

# Supplementary material of “Geometrized Transformer for Self-Supervised Homography Estimation”

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<https://github.com/ruc-aimc-lab/GeoFormer>

In this supplementary material, we report additional experimental results which are not included in the main paper due to space limits.

Table 7 shows the performance of varied methods on HPatches and ISC-HE. Table 8 shows the HPatches results divided between “Viewpoint” and “Illumination” splits. Visual localization results on the InLoc benchmark [15] are

given in Table 9.

## References

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Table 7: **Performance of different methods for homography estimation.** Per method, we use the star symbol (\*) to indicate its official release trained on the author-preferred dataset, while the no-star counterpart is re-trained by us on Oxford-Paris, using author-provided code.

Method	Testset: HPatches				Testset: ISC-HE			
	@3px	@5px	@10px	mAUC	@3px	@5px	@10px	mAUC
<i>Deep Homography:</i>								
CA-Unsupervised* [8], trained on custom dataset	20.5	31.7	40.1	30.8	8.9	16.4	23.4	16.2
CA-Unsupervised	37.6	41.9	45.3	41.6	10.7	17.5	23.8	17.3
HomoGAN* [5], trained on the same dataset as [8]	34.2	38.3	42.1	38.2	9.7	15.9	22.4	16.0
HomoGAN	31.3	36.8	41.8	36.6	9.1	15.6	22.3	15.7
BasesHomo* [16], trained on the same dataset as [8]	38.3	42.4	45.5	42.1	9.9	17.6	25.0	17.5
BasesHomo	39.6	43.2	45.9	42.9	10.1	17.1	23.9	17.0
<i>Detector-based matching:</i>								
SuperPoint* [2], trained on MS-COCO 2014 [7]	43.4	57.6	72.7	57.9	18.3	39.0	62.2	39.8
SuperPoint	42.1	56.2	69.9	56.1	16.2	37.8	61.4	38.5
R2D2* [9], trained on Aachen Day-Night dataset [13, 12]	50.6	63.9	76.8	63.8	18.2	39.6	62.9	40.2
R2D2	43.4	56.6	71.7	57.2	18.7	40.0	61.9	40.2
SuperGlue* [11], trained on MegaDepth [6]	53.9	68.3	81.7	68.0	19.6	42.2	66.9	42.9
SuperGlue	45.2	60.1	76.1	60.5	19.2	41.7	65.8	42.2
<i>Detector-free matching:</i>								
NCNet* [10], trained on PF-Pascal dataset [4]	48.3	50.1	59.8	52.7	9.6	25.3	51.2	28.7
NCNet	43.2	48.3	56.9	49.5	9.5	25.7	49.3	28.2
LoFTR* [14], trained on MegaDepth	65.3	75.2	84.6	75.0	18.6	38.8	60.5	39.3
LoFTR	58.5	69.8	81.1	69.8	18.7	41.0	64.8	41.5
ASpanFormer* [1], trained on MegaDepth	67.0	76.9	85.6	76.5	19.3	41.1	63.8	41.4
ASpanFormer	59.9	71.1	81.6	70.9	18.0	39.2	62.0	39.7
DKM* [3], trained on MegaDepth	70.7	<b>80.2</b>	<b>88.4</b>	79.8	19.1	40.4	63.4	41.0
DKM [3]	30.6	37.3	44.5	38.1	7.1	15.3	25.6	16.0
GeoFormer, trained on MegaDepth	<b>72.1</b>	80.0	87.7	<b>79.9</b>	<b>20.5</b>	<b>44.5</b>	<b>69.0</b>	<b>44.7</b>
GeoFormer	68.0	76.8	85.4	76.7	19.9	43.8	68.4	44.0

Table 8: **Split-view of HPatches results.**

	Illumination			Viewpoint		
	3px	5px	10px	3px	5px	10px
Training data: <i>Oxford-Paris</i>						
LoFTR	77.0	85.6	92.6	40.7	54.7	70.2
ASpanFormer	74.4	84.1	91.8	46.0	58.6	71.8
DKM	31.3	36.7	41.6	30.7	38.5	46.6
GeoFormer	<b>86.6</b>	<b>91.6</b>	<b>95.6</b>	<b>49.3</b>	<b>62.4</b>	<b>75.9</b>
Training data: <i>MegaDepth</i>						
LoFTR	79.6	87.2	93.4	51.7	63.7	76.3
ASpanFormer	79.5	87.2	93.4	54.9	67.0	78.1
DKM	79.9	87.5	93.6	<b>62.1</b>	<b>73.4</b>	<b>83.6</b>
GeoFormer	<b>89.1</b>	<b>93.2</b>	<b>96.4</b>	55.7	67.4	79.3

Table 9: **Performance of visual localization on InLoc.**  
Training data: MegaDepth.

Method	DUC1	DUC2
	(0.25m, 10°) / (0.5m, 10°) / (1m, 10°)	
LoFTR	46.5 / 70.2 / 81.3	48.9 / 67.9 / 80.9
GeoFormer	40.9 / 65.7 / 75.8	45.0 / 62.6 / 77.1

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