

# Supplementary Material: Semantics-Adding Flaw-Erasing Network for Semantic Human Matting

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In this supplemental, we show some equations used in FEM. Moreover, we also provide the tables about our human matting dataset. Finally, we provide more visual results.

## 1 FEM

The detail of the inferior pixels indicated by  $\mathcal{FM}$  can be enhanced via the following operations:

$$\begin{aligned}
 i^t &= \sigma(W_{xi} * \tilde{F}_{in,c,s,d} + W_{hi} * h^{t-1} + W_{ci} \circ c^{t-1} + b_i), \\
 f^t &= \sigma(W_{xf} * \tilde{F}_{in,c,s,d} + W_{hf} * h^{t-1} + W_{cf} \circ c^{t-1} + b_f), \\
 c^t &= f^t \circ c^{t-1} + i^t \circ \tanh(W_{xc} * \tilde{F}_{in,c,s,d} + W_{hc} * h^{t-1} + b_c), \\
 o^t &= \sigma(W_{xo} * \tilde{F}_{in,c,s,d} + W_{ho} * h^{t-1} + W_{co} \circ c^{t-1} + b_o), \\
 h^t &= o^t \circ \tanh(c^t),
 \end{aligned} \tag{1}$$

where  $i^t$ ,  $f^t$  and  $o^t$  stand for the input, forget and output gates, respectively, and  $c^t$  stores the earlier information.  $\circ$  denotes the Hadamard product.  $h^0$  and  $c^0$  are initialized to 0. At the last step  $N$ , refined features  $F_{refine}^N = h^N$  are generated with the memory of the previous refined features. The final alpha matte is generated by  $\alpha = \sigma(W_*(F_{refine}^N))$ , where we use convolutional layers to reduce the dimension. We set  $N = 3$  in practice.

## 2 Human Matting Dataset

Table 1 compares our dataset with the existing human matting datasets:

Table 1: Configurations of different human matting datasets.

Dataset	Publicly available	Train Set		Test Set	
		Human	Image	Human	Image
Shen et al. [24]	✓	1700	1700	300	300
Trimap+DIM [25]	✓	202	20200	11	220
Distinctions646 [26]	✓	362	36200	11	220
SHM [27]	×	34493	34493	1020	1020
BSHM [28]	×	9324	93240	125	1250
BSHM (Coarse)	×	+10597	+105970	-	-
Ours	✓	4494	44940	235	2350

Table 2 shows details of the three subsets of our dataset:

Table 2: Foreground configurations of the proposed dataset.

Sub-dataset	Train Set	Test Set	Total
Single Frontal Subset	2177	106	2283
Single Pose-varied Subset	1901	110	2011
Multiple Subset	416	19	435
<b>Total</b>	4494	235	4729

## 3 More results

Fig. 1, we provide more visual results:

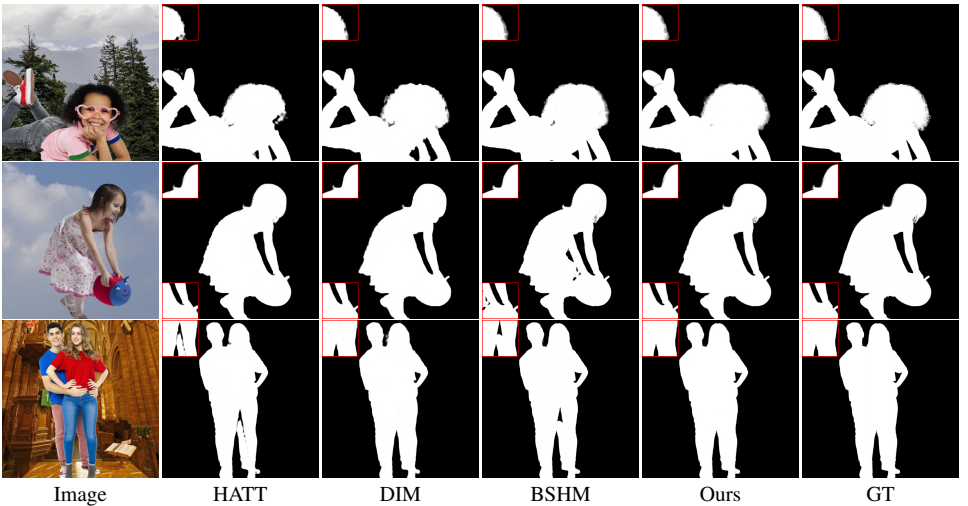


Figure 1: Qualitative comparison on the proposed dataset.

## References

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