

# MaterialGAN: Reflectance Capture using a Generative SVBRDF Model



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## **Material acquisition**





## **Related works**





Dana and Wang 2004



#### Weyrich et al. 2006



#### Spherical Gantry

### High quality Need dense captures with complex devices

### **Related works**





Aittala et al. 2015



Hui et al. 2017



Li et al. 2017



Li et al. 2018





Deschaintre et al. 2019



Gao et al. 2019

One or more images captured by handphone Quality is not good as using specialized devices

## **SVBRDF**





Inputs



SVBRDF maps

## Challenges





**Under-constrained** 

## Challenges





□ User interaction

- □ Linear low-dimensional BRDF models
- **Prior :** Stationary stochastic textures
  - □ Learned with Neural Network
  - ...

## Challenges





**Unnatural SVBRDF maps** 

## **Our technique**



Need a good prior to make SVBRDF maps look more natural



## **GAN review**



GAN Goodfellow et al. 2014

Image



StyleGAN2 Karras et al. 2019 Video



MoCoGAN Tulyakov et al. 2018

3D shape



MC-GAN Li et al. 2019



### **Our method**

Please check our main paper and supplemental materials for detailed analyses and comparisons



### MaterialGAN = StyleGAN2 + SVBRDF



Z: latent vector of MaterialGAN

## **MaterialGAN**





### SVBRDF maps

Rendering

# **Training details of MaterialGAN**



- □ Treat SVBRDF maps as 9-channel "image"
- □ 100,000 training data (including augmentation) from Deschaintre et al. 2018
- □ Resolution 256x256
- □ Tensorflow
- □ 8× Nvidia Tesla V100, 5 days



























## **MaterialGAN latent space**





## **Final optimization pipeline**







### **Results**

Please check our main paper and supplemental materials for detailed analyses and comparisons

## **Implementation details**



- □ Single point source light, collocated with the camera
- □ Support single input to 25 inputs (GPU with 16GB memories)
- □ 39 synthetic testing data from Deschaintre et al. 2018 and Adobe Stocks dataset
- □ 39 cellphone captures for testing
- Pytorch
- □ Titan RTX, 2000 iterations takes about 2 minutes

## **Optimization results**





Inputs





Optimized maps

Renderings







Reference (Inputs)



Reference (Novel view)





**Estimated Maps** 

Ours (Optimized)



Ours (Novel view)



## Wood





### Reference (Inputs)



#### Reference (Novel view)





Ours (Optimized)



Ours (Novel view)

**Estimated Maps** 



### Leather





Ours (Optimized)

Ours (Novel view)









Reference (Inputs)



Reference (Novel view)







Ours (Optimized)



Ours (Novel view)



## Book





### Reference (Inputs)



Reference (Novel view)

.... Prince







Ours (Optimized)

Ours (Novel view)

Prince









## **Comparison with [Gao et al. 2019]**

Please check our main paper and supplemental materials for detailed analyses and comparisons



### **Ours is less sensitive to initialization**





#### Ours is not require post-refinement to get sharp maps



## **Latent Space Interpolation**









- □ Our model relies on simple BRDF model
- □ Flat surface only
- □ Relies on known illumination
- Dataset is not general enough

## Conclusion

SVBRDF acquisition from a small number of input captures with smartphone

An optimization framework with a powerful material prior (MaterialGAN)

High quality SVBRDF reconstruction without any good initialization









# Thank you!







#### □ Anonymous reviewers

□ TJ Rhodes from Adobe Research for help with material capture hardware setup

### □ NSF IIS-1813553



Project page



Github repo