

Deep Learning for Graphics

Data sets

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Timetable

	Niloy	lasonas	Paul	Vova	Kostas	Tobias
Introduction	Х	Х	Х			Х
Theory	Х					
NN Basics		Х				Х
Supervised Applications						
Data						Х
Unsupervised Applications			Х			
Beyond 2D	Х		Х	Х		
Outlook	Х	Х	Х	Х	Х	Х



Modalities

- 2D images
- 3D datasets
- Data augmentation
- Synthetic image training data



2D image data sets



- MNIST
 - Handwritten digits
 - 28x28 images
 - 10 classes
 - 60k train/10k test

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- CIFAR10
 - Object images
 - 10 classes
 - 32 x 32 pixels
 - 50k train/10k test

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- PASCAL VOC
 - Multiple objects per image
 - 20 classes
 - Labels for classification, segmentation, dete







- ImageNet
 - The main "fuel" for deep learning
 - 1000 classes
 - Classification/Detection (200 classes)
 - Structure from WordNet





- MS COCO
 - Boost to DL class/instance segmentation and keypoint detection
 - 80 classes
 - 200k images
 - Instance segmentation masks (>1 mil)
 - Human keypoints (250k)







LDR 2 HDR

http://hdrv.org/hdrcnn/material/testset/index.html





3D data sets



3D Datasets

- ShapeNet
 - Similar to ImageNet but for CAD models
 - 55 common categories
 - 10k+ models
 - ShapeNetCore
 - 12k models
 - Additional annotations (real world dimensions, materials,...)

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3D Datasets

• SMPL

- Parametric human shape model
- 72 parameters control pose and human shape
- Fully differentiable
- Useful for human shape estimation, motion capture etc







• Augment existing data with image operations to reduce overfitting

• Much larger dataset

• Learn expected transformations





• Mirror





- Mirror
- Rotation





- Mirror
- Rotation
- Translation





- Mirror
- Rotation
- Translation
- Zoom





- Mirror
- Rotation
- Translation
- Zoom
- Blur





- Mirror
- Rotation
- Translation
- Zoom
- Blur
- Noise





- Mirror
- Rotation
- Translation
- Zoom
- Blur
- Noise
- Color transforms



Library https://github.com/codebox/imagersaugmentorg for Graphics"

Creating Your Own



Material/Illumination





Material/Illumination





Decomposition





Synthetic data



Synthetic Data











Synthetic Data for DL

- 3D models + renderer = unlimited data
- Suitable for data hungry approaches such as deep networks
- Higher fidelity -> smaller discrepancy between synthetic and real



How To Generate Synthetic Data

- What you need
 - 3D models with task-specific annotation
 - Renderer
- Example: Indoor depth estimation (McCormac et al)

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3D Room with furniture



NVIDIA OptiX renderer





EG Course "Deep Learning for Graphics"

How To Generate Synthetic Data

• How much fidelity?







Path tracer, global Illumination (VRay)

OpenGL

Natural Illumination + AO



Selecting viewpoint

- Object in the center
 - Sample from the hemisphere
 - Multiple FOV, Target, Up, etc
- Scene
 - Simulate human camera path
 - Optimize camera position for a particular objective (eg segmentation)





Selecting Appearance

- Materials
 - Capture the variability of real world objects
 - BRDF, textures (MERL DB)
- Illumination
 - Capture the effect of environment
 - Increase realism
 - Laval indoor DB http://indoor.hdrdb.com/







Thank you!



http://geometry.cs.ucl.ac.uk/dl4g/

