

# Depth Prediction and RGBD Images for Recognition

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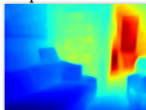
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May 25, 2016

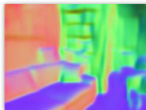
# Related work and motivation



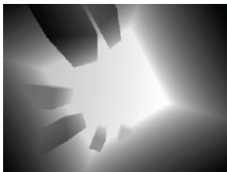
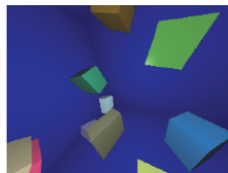
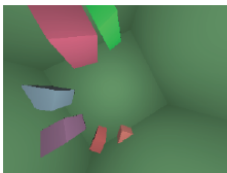
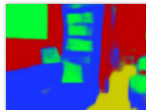
Depth



Normals



Labels



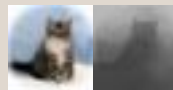
## our project: depth estimation & Classification on RGBD images

### implement previous work



### Go further

(2) Build a RGBD CIFAR10 based on indoor depth knowledge



(3) Compare RGBD and RGB  
 $label = f(RGBD)$   
 $label = f(RGB)$

first part: implement previous work

infer depth from RGB image

# Infer depth from RGB image: Loss definition

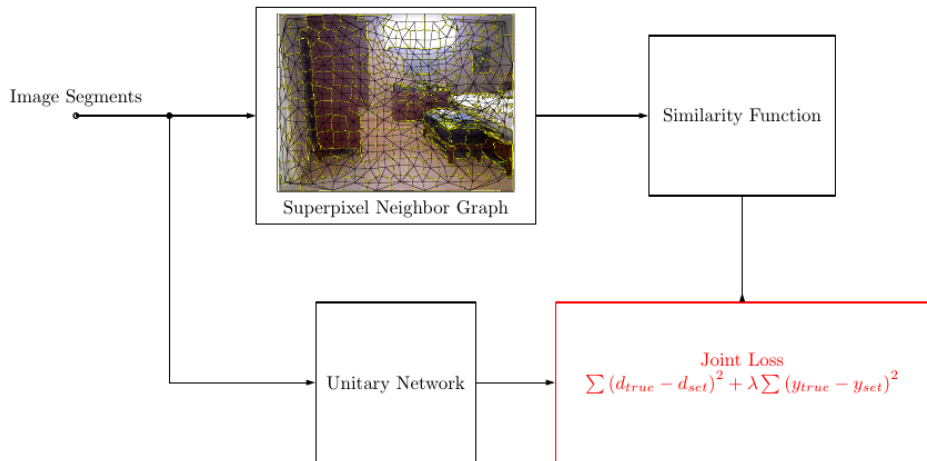
At training time, we combine two objective function<sup>1</sup>

- 1 regress to ground truth depth image(Kinect, PrimeSense)  
 $\sum_p (y_p - \hat{y}_p)^2$ , p stands for pixel.
- 2 Similarity between superpixels.  $R_{pq} = \sum_{k=1}^K \beta_k S_{pq}^{(k)}$   
 $\beta$  is trainable weight. S is similarity function.

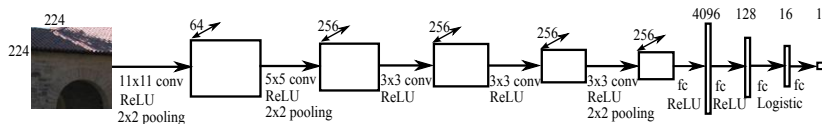
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<sup>1</sup>Fayao Liu, Chunhua Shen, and Guosheng Lin. "Deep convolutional neural fields for depth estimation from a single image". In: *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*. 2015, pp. 5162–5170.

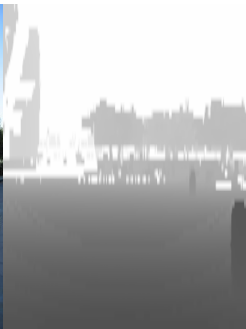
# Architecture: Deep convolutional Neural Field



# Infer depth from RGB image: Supervised part



using traditional CNN.





# Compare performance with original paper

Method	Error (lower is better)			Accuracy (higher is better)		
	rel	log10	rms	$\delta < 1.25$	$\delta < 1.25^2$	$\delta < 1.25^3$
Our implementation	<b>0.252</b>	0.103	0.860	0.544	<b>0.861</b>	0.943
Original paper	0.230	<b>0.095</b>	<b>0.824</b>	<b>0.614</b>	0.883	<b>0.971</b>

Table: Sanity check (**Bold** is better)

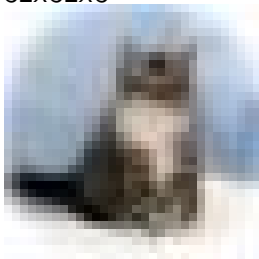
second part: go further

# Classification on RGBD images

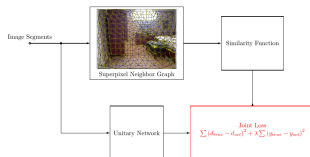
# build RGBD CIFAR dataset



32x32x3



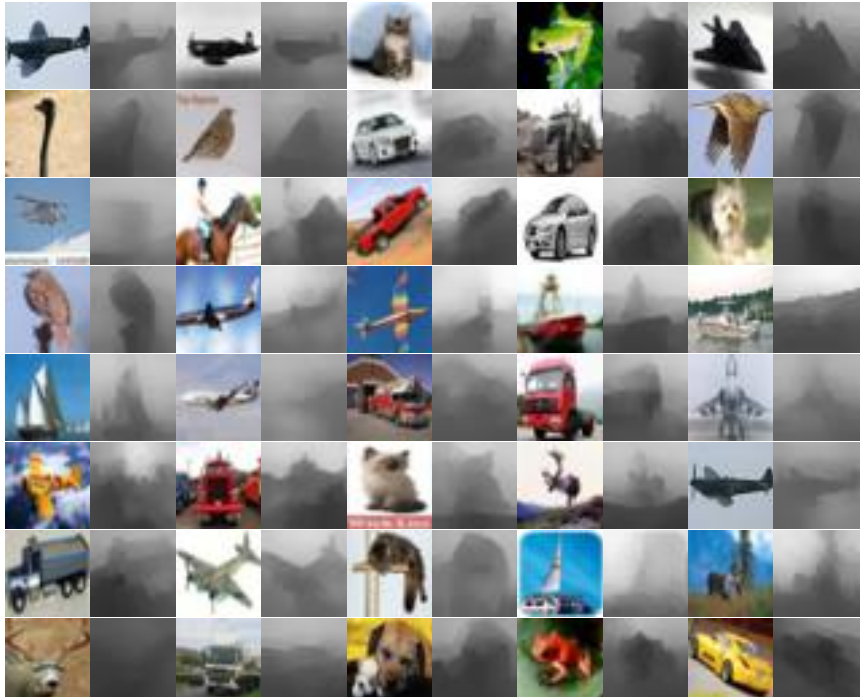
400x400x3



400x400x1



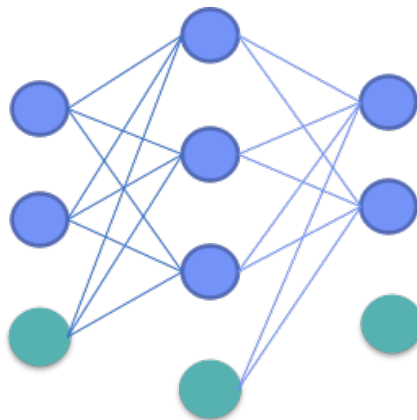
32x32x4



# architecture

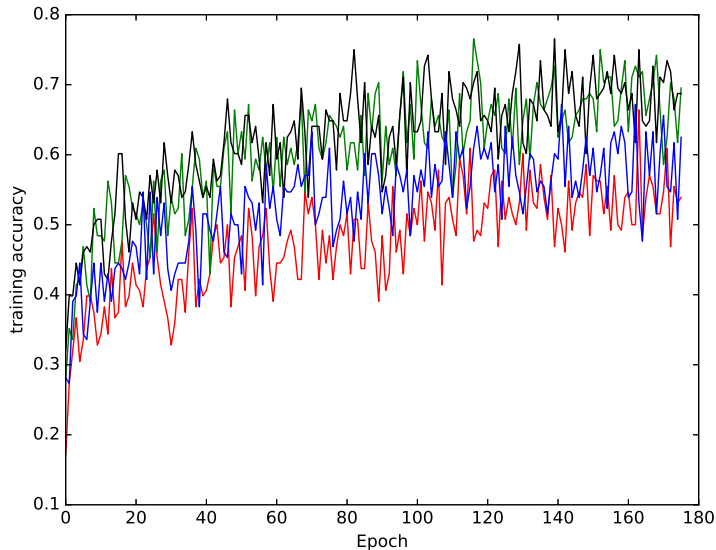


32x32x4

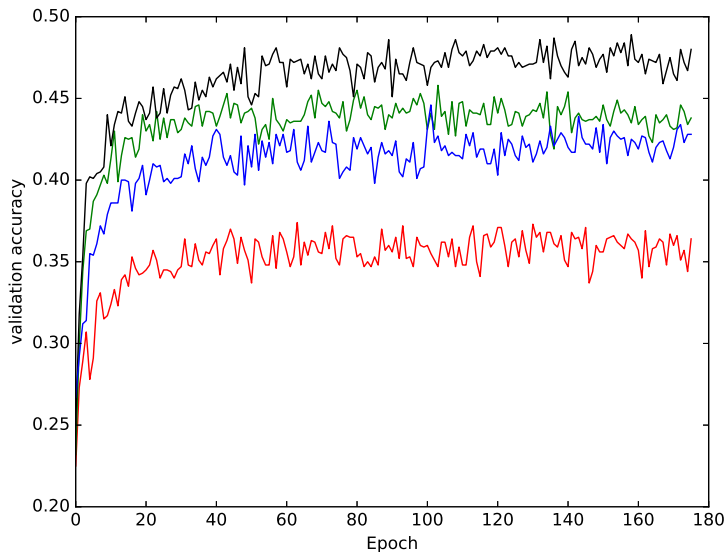


airplane  
automobile  
bird  
cat  
deer  
dog  
frog  
horse  
ship  
truck

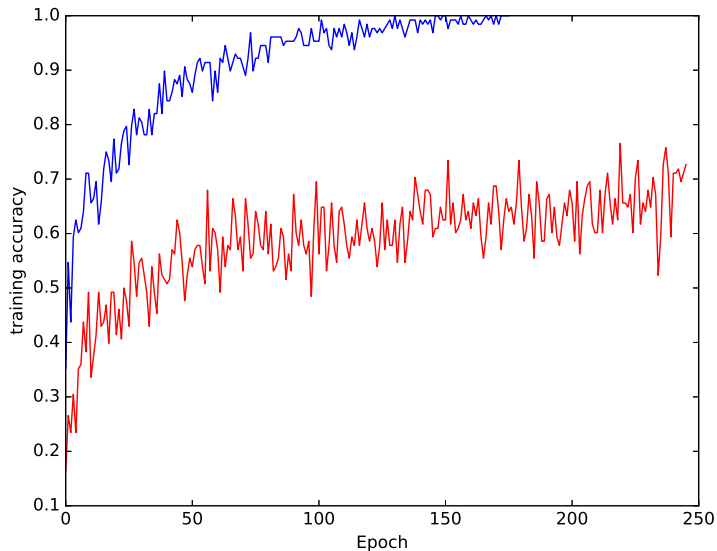
# R vs G vs B vs D: training time



# R vs G vs B vs D: testing time

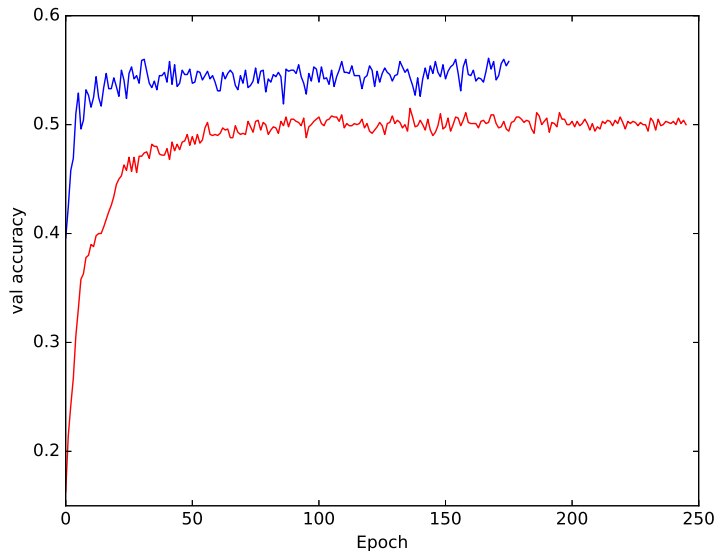


# RGBD vs RGB: training time

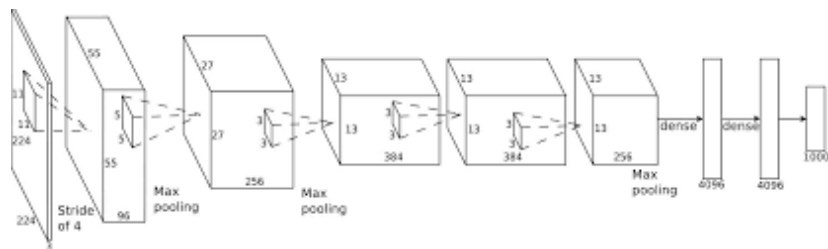


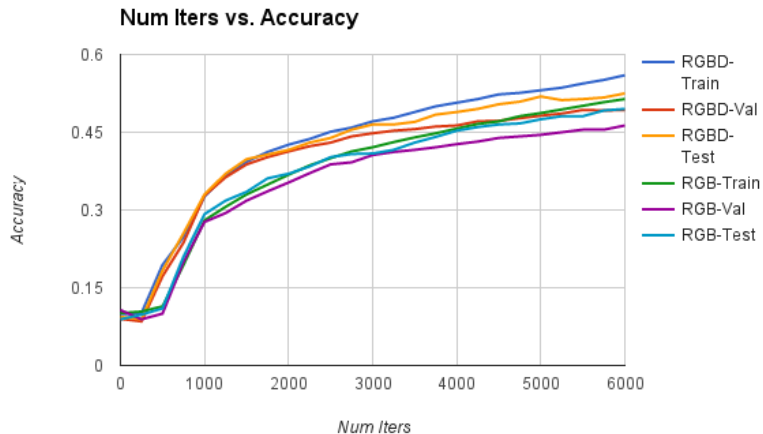


# RGBD vs RGB: testing time



# architecture





# our contribution

- 1 reproduce previous work on depth estimation
- 2 create the first RGBD CIFAR10 dataset
- 3 define a new metric for depth prediction problem
- 4 prove that depth channel has a better feature representation
- 5 show that training on RGBD images can somehow improve accuracy

# questions?<sup>2</sup>

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<sup>2</sup>code, references, report and slides can be access here:  
<https://github.com/yihui-he/Depth-estimation-with-neural-network>