

## Learning to predict stereo reliability enforcing local consistency of confidence maps Matteo Poggi, Stefano Mattoccia University of Bologna

## **Confidence measures and local consistency**

- In stereo matching, confidence measures provide a per-pixel estimation of the correctness of the assigned disparity
- Our proposal: using confidence prediction from the neighboring pixel to improve the effectiveness of the measure
- $\blacktriangleright$  Processing confidence maps (e.g., depicted by PKR measure), we obtain a new map (e.g., we call, in this case, PKR+)



## **Enforcing local consistency with deep learning**

Patch-based CNN trained to enforce local consistency for a single confidence measure



- Perceptive field: 9x9
- $\blacktriangleright$  Four convolutional layer (3x3), each extracting 128 feature maps
- $\blacktriangleright$  Two convolutional layers (1x1), each extracting 384 feature maps
- Final regression neuron
- $\succ$  Trained by Stocastic Gradient Descent (SGD), minimizing binary cross entropy as loss function (BCE)

**Building the training set** 

- $\succ$  Compute disparity maps and confidence maps, according to given stereo algorithm and confidence measure, on stereo images with available ground-truth (e.g., from KITTI or Middlebury datasets)
- $\succ$  For each pixel, compare the computed disparity with the available ground-truth
- $\succ$  Fixing a threshold T, label as correct (1) pixels having a disparity error lower than T, or wrong (0) if greater.
- $\succ$  Train the network on the obtained binary labels

## Impact of the amount of training data

- $\blacktriangleright$  We train on a subset of images from KITTI 2012 (20 images, T=3)
- $\succ$  Each pixel with available ground-truth represent a training sample
- $\blacktriangleright$  Training with 5, 15, 25 and 35 images (about 0.7, 1.5, 2.0, 2.7 and 3.5 million samples)
- $\succ$  Evaluation of the networks on Middlebury v3
- $\succ \Delta = (AUC AUC +)/(AUC AUCopt)$
- $\succ$  (Train on Middlebury v3.  $\rightarrow$  only 1.2% better)





Source code and trained networks http://vision.disi.unibo.it/~mpoggi/code.html

Acknowledgment: the Titan X Pascal used for this research was donated by the NVIDIA Corporation

## **Experimental results**

- **AD-CENSUS** (and **MC-CNN**, in the paper)



# Enseble [2], GCP [3], Park [4], O1 [5], CCNN [6]



Exploiting local consistency with our network **always improves the input confidence measure in term of AUC** from  $\Delta$ =75% to 9%

### References

- [1] Hu and Mordohai, PAMI 2012
- [2] Haeusler et al., CVPR 2013
- [3] Spyropoulos et al., CVPR 2014
- [4] Park and Yoon, CVPR 2015
- [5] Poggi and Mattoccia, 3DV 2016
- [6] Poggi and Mattoccia, BMVC 2016





# Training on KITTI 2012 (20 images, T=3), evaluation on the KITTI **2012**, **KITTI 2015** and **Middlebury v3** datasets. Stereo algorithm

Testing on 18 confidence measures from literature [1]



Testing on 5 state-of-the-art machine-learning techniques: