

Supplementary Material for Submission: “PatchMatch Stereo - Stereo Matching with Slanted Support Windows”

BMVC 2011 Submission # 289

The Global Algorithm

In Section 2.4 of the paper we have described how to build a cost volume using our slanted adaptive support weight windows that can be used by global stereo methods. We embed this cost volume into a global algorithm whose details are described here.

The global algorithm searches a disparity map D that assigns each pixel of both views to a discrete disparity and minimizes the energy defined as

$$E(D) = E_{data}(D) + \lambda E_{smooth}(D). \quad (1)$$

Here, λ is a parameter that balances the influence of data and smoothness terms. We take the data term from [9] which accomplishes symmetrical occlusion handling and is defined as

$$E_{data}(D) = \sum_{p \in \mathcal{I}} \begin{cases} c(p, d_p) : d_p = d_{p'} \\ \lambda_{occ} : d_p < d_{p'} \\ \infty : \text{otherwise} \end{cases} \quad (2)$$

where \mathcal{I} denotes all pixels of both views. We write p' to denote p 's matching point according to p 's disparity in D . The function $c(p, d_p)$ looks up the costs for matching pixel p at disparity d_p in our cost volume and the parameter λ_{occ} puts a penalty on occluded pixels. As a smoothness term, we use the second order term of [9] that puts a penalty on disparity curvature:

$$E_{smooth}(D) = \sum_{\langle p, q, r \rangle \in \mathcal{N}} \min(|d_p - 2d_q + d_r|, \tau_{smooth}) \quad (3)$$

Here, \mathcal{N} denotes the set of all 3×1 and 1×3 patches in left and right images. The parameter τ_{smooth} truncates the smoothness costs in order to allow for sharp jumps in disparity at depth discontinuities. This smoothness term is a natural choice for our algorithm, since it overcomes the bias towards fronto-parallel surfaces that competing terms suffer from (e.g., truncated linear model).

For optimization we use the α -expansion algorithm [10]. Note that the graph used to compute an optimal α -expansion can contain non-submodular edges. We use QPBO [11] to derive a potentially incomplete binary labelling of pixels where label 0 means that the old disparity is kept and label 1 means that disparity α is taken. Unlabelled pixels are set to label 0, i.e., we apply QPBOF.

References

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