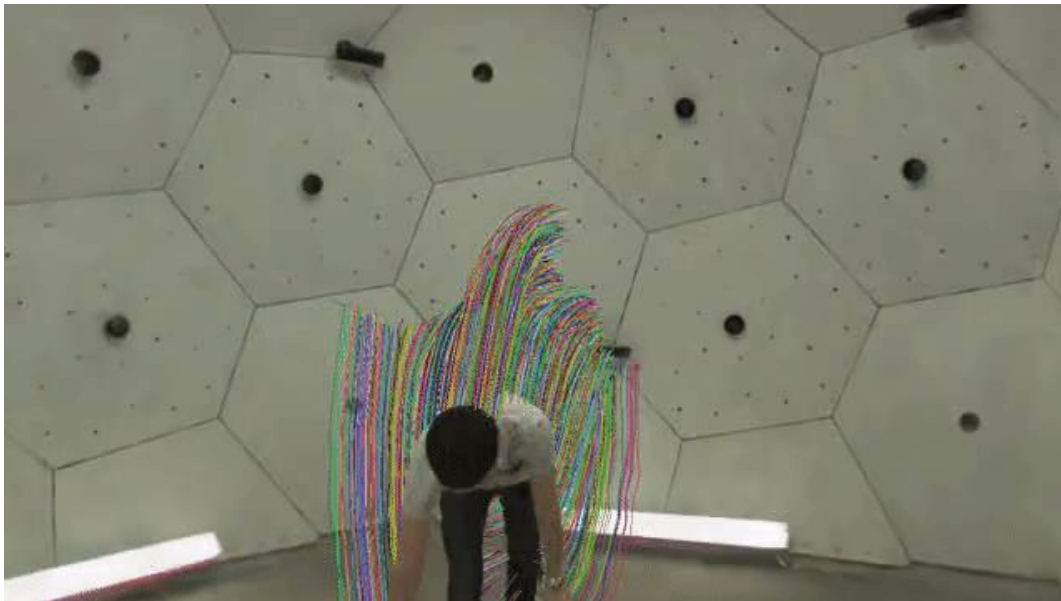
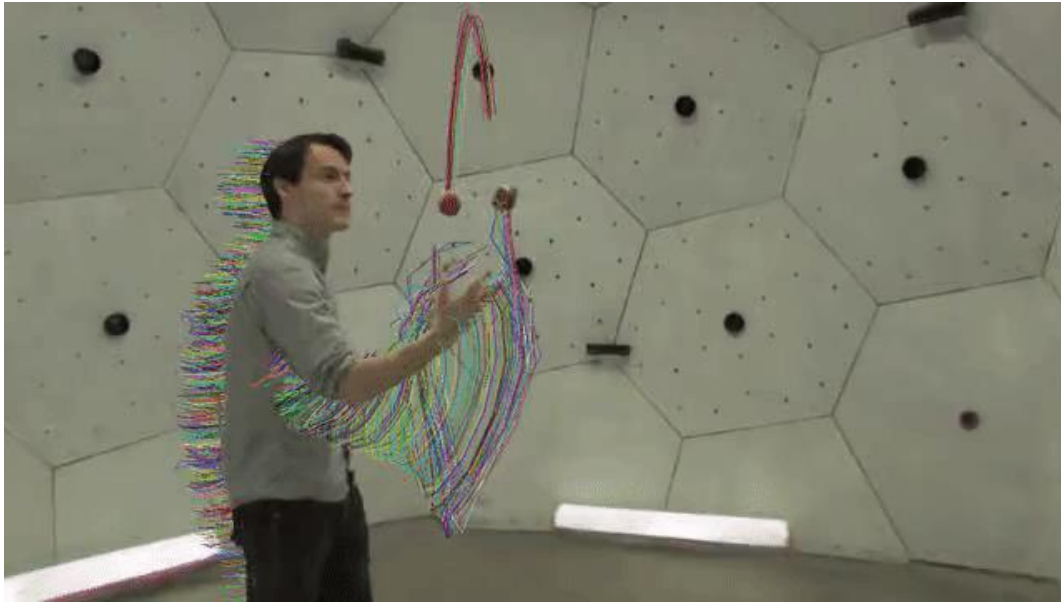
$$\mathcal{L}_{i,j}^{\text{rigid}} = w_{i,j} \left\| (\mu_{j,t-1} - \mu_{i,t-1}) - R_{i,t-1} R_{i,t}^{-1} (\mu_{j,t} - \mu_{i,t}) \right\|_2$$

$$\mathcal{L}^{\text{rigid}} = \frac{1}{k|\mathcal{S}|} \sum_{i \in \mathcal{S}} \sum_{j \in \text{knn}_{i;k}} \mathcal{L}_{i,j}^{\text{rigid}}$$

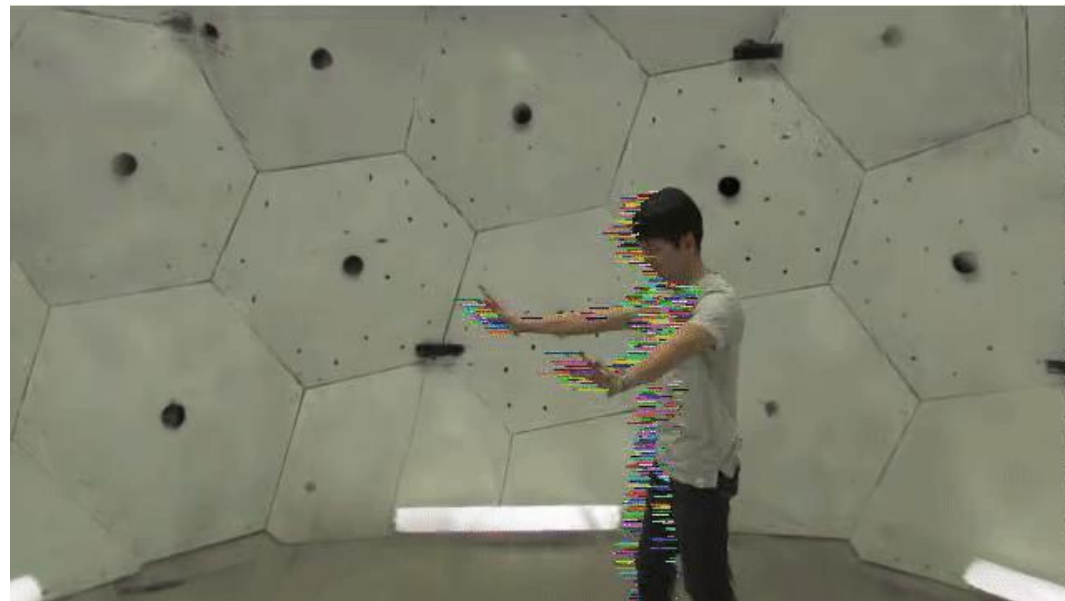
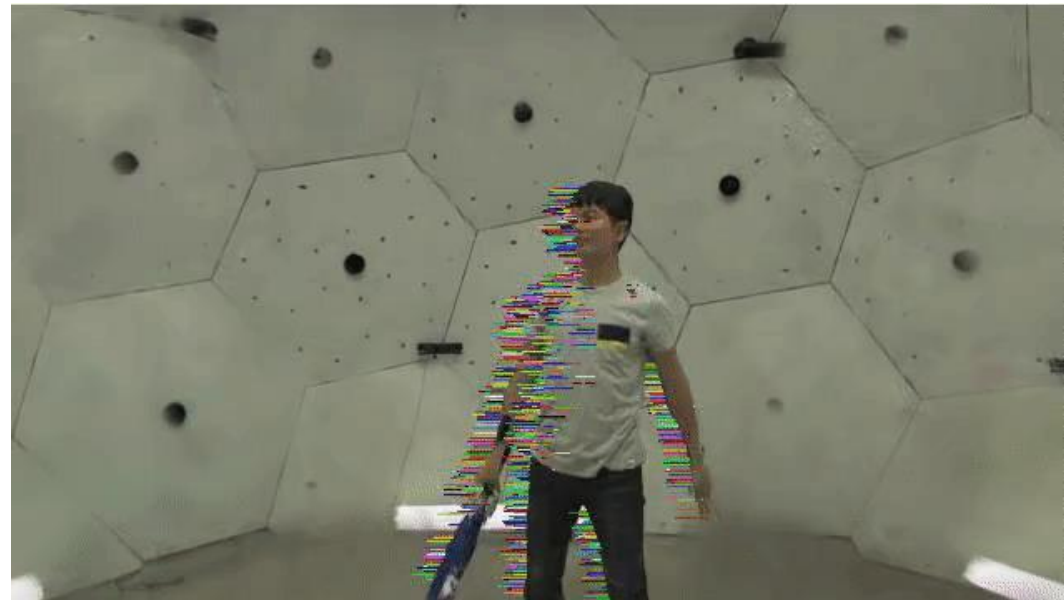
# Making 3D Gaussians Move



# Making 3D Gaussians Move



# Full 6-DoF Tracking



# It works!



It works!

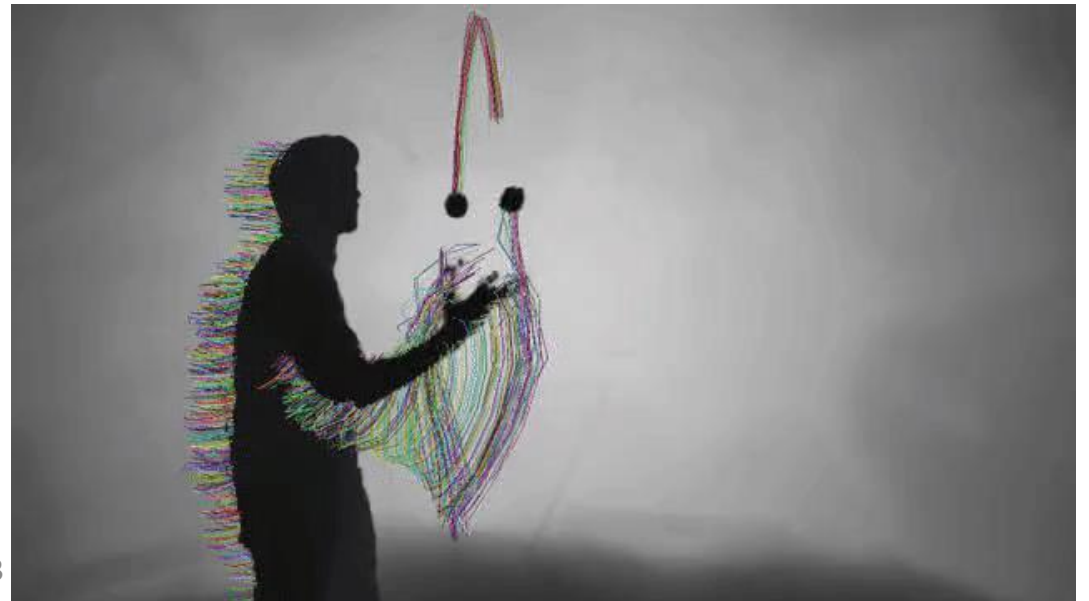
Median 3D Tracking Error: 1.90 cm

Median 2D Tracking Error: 1.54 pix

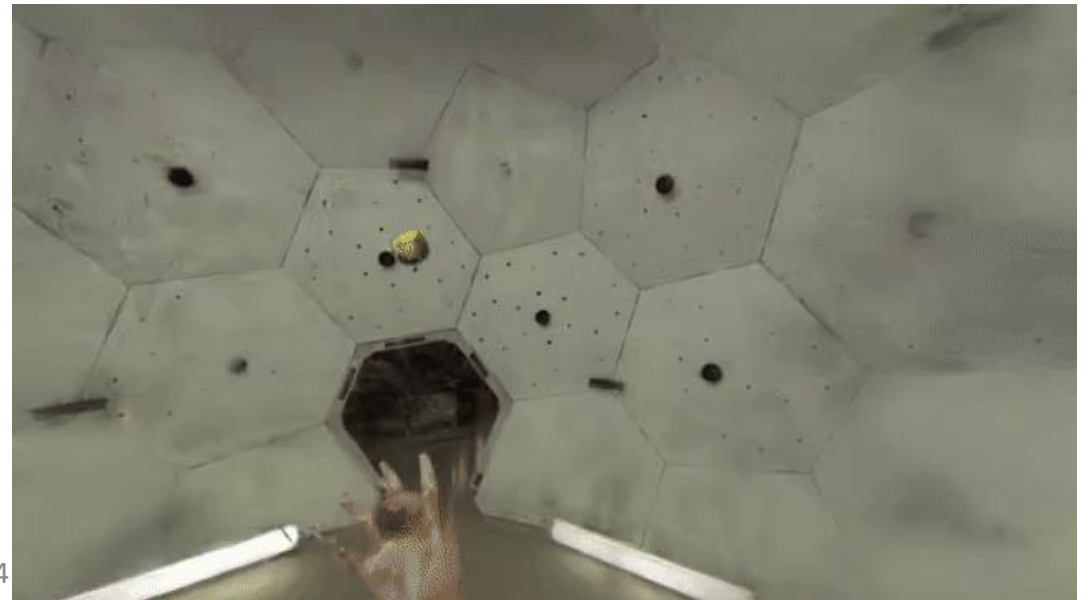
PSNR: 29.48 dB



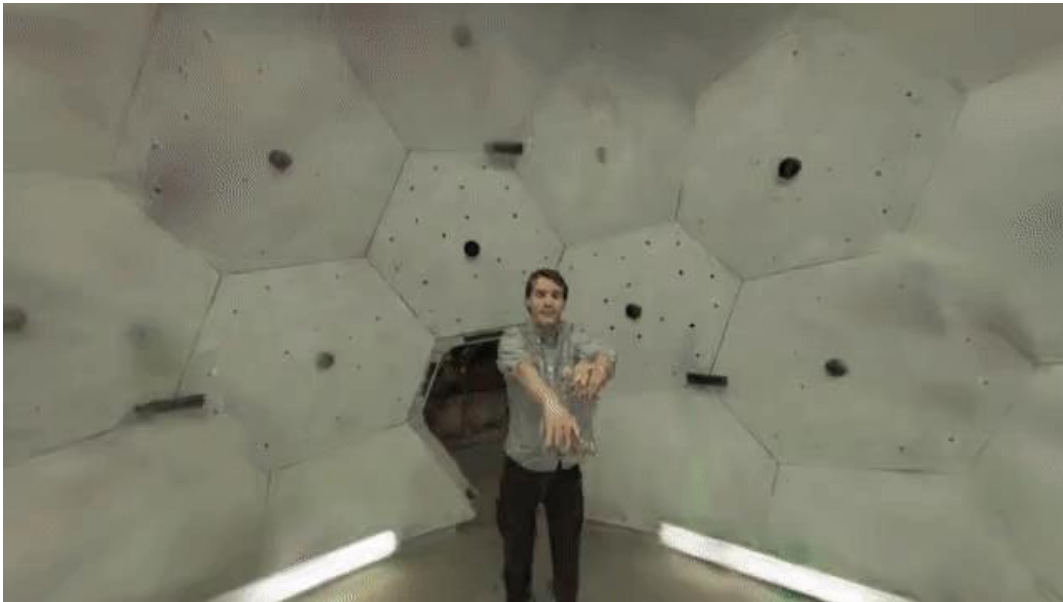
# Dense Metric 3D Tracking



# Creative Applications: Gaussian-eye view



# Creative Applications: Gaussian-eye view



# Creative Applications: Compositional Dynamic Scenes



# Creative Applications: Compositional Dynamic Scenes



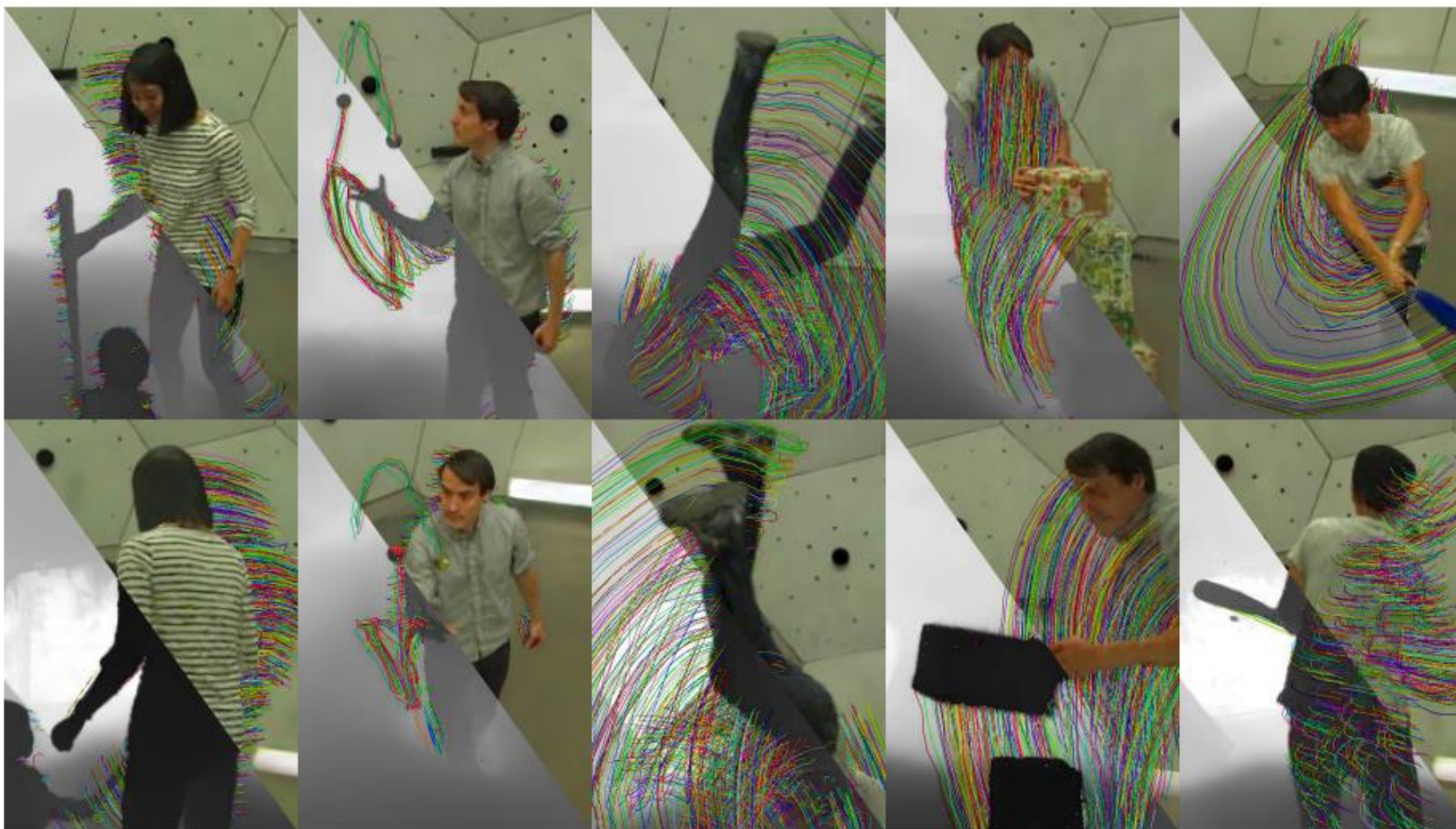
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yunchaoy@andrew.cmu.edu

Jia-Wei Liu

National University of Singapore

jiawei.liu@u.nus.edu

Mike Zheng Shou

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mike.zheng.shou@gmail.com

Shuran Song

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Carnegie Mellon University

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lizhan@pdx.edu

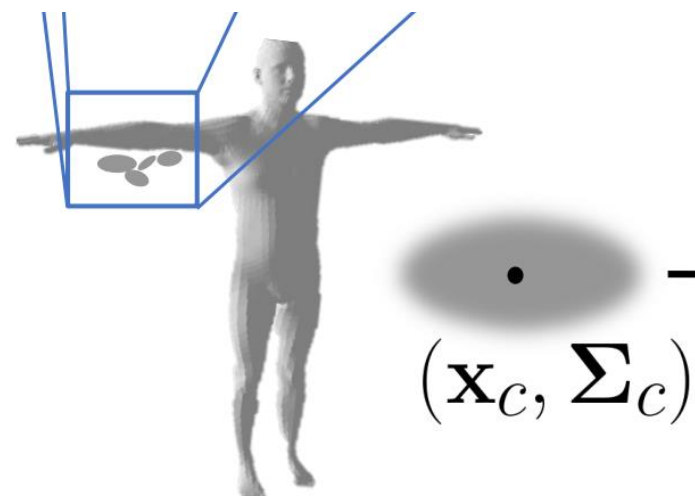
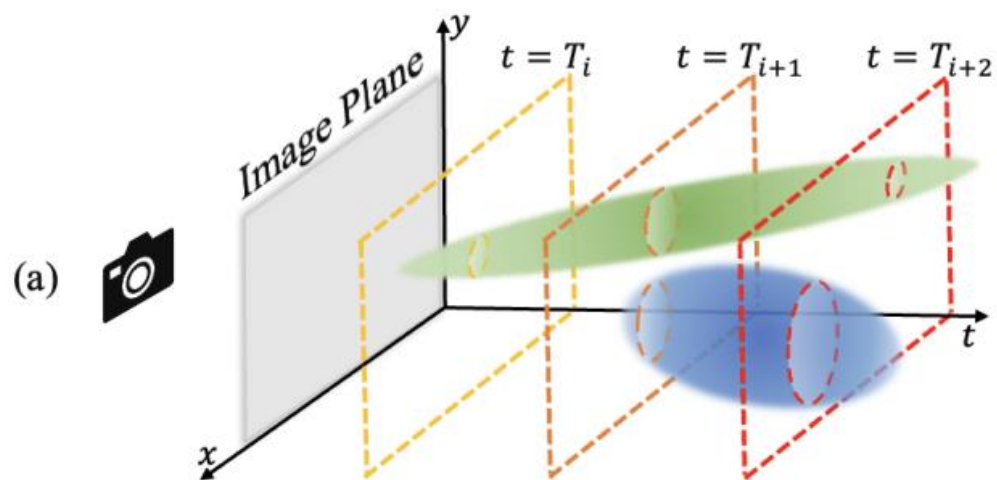
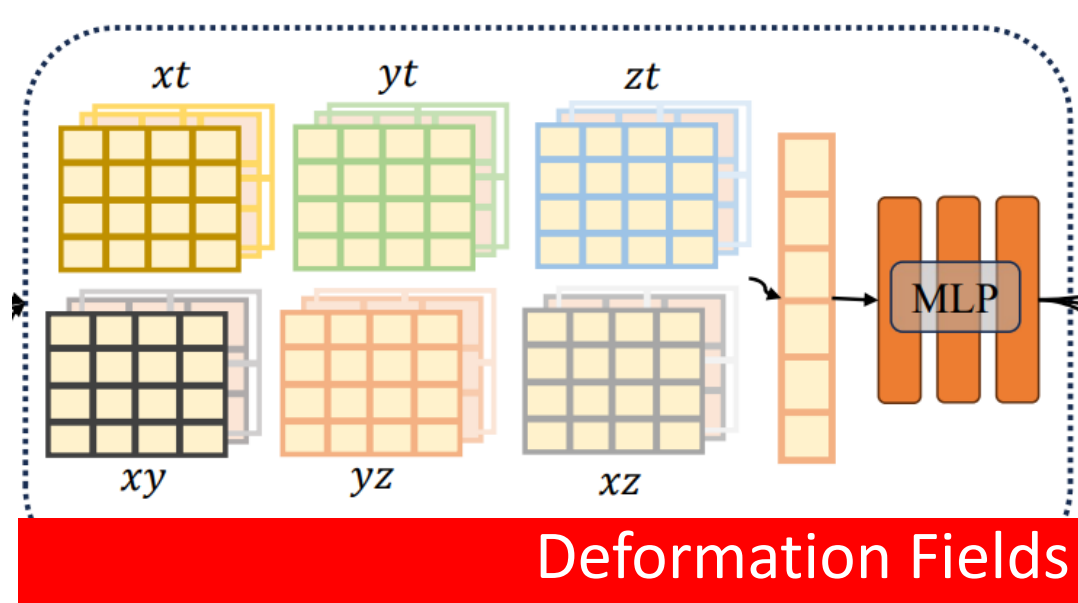
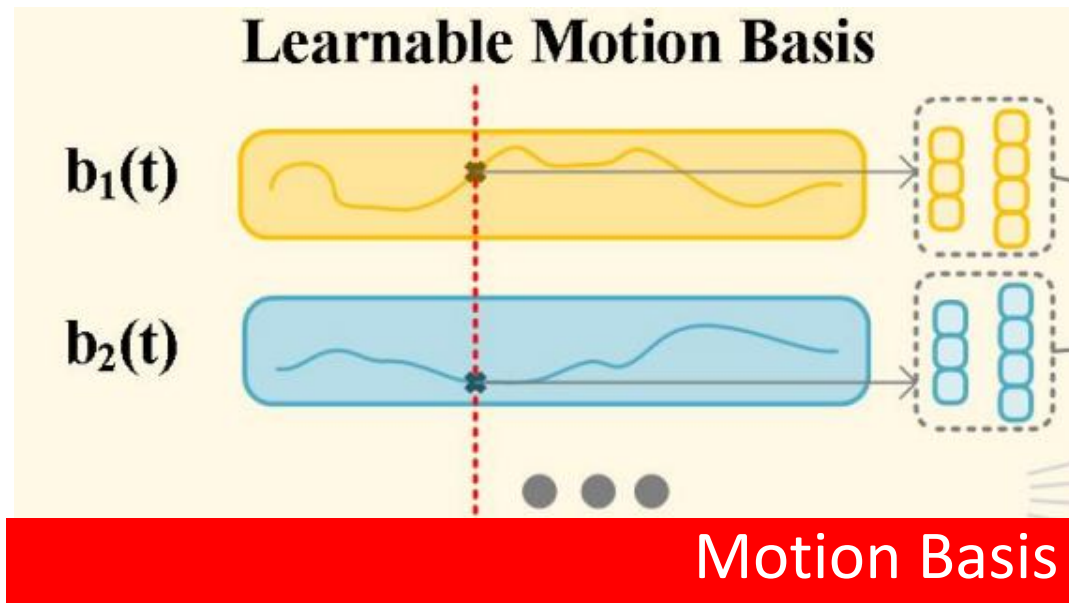
zhang.chen@oppo.com

zhong.li@oppo.com

yi.xu@oppo.com

<https://oppo-us-research.github.io/SpacetimeGaussians-website/>

# Alternative Dynamic Gaussian Representations

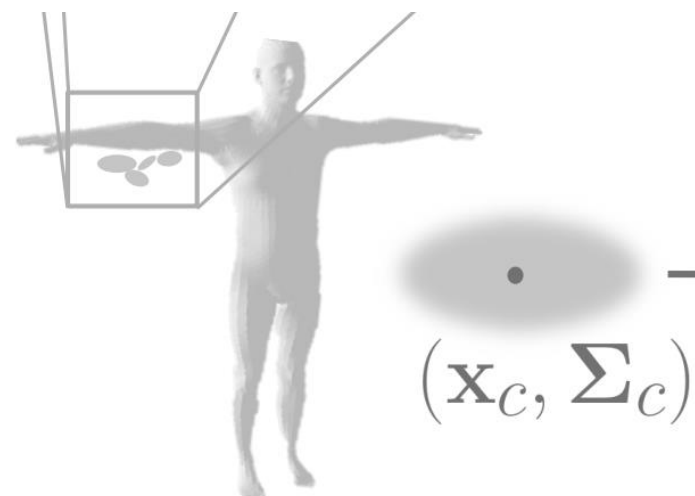
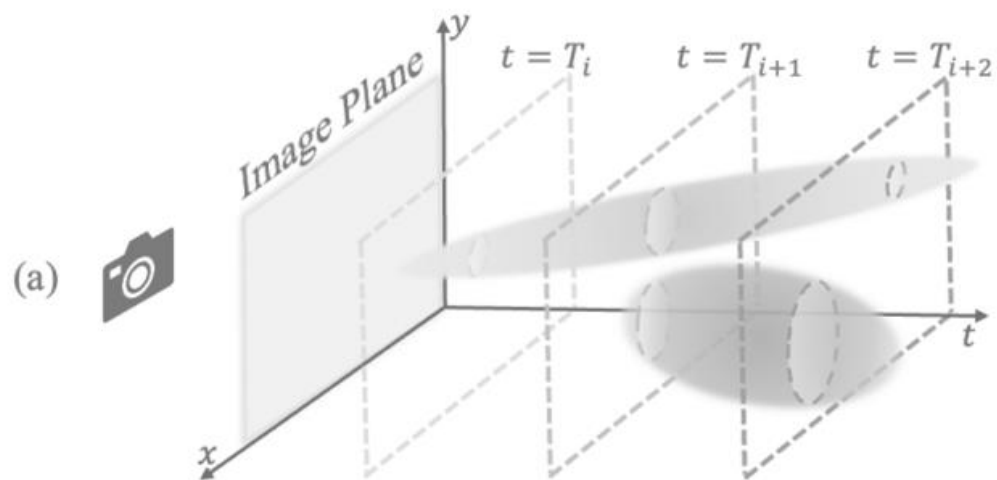
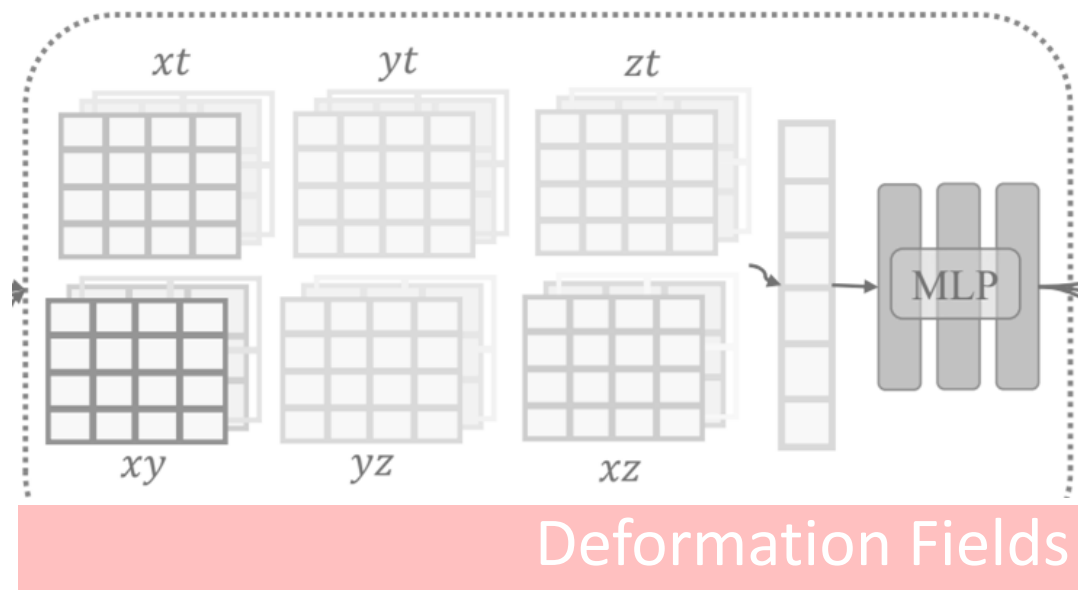
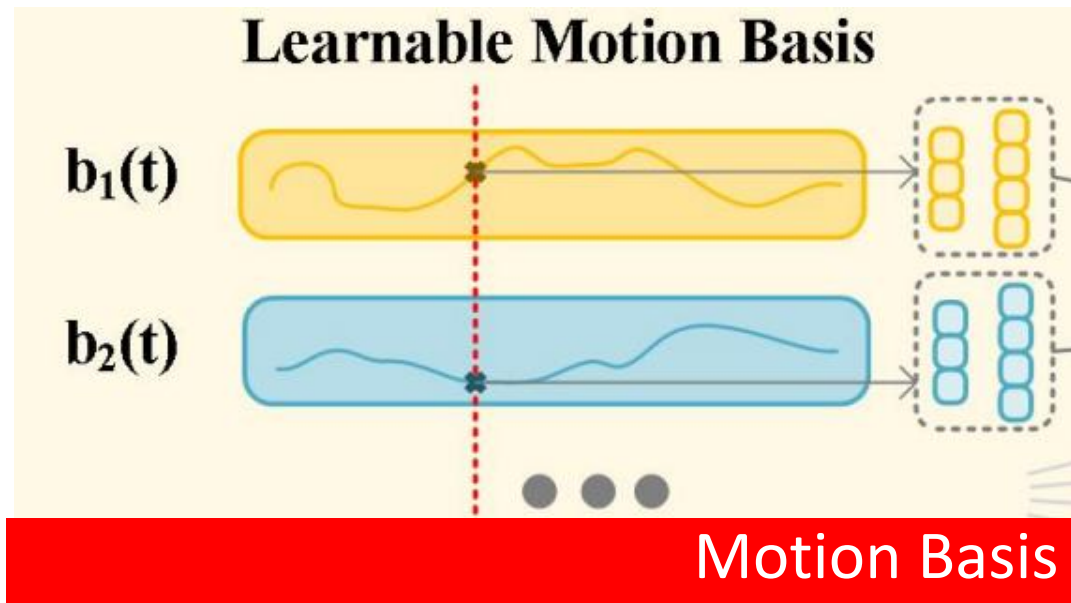


4D Gaussians

Shape Templates



# Alternative Dynamic Gaussian Representations



4D Gaussians 41

Shape Templates

# Motion Basis Representations

## **An Efficient 3D Gaussian Representation for Monocular/Multi-view Dynamic Scenes**

Kai Katsumata    Duc Minh Vo    Hideki Nakayama

The University of Tokyo, Japan

{katsumata, vmduc, nakayama}@nlab.ci.i.u-tokyo.ac.jp

## Motion as Fourier Coefficients / Optical Flow supervision

$$x(t) = w_{x,0} + \sum_{i=1}^L w_{x,2i-1} \sin(\pi t) + w_{x,2i} \cos(\pi t),$$

$$y(t) = w_{y,0} + \sum_{i=1}^L w_{y,2i-1} \sin(\pi t) + w_{y,2i} \cos(\pi t),$$

$$z(t) = w_{z,0} + \sum_{i=1}^L w_{z,2i-1} \sin(\pi t) + w_{z,2i} \cos(\pi t),$$

# Motion Basis Representations

## Spacetime Gaussian Feature Splatting for Real-Time Dynamic View Synthesis

Zhan Li<sup>1,2\*</sup>

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Yi Xu<sup>1</sup>

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<sup>2</sup> Portland State University

[lizhan@pdx.edu](mailto:lizhan@pdx.edu)

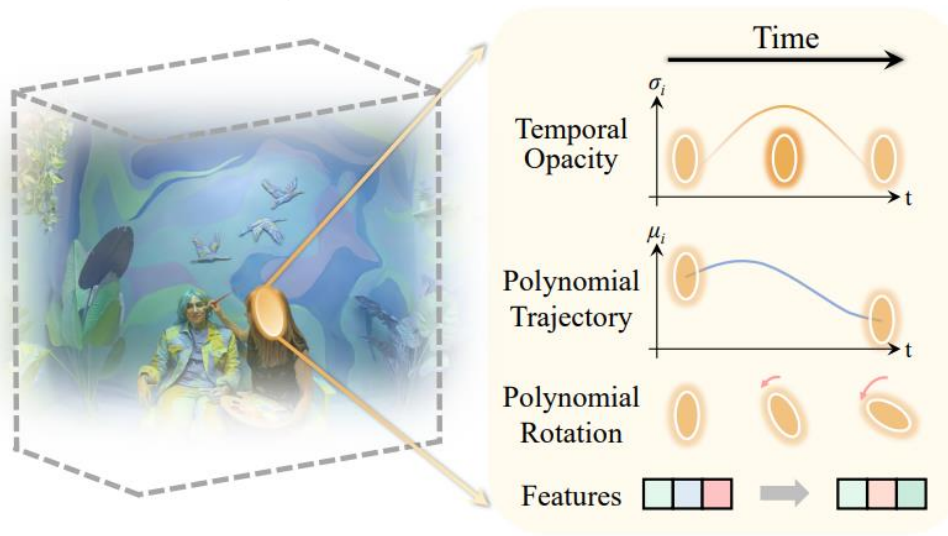
[zhang.chen@oppo.com](mailto:zhang.chen@oppo.com)

[zhong.li@oppo.com](mailto:zhong.li@oppo.com)

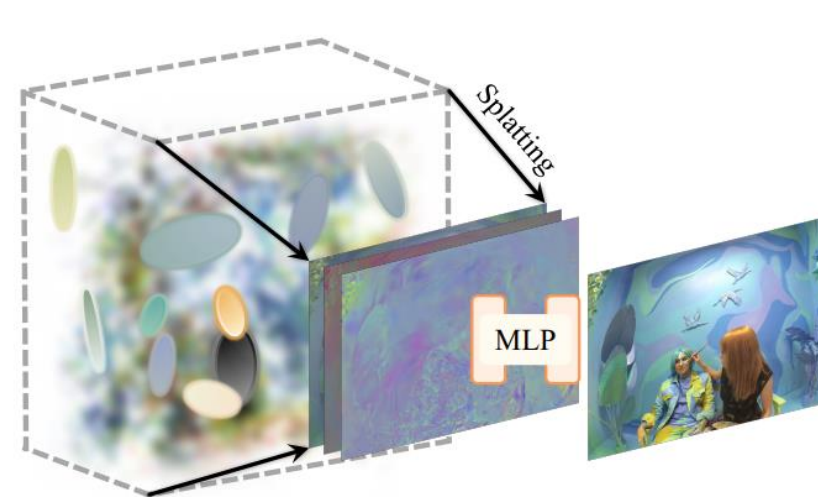
[yi.xu@oppo.com](mailto:yi.xu@oppo.com)

<https://oppo-us-research.github.io/SpacetimeGaussians-website/>

Motion as Polynomial Coefficients / Temporally Local Opacity /  
Splats Features instead of Colors.



(a) Spacetime Gaussians



(b) Feature Splatting and Rendering

# Motion Basis Representations

## Gaussian-Flow: 4D Reconstruction with Dynamic 3D Gaussian Particle

Youtian Lin<sup>1</sup>

Zuozhuo Dai<sup>2</sup>

Siyu Zhu<sup>3</sup>

Yao Yao<sup>1✉</sup>

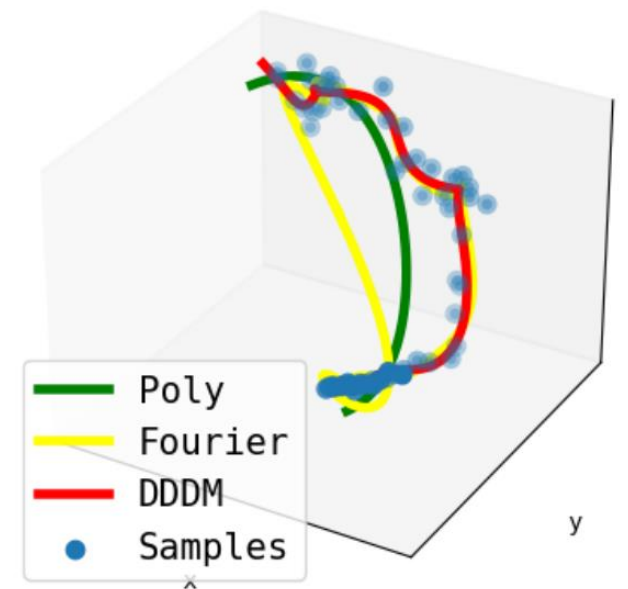
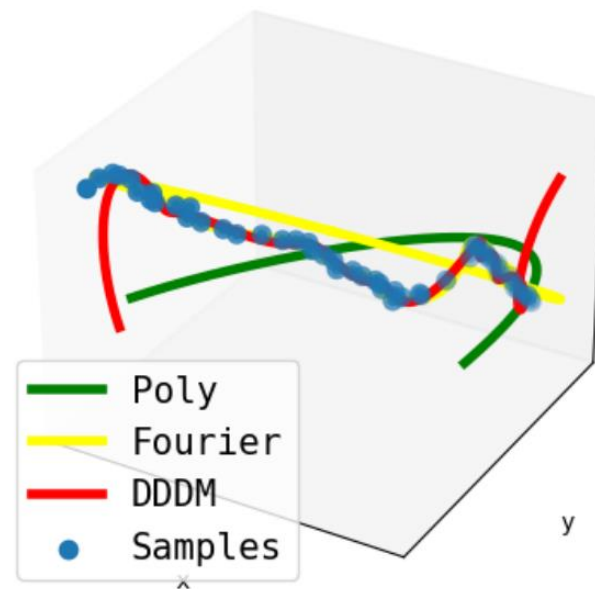
<sup>1</sup>Nanjing University

<sup>2</sup>Alibaba Group

<sup>3</sup>Fudan University

Combines Polynomial + Fourier Coefficients for modelling motion

$$D(t) = P_N(t) + F_L(t)$$



# Motion Basis Representations

## DynMF: Neural Motion Factorization

### for Real-time Dynamic View Synthesis with 3D Gaussian Splatting

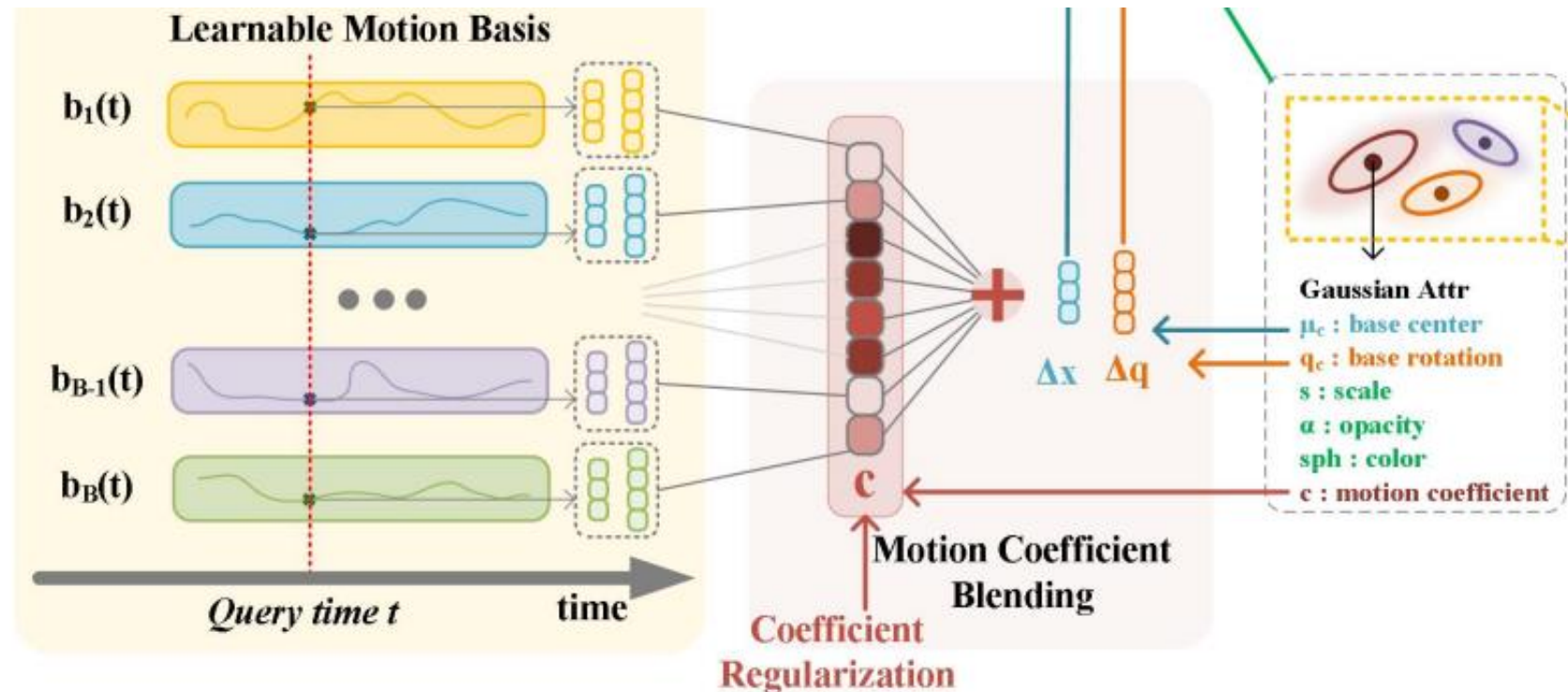
Agelos Kratimenos

Jiahui Lei

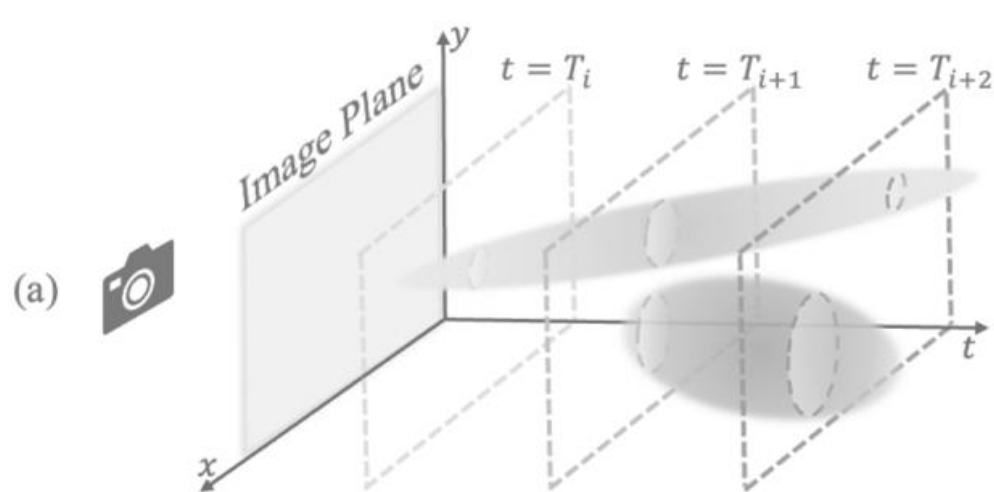
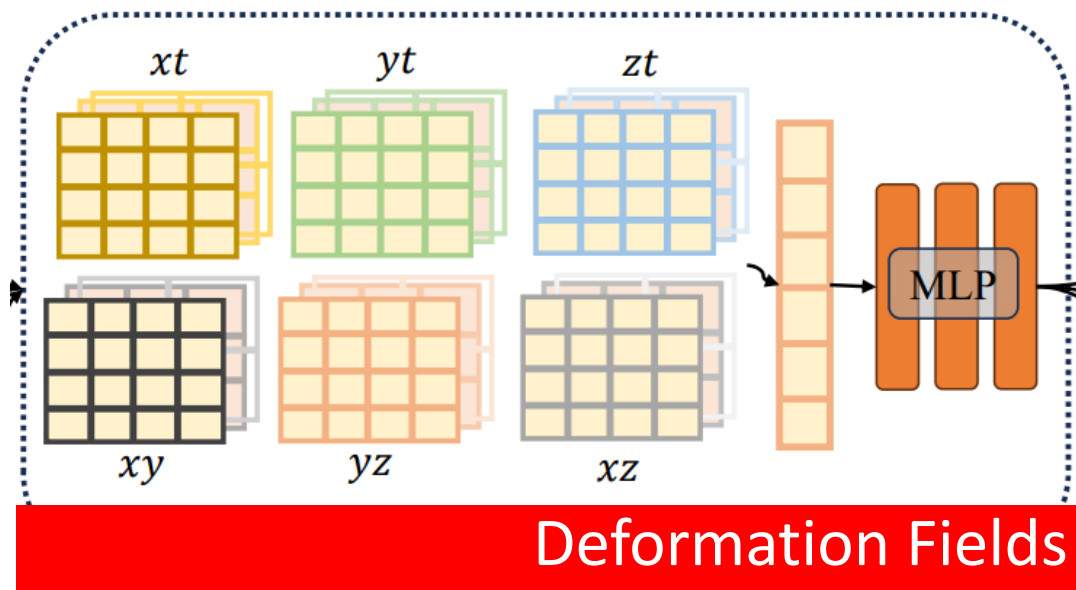
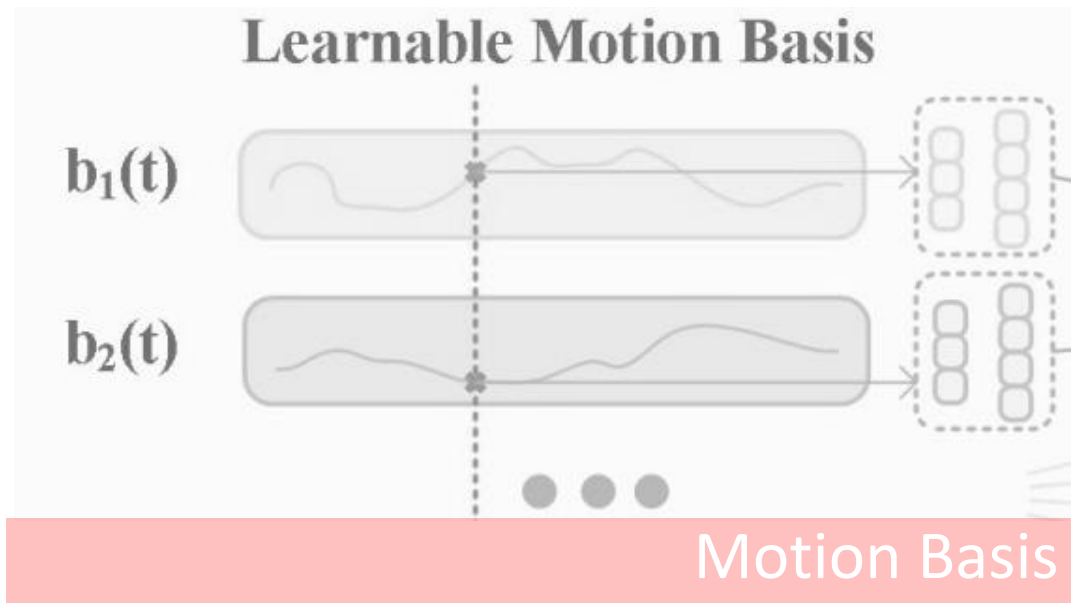
Kostas Daniilidis

University of Pennsylvania

Uses MLPs to represent small basis set / Each Gaussians motion is linear combo of MLP bases. Bases can be sparse (10 or 16).



# Alternative Dynamic Gaussian Representations



4D Gaussians

Shape Templates

# Deformation Field Representations

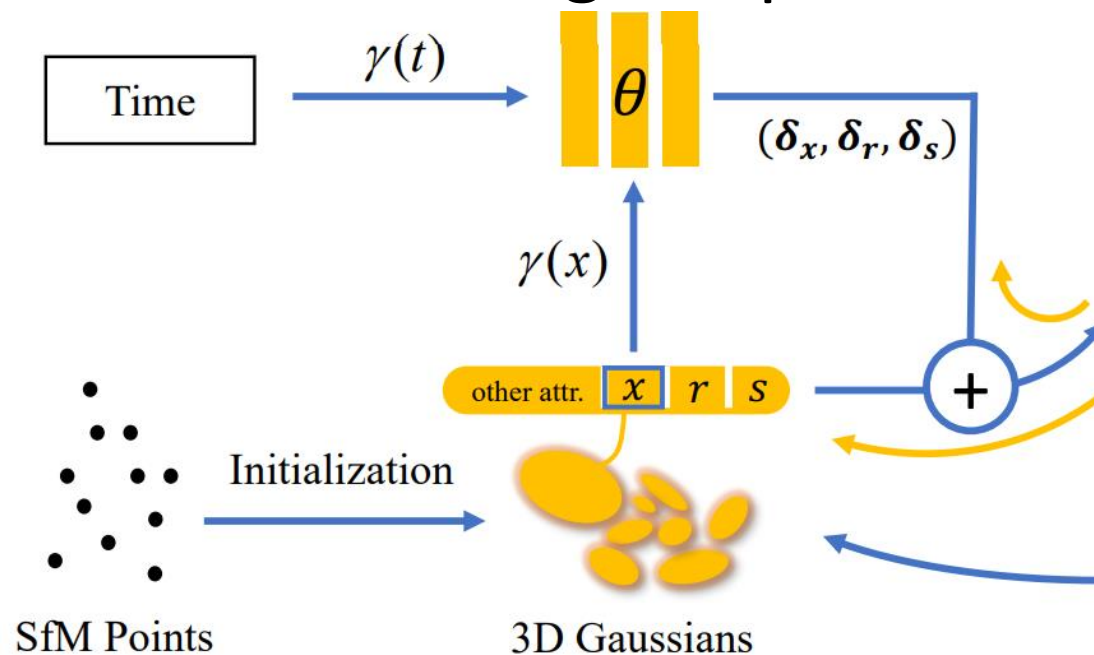
## Deformable 3D Gaussians for High-Fidelity Monocular Dynamic Scene Reconstruction

Ziyi Yang<sup>1,2</sup> Xinyu Gao<sup>1</sup> Wen Zhou<sup>2</sup> Shaohui Jiao<sup>2</sup> Yuqing Zhang<sup>1</sup> Xiaogang Jin<sup>1†</sup>

<sup>1</sup>State Key Laboratory of CAD&CG, Zhejiang University <sup>2</sup>ByteDance Inc.

Dense MLP representation over space / time defining the push-forward deformation of Gaussians

$$(\delta \mathbf{x}, \delta \mathbf{r}, \delta \mathbf{s}) = \mathcal{F}_{\theta}(\gamma(\text{sg}(\mathbf{x})), \gamma(t)),$$

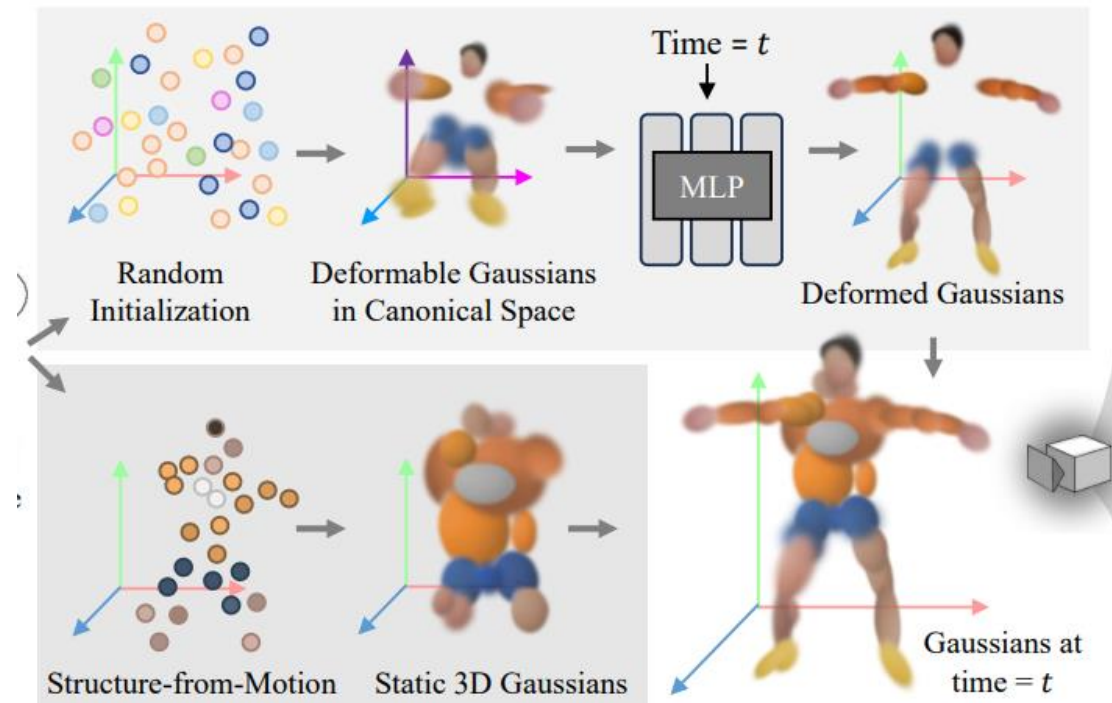


# Deformation Field Representations

## GauFRE 🍷: Gaussian Deformation Fields for Real-time Dynamic Novel View Synthesis

Yiqing Liang<sup>‡</sup>, Numair Khan, Zhengqin Li, Thu Nguyen-Phuoc,  
Douglas Lanman, James Tompkin<sup>‡</sup>, Lei Xiao  
Meta      <sup>‡</sup>Brown University

Adds a set of static Gaussians that cannot move.





# Deformation Field Representations

## 4D Gaussian Splatting for Real-Time Dynamic Scene Rendering

Guanjun Wu<sup>1\*</sup>, Taoran Yi<sup>2\*</sup>, Jiemin Fang<sup>3†</sup>, Lingxi Xie<sup>3</sup>, Xiaopeng Zhang<sup>3</sup>,  
Wei Wei<sup>1</sup>, Wenyu Liu<sup>2</sup>, Qi Tian<sup>3</sup>, Xinggang Wang<sup>2‡</sup>

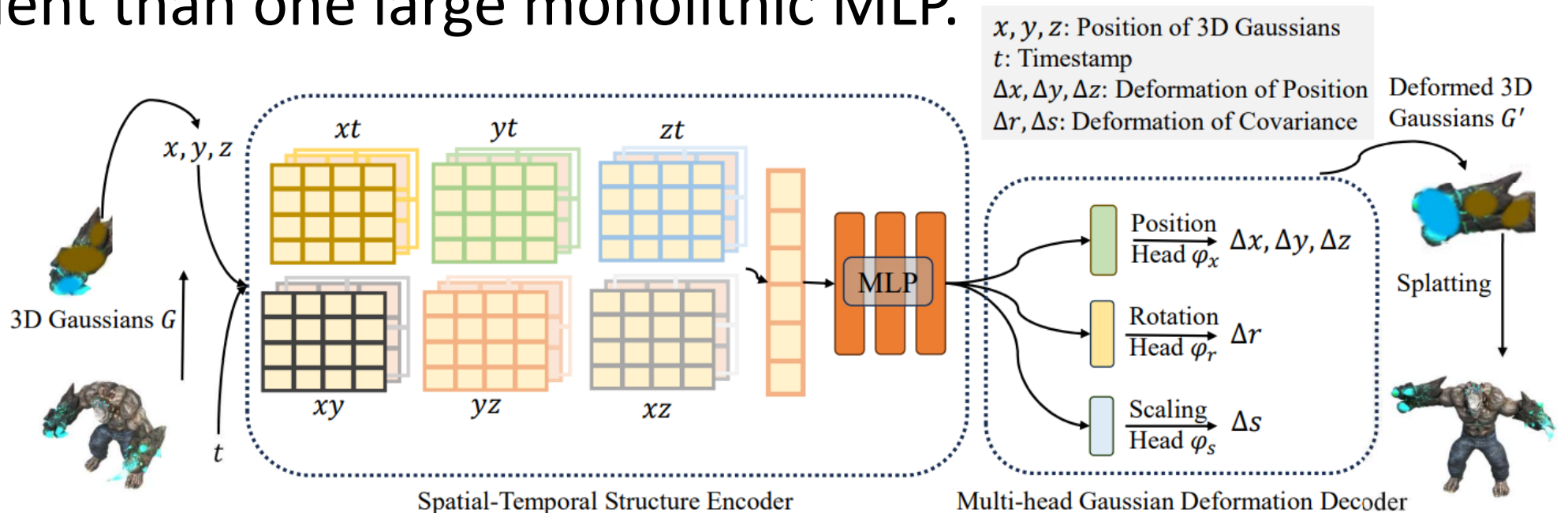
<sup>1</sup>School of CS, Huazhong University of Science and Technology

<sup>2</sup>School of EIC, Huazhong University of Science and Technology <sup>3</sup>Huawei Inc.

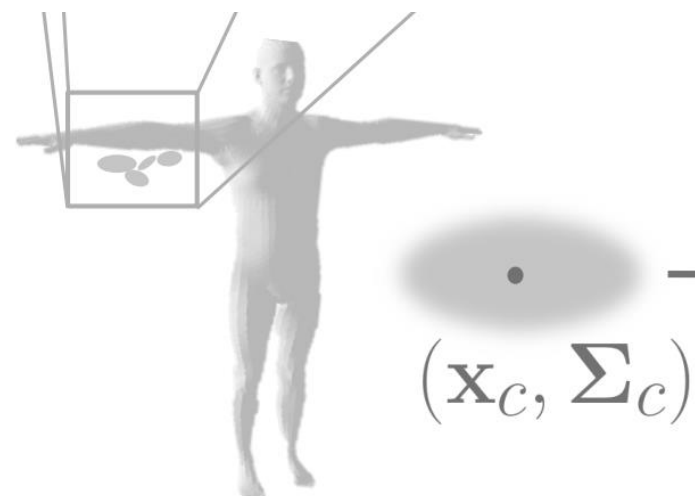
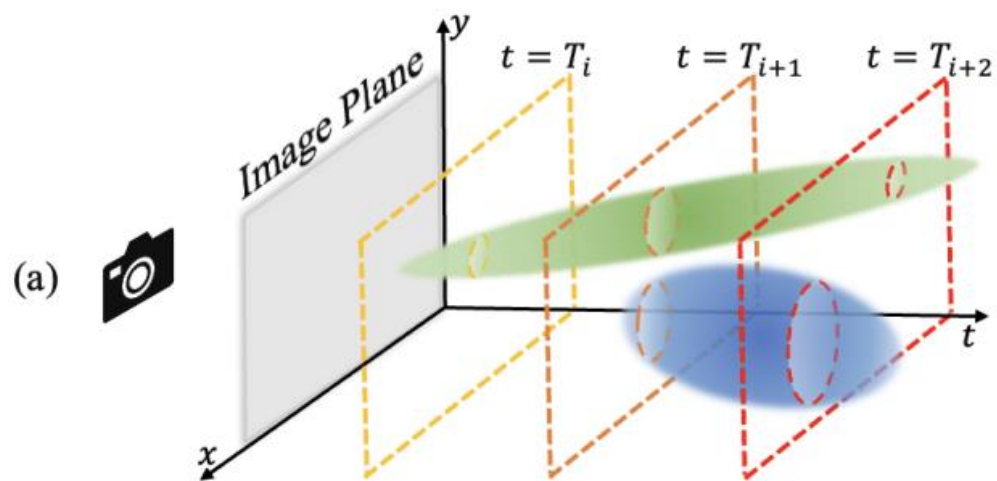
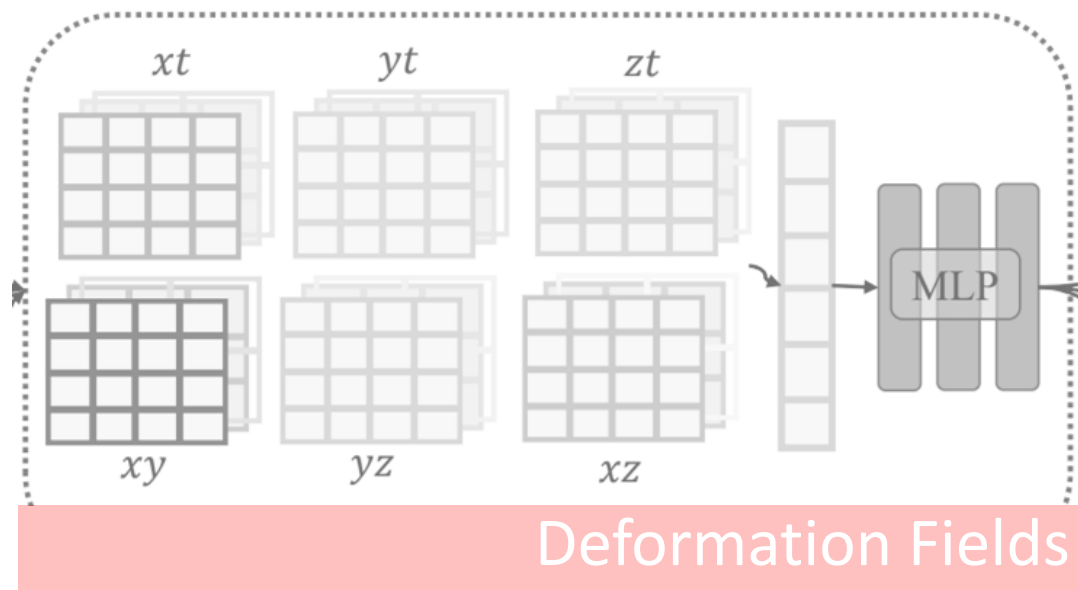
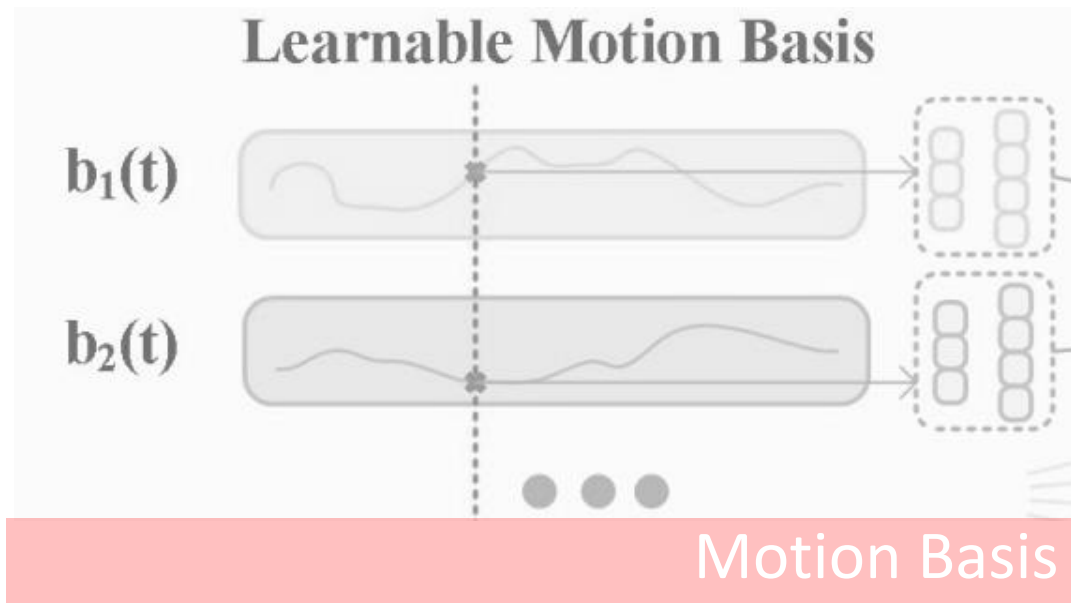
{guajuwu, taoranyi, weiw, liuw, xgwang}@hust.edu.cn

{jaminfong, 198808xc, zxphistory}@gmail.com tian.qil@huawei.com

Using Multi-Res Hex-plane + tiny MLP for push-forward deformation is more efficient than one large monolithic MLP.



# Alternative Dynamic Gaussian Representations



4D Gaussians

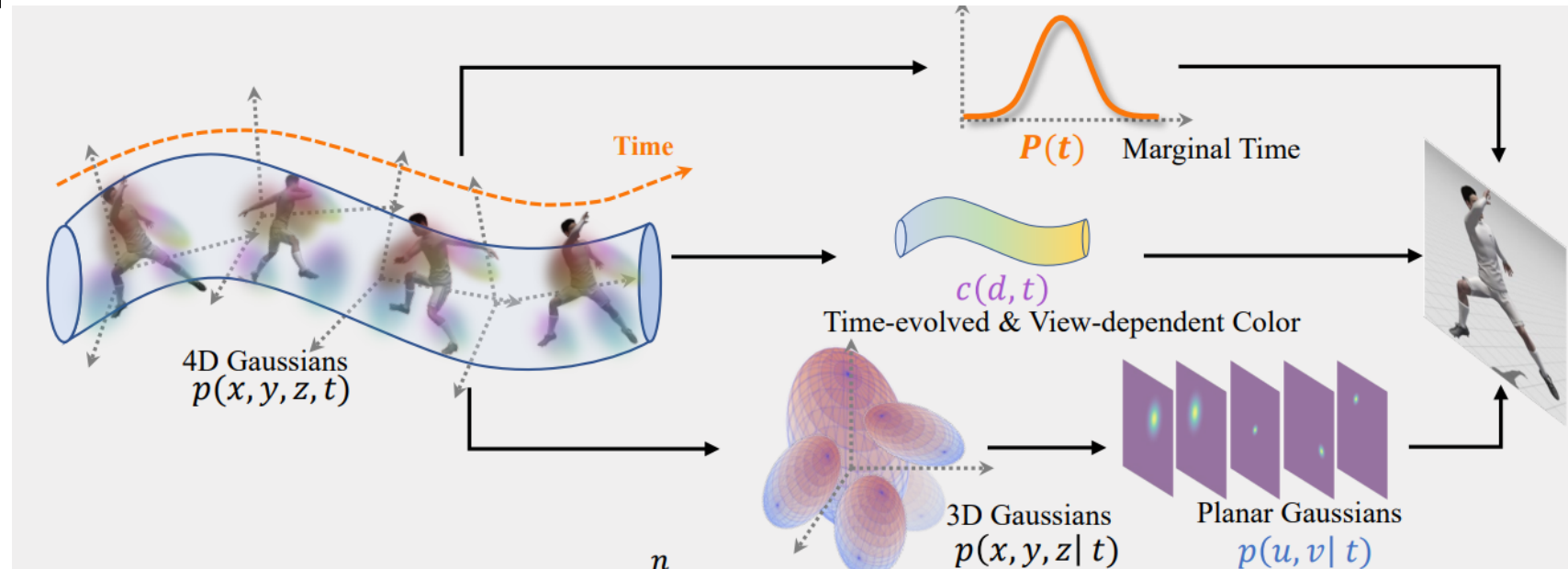
Shape Templates

# 4D Gaussian Representations

## REAL-TIME PHOTOREALISTIC DYNAMIC SCENE REPRESENTATION AND RENDERING WITH 4D GAUSSIAN SPLATTING

Zeyu Yang, Hongye Yang, Zijie Pan, Li Zhang\*  
Fudan University

Represents scenes as actual 4D Gaussians, which are spliced into 3D Gaussians per timestep

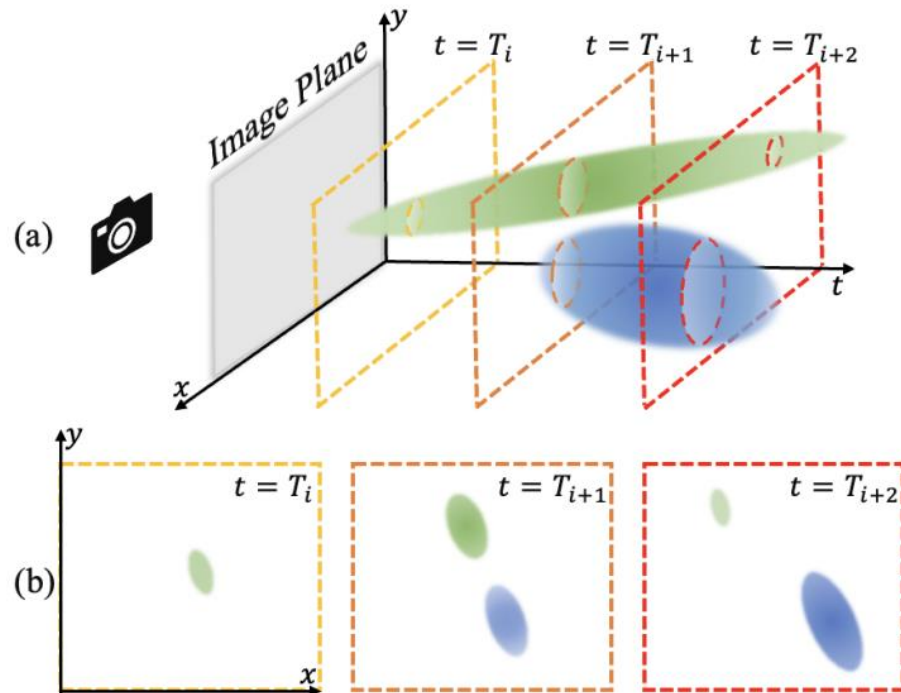


# 4D Gaussian Representations

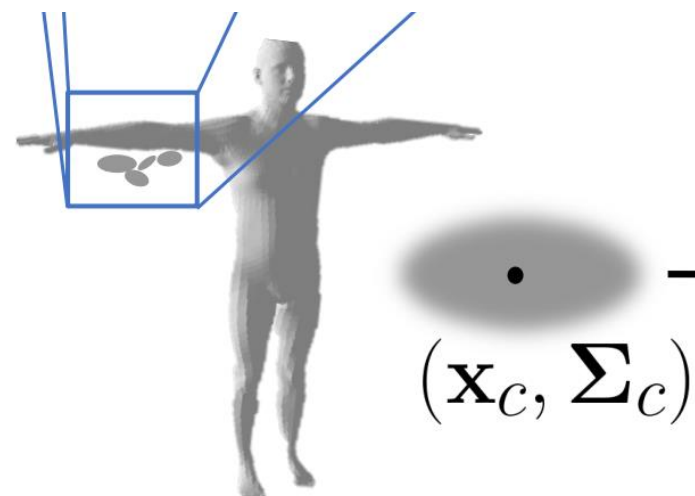
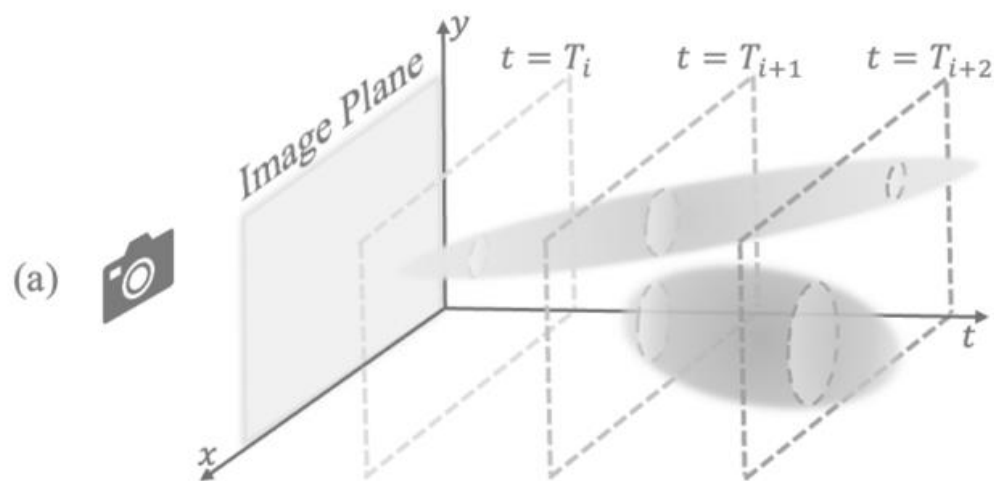
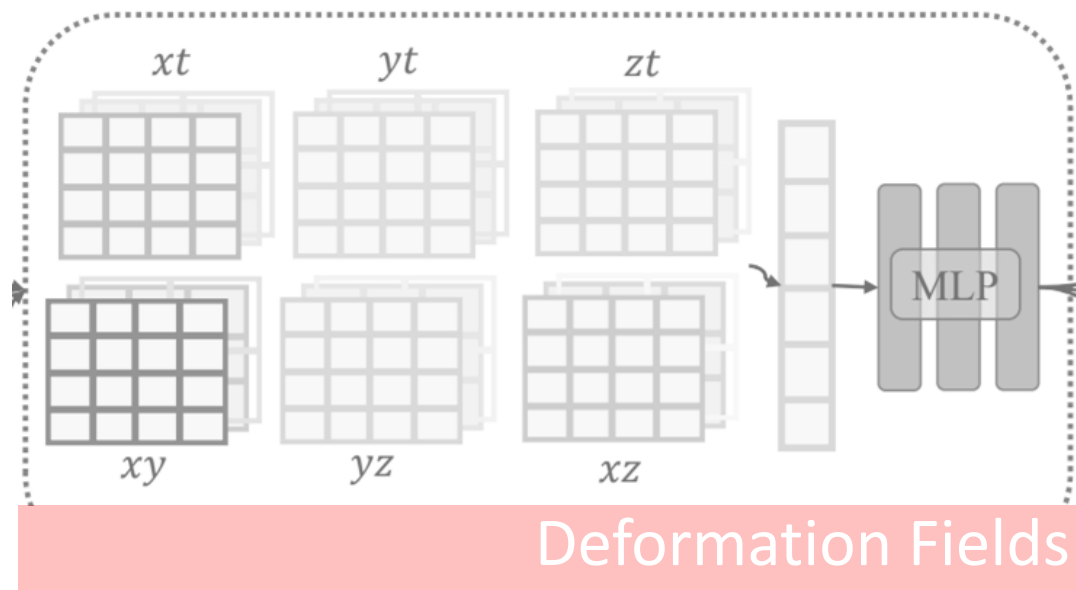
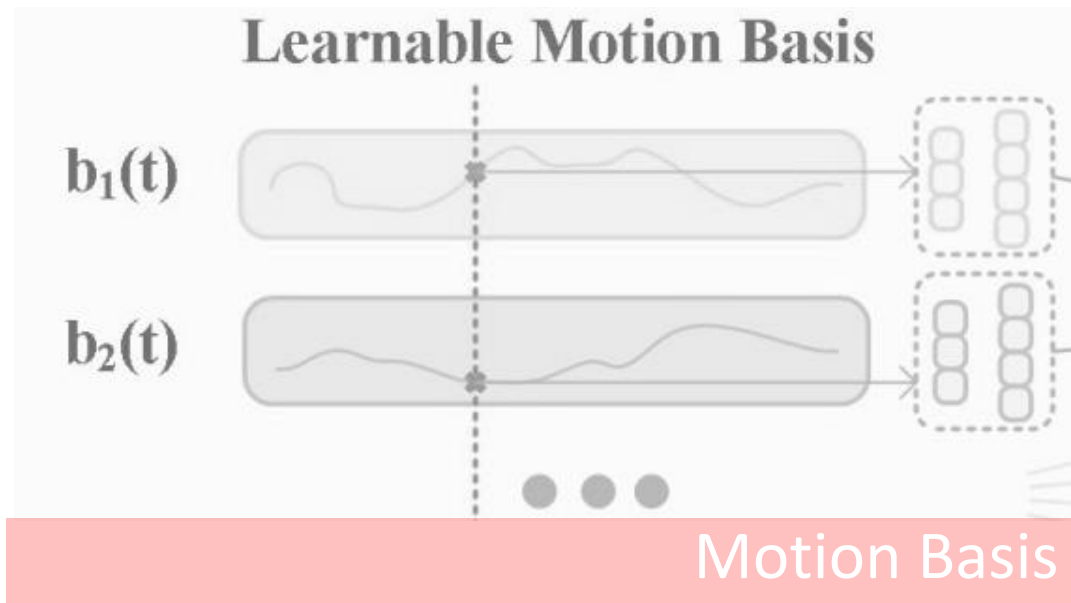
## 4D Gaussian Splatting: Towards Efficient Novel View Synthesis for Dynamic Scenes

Yuanxing Duan<sup>1\*</sup> Fangyin Wei<sup>2\*</sup> Qiyu Dai<sup>1,4</sup> Yuhang He<sup>1</sup> Wenzheng Chen<sup>3†</sup> Baoquan Chen<sup>1,4†</sup>  
<sup>1</sup>Peking University <sup>2</sup>Princeton University <sup>3</sup>NVIDIA <sup>4</sup>National Key Lab of General AI, China

Uses a different (rotor-based) 4D gaussian covariance representation.  
More naturally decomposes into 3D + 1D, also VERY fast CUDA impl.



# Alternative Dynamic Gaussian Representations



# Shape Templates

## Drivable 3D Gaussian Avatars

Wojciech Zielonka<sup>3,1\*</sup>, Timur Bagautdinov<sup>1</sup>, Shunsuke Saito<sup>1</sup>,  
Michael Zollhöfer<sup>1</sup>, Justus Thies<sup>2,3</sup>, Javier Romero<sup>1</sup>

<sup>1</sup>Meta Reality Labs Research    <sup>2</sup>Technical University of Darmstadt  
<sup>3</sup>Max Planck Institute for Intelligent Systems, Tübingen, Germany

<https://zielon.github.io/d3ga/>

## Animatable Gaussians: Learning Pose-dependent Gaussian Maps for High-fidelity Human Avatar Modeling

Zhe Li<sup>1</sup>, Zerong Zheng<sup>2</sup>, Lizhen Wang<sup>1</sup>, Yebin Liu<sup>1</sup>

<sup>1</sup> Department of Automation, Tsinghua University    <sup>2</sup> NNNKosmos Technology

<https://animatable-gaussians.github.io/>

## SplatArmor: Articulated Gaussian splatting for animatable humans from monocular RGB videos

Rohit Jena<sup>1\*</sup>    Ganesh Iyer<sup>2</sup>    Siddharth Choudhary<sup>2</sup>    Brandon M. Smith<sup>2</sup>

Pratik Chaudhari<sup>1</sup>    James C. Gee<sup>1</sup>  
<sup>1</sup>University of Pennsylvania    <sup>2</sup>Amazon.com, Inc

## GART: Gaussian Articulated Template Models

Jiahui Lei<sup>1</sup>    Yufu Wang<sup>1</sup>    Georgios Pavlakos<sup>2</sup>    Lingjie Liu<sup>1</sup>    Kostas Daniilidis<sup>1,3</sup>  
<sup>1</sup> University of Pennsylvania    <sup>2</sup> UC Berkeley    <sup>3</sup> Archimedes, Athena RC  
{leijh, yufu, lingjie.liu, kostas}@cis.upenn.edu, pavlakos@berkeley.edu

## Human Gaussian Splatting: Real-time Rendering of Animatable Avatars

Arthur Moreau\*    Jifei Song\*    Helisa Dharmo    Richard Shaw    Yiren Zhou  
Eduardo Pérez-Pellitero  
Huawei Noah's Ark Lab

## HUGS: Human Gaussian Splats

Muhammed Kocabas<sup>1</sup>    Jen-Hao Rick Chang<sup>2</sup>    James Gabriel<sup>2</sup>    Oncel Tuzel<sup>2</sup>    Anurag Ranjan<sup>2</sup>  
<sup>1</sup>Apple    <sup>2</sup>Max Planck Institute for Intelligent Systems    <sup>3</sup>ETH Zurich

## Gaussian Shell Maps for Efficient 3D Human Generation

Rameen Abdal\*<sup>1</sup>    Wang Yifan\*<sup>1</sup>    Zifan Shi\*<sup>1,2</sup>    Yinghao Xu<sup>1</sup>    Ryan Po<sup>1</sup>    Zhengfei Kuang<sup>1</sup>  
Qifeng Chen<sup>2</sup>    Dit-Yan Yeung<sup>2</sup>    Gordon Wetzstein<sup>1</sup>  
<sup>1</sup>Stanford University    <sup>2</sup>HKUST

# Shape Templates

## SplatArmor: Articulated Gaussian splatting for animatable humans from monocular RGB videos

Rohit Jena<sup>1\*</sup>

Ganesh Iyer<sup>2</sup>

Siddharth Choudhary<sup>2</sup>

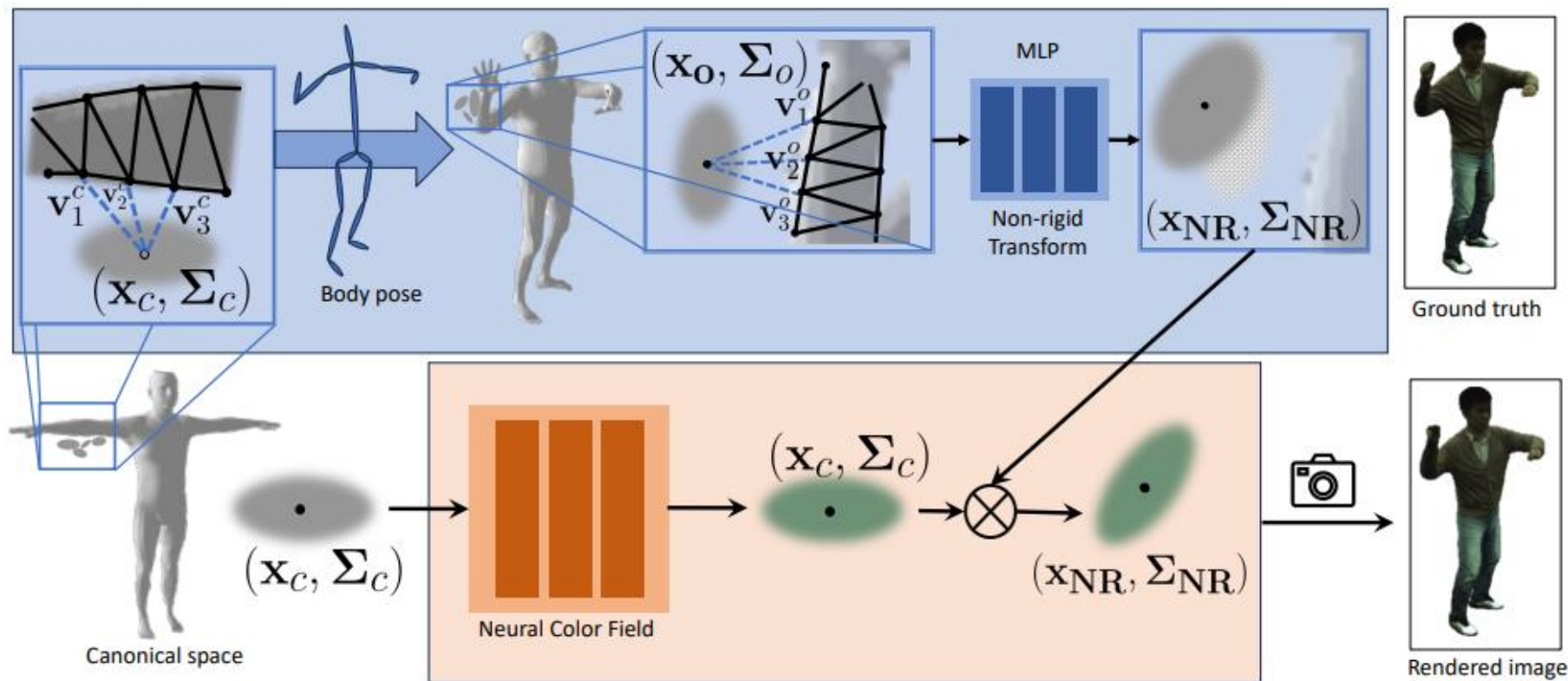
Brandon M. Smith<sup>2</sup>

Pratik Chaudhari<sup>1</sup>

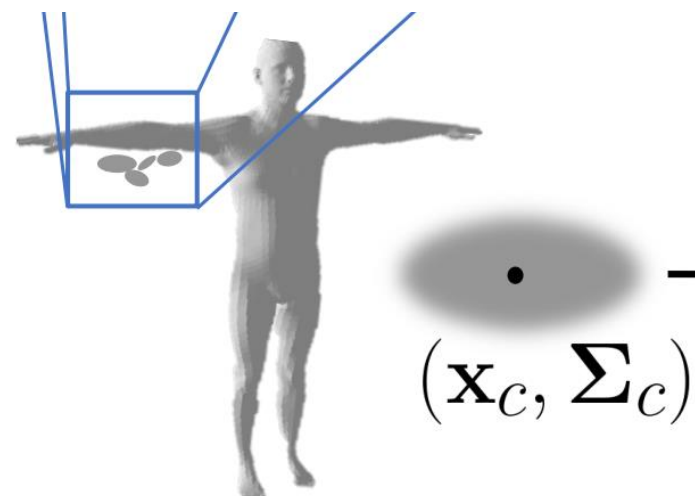
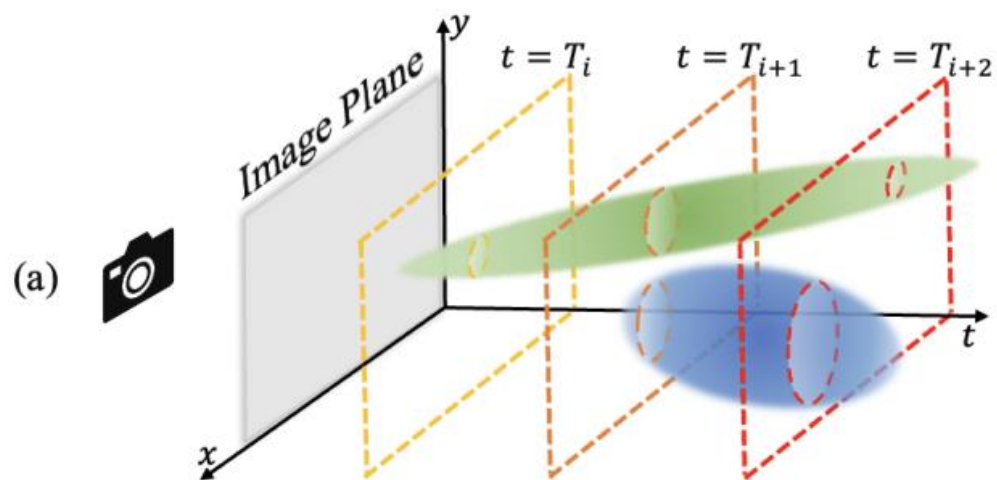
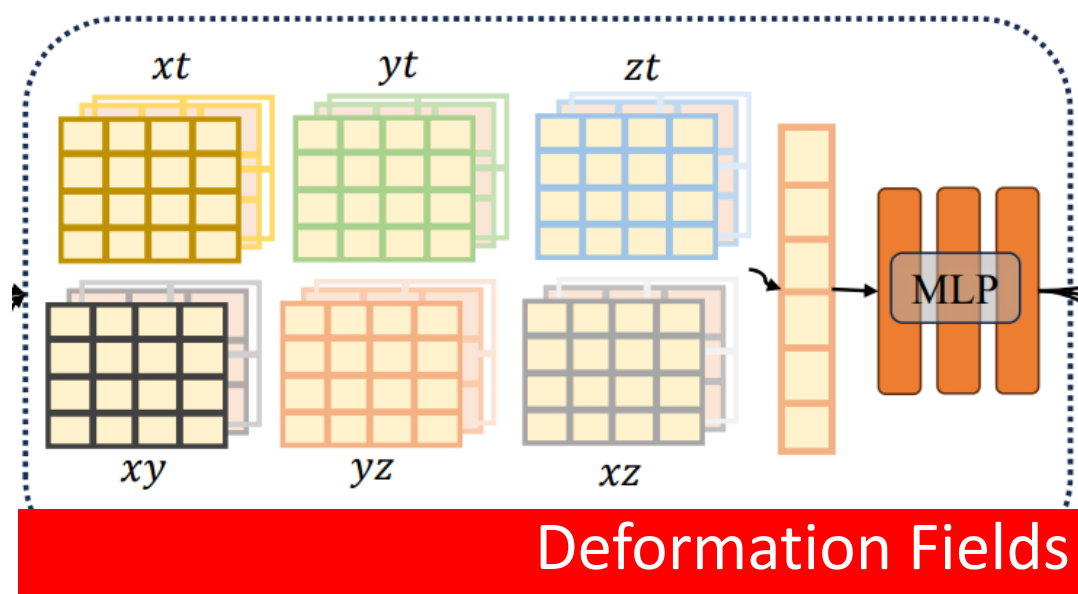
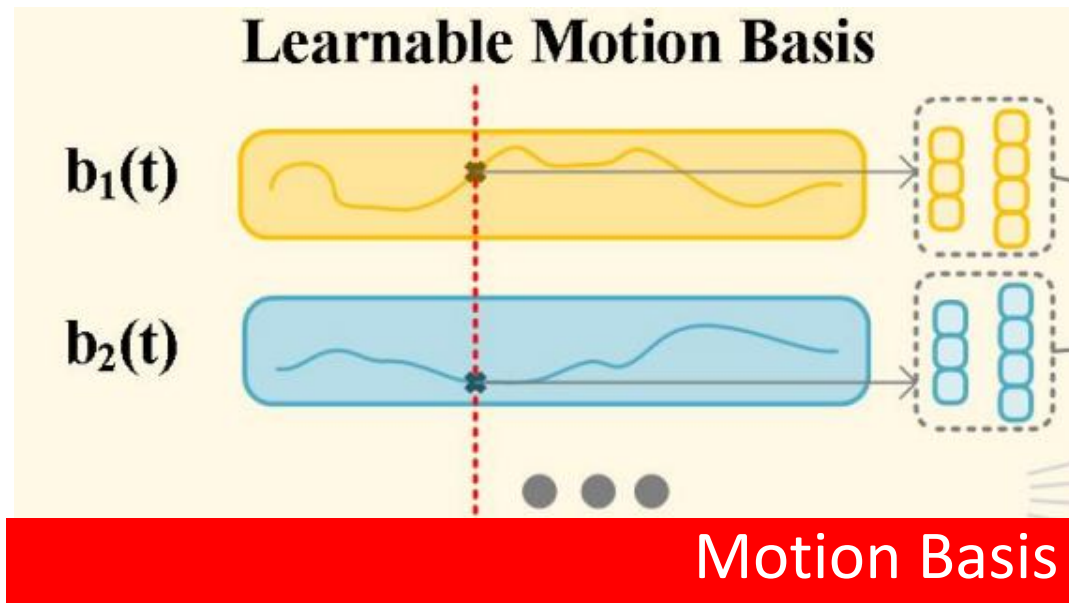
James C. Gee<sup>1</sup>

<sup>1</sup>University of Pennsylvania

<sup>2</sup>Amazon.com, Inc



# Alternative Dynamic Gaussian Representations



4D Gaussians

Shape Templates



# Thanks

