Segmentation-Aware Convolutional Networks Using Local Attention Masks Supplementary Material



output

reshape

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Multi-dim. embeddings

im2dist

Unrolled masks

Figure 1: Implementation of convolution in Caffe, compared with the implementation of segmentation-aware convolution. Convolution involves re-organizing the elements of each (potentially overlapping) patch into a column (*i.e.*, *im2col*), followed by a matrix multiplication with weights. Segmentation-aware convolution works similarly, with an image-to-column transformation on the input, an image-to-distance transformation on the embeddings (*i.e.*, *im2dist*), a pointwise multiplication of those two matrices, and then a matrix multiplication with weights. The variables H, W denote the height and width of the input, respectively; E denotes the number of channels in the input; K denotes the dimensionality of a patch (*e.g.*, K = 9 in convolution with a 3×3 filter); F denotes the number of filters (and the dimensionality of the output). In both cases, an $H \times W \times E$ input is transformed into an $H \times W \times F$ output.

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