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# V-SENSE

## A Spatio-Angular Binary Descriptor for Fast Light Field Inter View Matching

Dr Martin Alain ([alainm@tcd.ie](mailto:alainm@tcd.ie)), Prof Aljosa Smolic  
IEEE International Conference on Image Processing 2020  
Session 3D-02 Light-Field Image Processing and Compression

# Outline

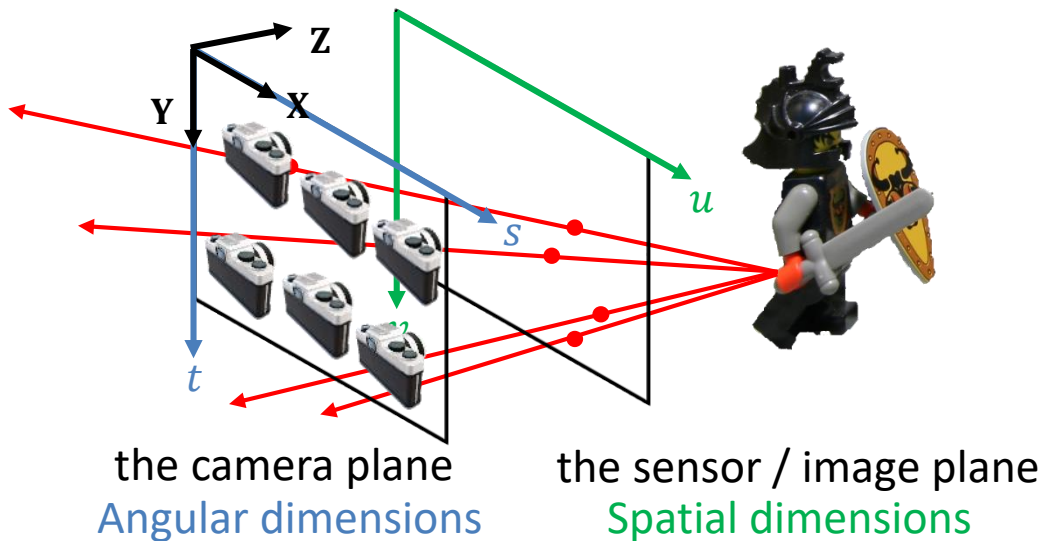
- **Introduction**
- **Proposed binary descriptor for light fields**
- **Experiments and results**
- **Conclusion**

# Outline

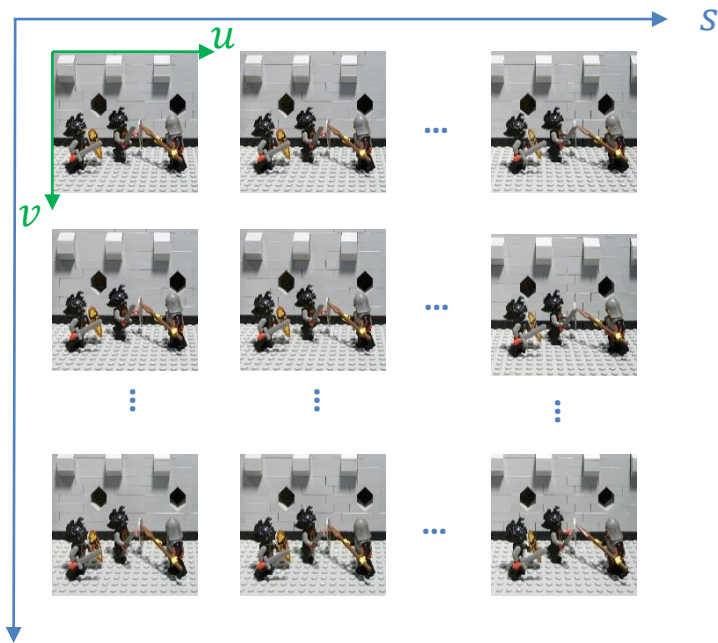
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# Light field imaging

2 parallel planes parameterization

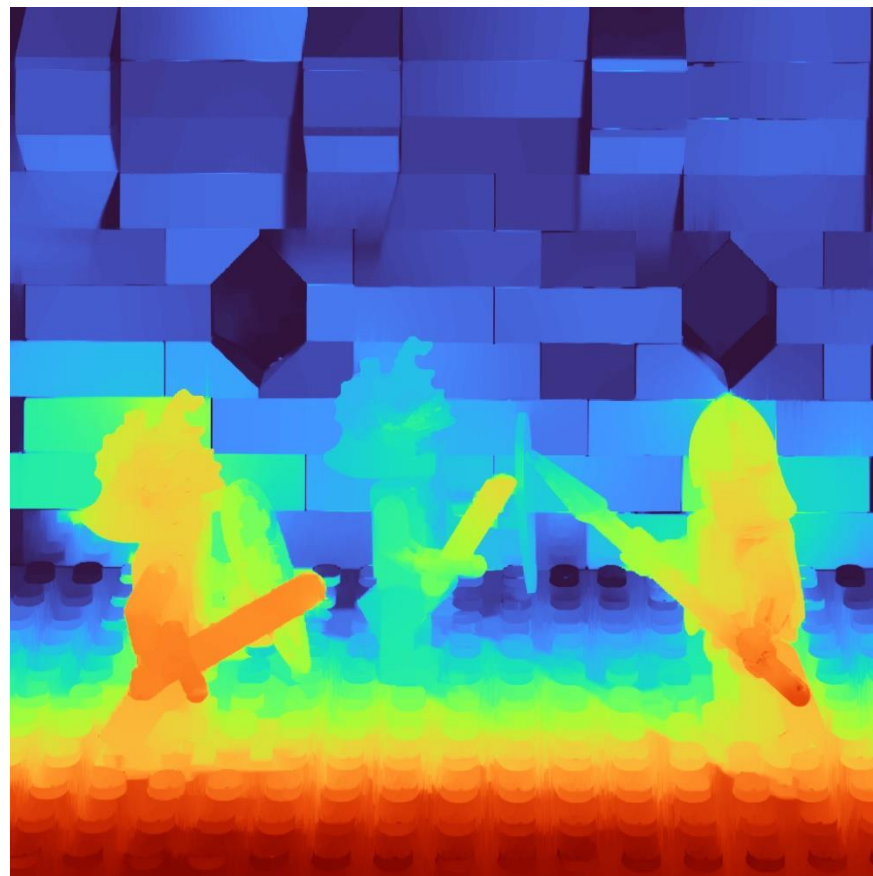
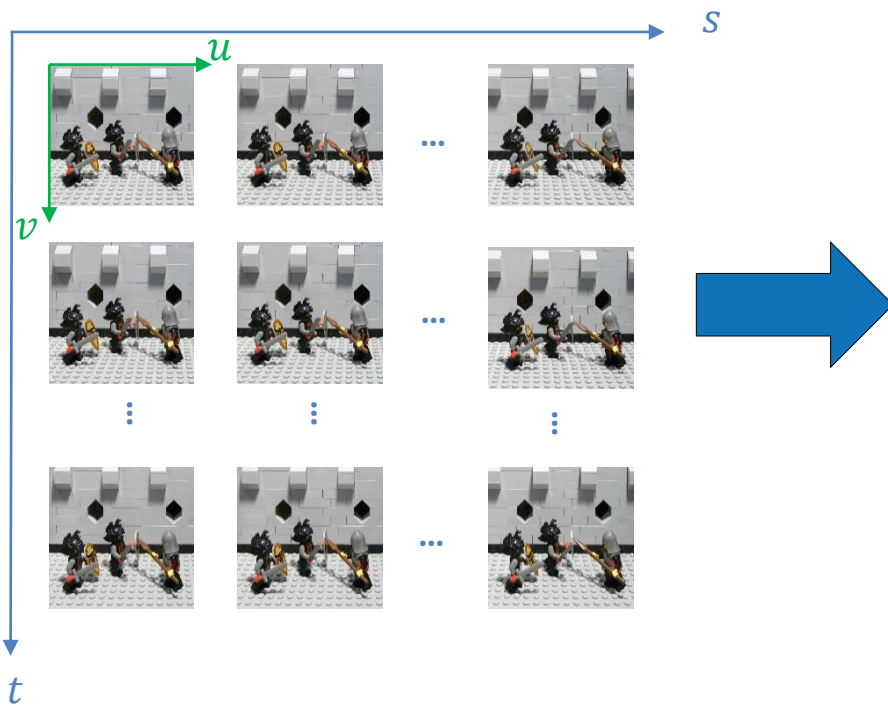


The light field  $L = L(s, t, u, v)$



# Light field applications

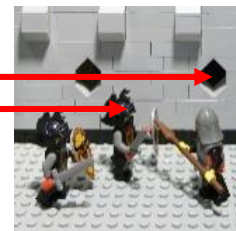
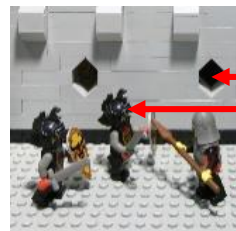
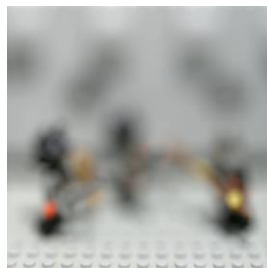
Disparity / Depth estimation



# Light field applications

Disparity / Depth estimation

- **EPI line slope**
- **Defocus cues**
- **Inter-view matching**
- **+ Learning-based**

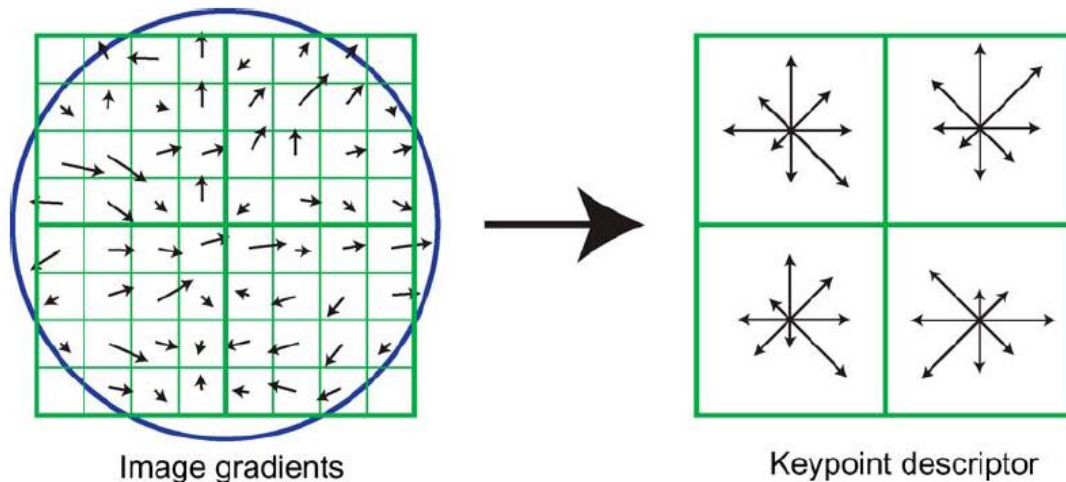


Chen, Yang, Martin Alain, and Aljosa Smolic. "Fast and accurate optical flow based depth map estimation from light fields." *Irish Machine Vision and Image Processing Conference (IMVIP)*. 2017

# Descriptors

## 2D pixel descriptor

- **SIFT [1]**
- **DAISY [2]**
- **BOOM [3]**



[1] Lowe, David G. "Object recognition from local scale-invariant features." *Proceedings of the seventh IEEE international conference on computer vision*. Vol. 2. Ieee, 1999.

[2] Tola, Engin, Vincent Lepetit, and Pascal Fua. "Daisy: An efficient dense descriptor applied to wide-baseline stereo." *IEEE transactions on pattern analysis and machine intelligence* 32.5 (2009): 815-830.

[3] Schaffner, Michael, et al. "Towards edge-aware spatio-temporal filtering in real-time." *IEEE Transactions on Image Processing* 27.1 (2017): 265-280.

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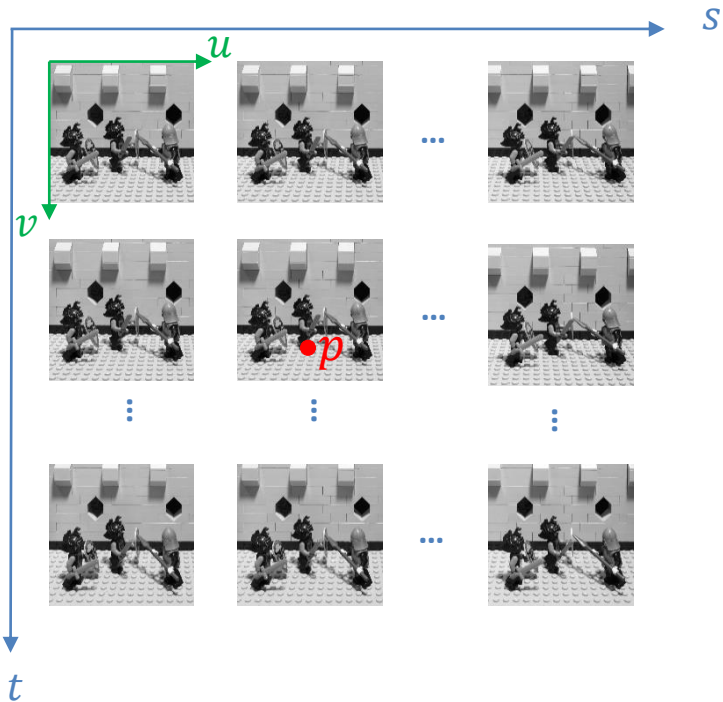
# SABOM

Spatio-Angular Binarised Orientation Maps

- **Extension of BOOM for light fields**
- **Exploits the 4D light field gradient**
- **Application to fast disparity map estimation**

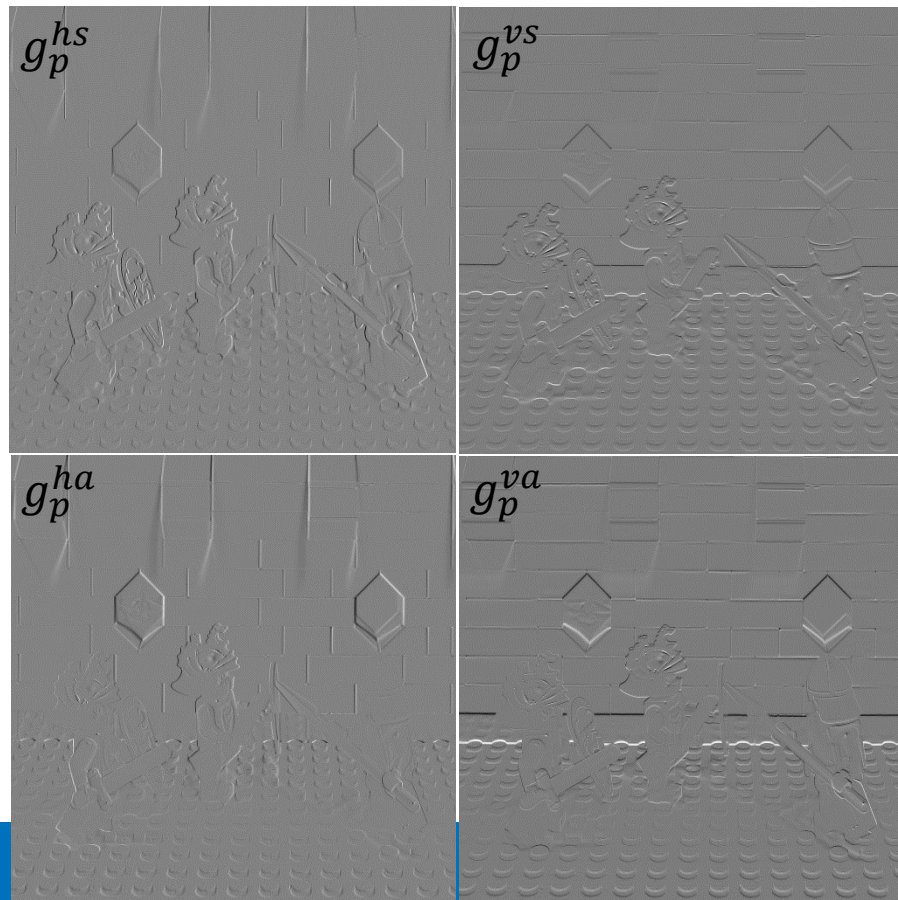
# SABOM

4D Gradient computation




$$\mathbf{g}_p = \begin{bmatrix} \mathbf{g}_p^s & \mathbf{g}_p^a \end{bmatrix} = \begin{bmatrix} g_p^{hs} & g_p^{vs} & g_p^{ha} & g_p^{va} \end{bmatrix}$$

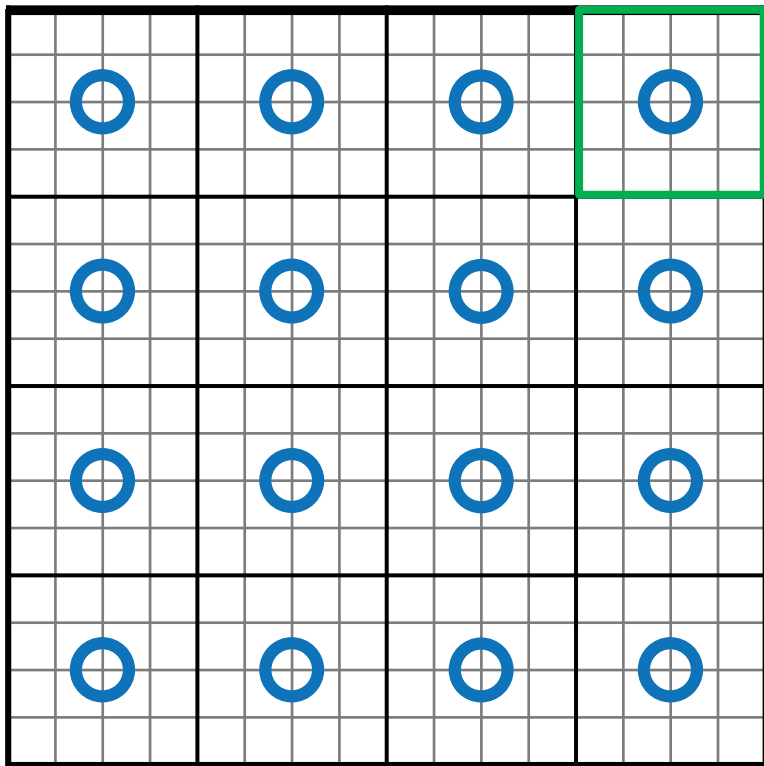
4D                      2D                      Scalar



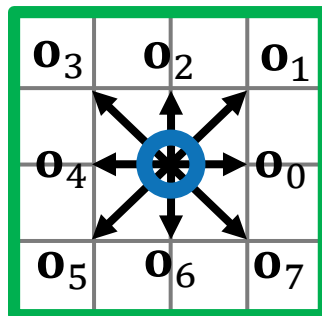
# SABOM

16 × 16 Input patch  $\mathbf{P}$

 16 spatial bin centers



8 spatial orientations



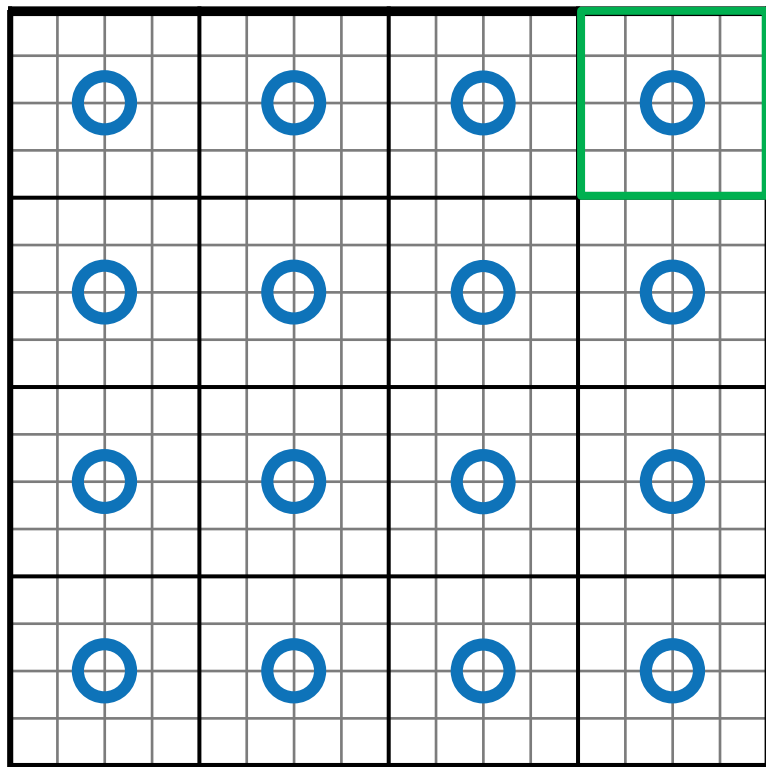
4 × 4 Pixel block  $\mathcal{B}^s$


$$r_{kl}^s = \sum_{p \in \mathcal{B}_l^s} \max(0, \langle \mathbf{o}_k | \mathbf{g}_p^s \rangle)$$

$k = 0 \dots 7, l = 0 \dots 15$

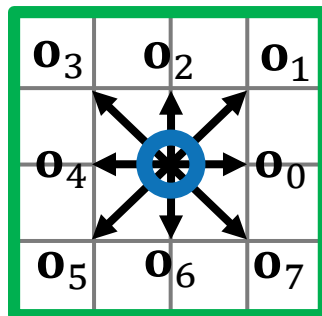
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Naive binarisation:

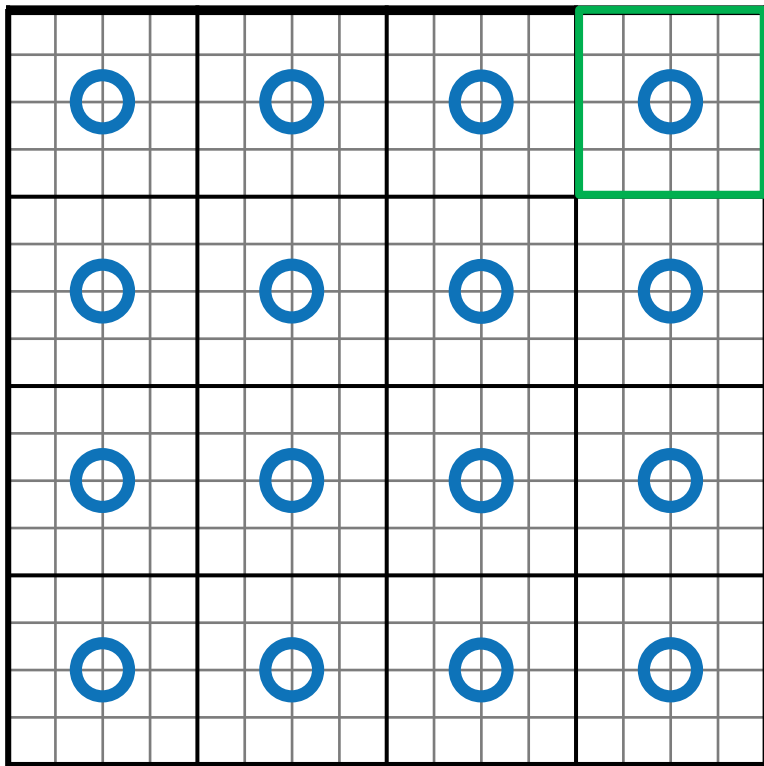
~~$$r_{kl}^s = \frac{r_{kl}^s}{\sum_{p \in \mathcal{B}_l^s} \|\mathbf{g}_p^s\|_2}$$

$$s^s = \sum_{p \in \mathcal{P}} \|\mathbf{g}_p^s\|_2$$~~

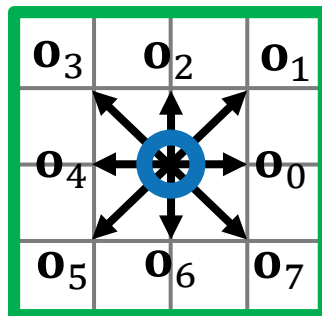
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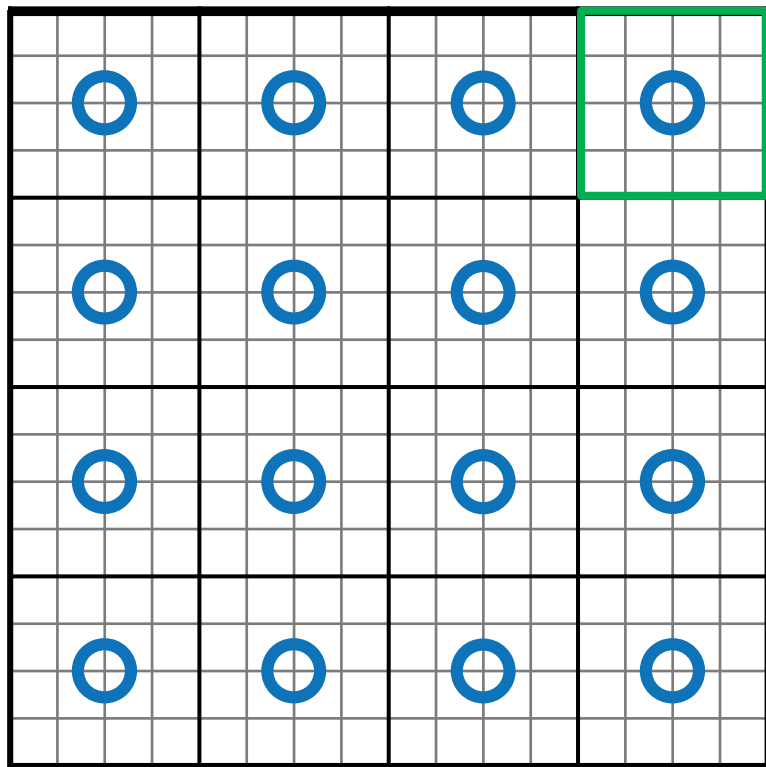
$k = 0 \dots 7, l = 0 \dots 15$


$$s^s = \sum_{p \in \mathcal{P}} \alpha \cdot \|\mathbf{g}_p^s\|_\infty + \beta \cdot \|\mathbf{g}_p^s\|_1$$

Seol, Changkyu, and Kyungwhoon Cheun. "A low complexity Euclidean norm approximation." *IEEE Transactions on Signal Processing* 56.4 (2008): 1721-1726.

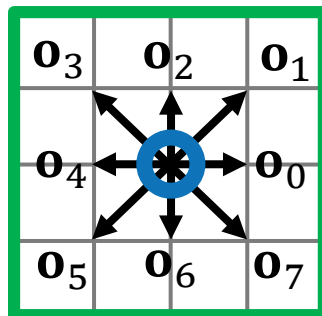
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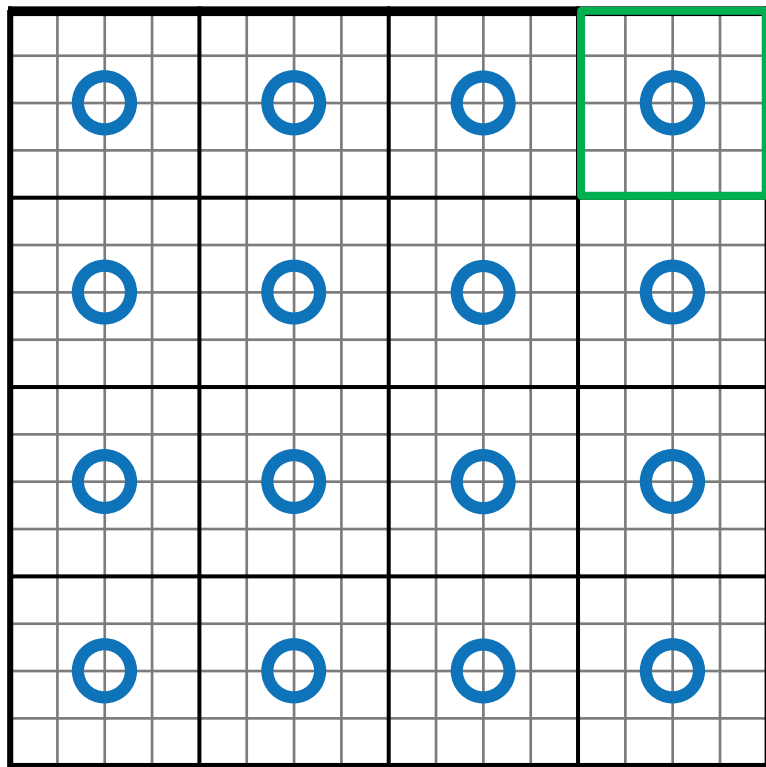
$k = 0 \dots 7, l = 0 \dots 15$


$$s^s = \sum_{p \in \mathcal{P}} \alpha \cdot \|\mathbf{g}_p^s\|_\infty + \beta \cdot \|\mathbf{g}_p^s\|_1$$

$$b_{kl}^s = r_{kl}^s \cdot \theta > s^s, \theta = \frac{1}{\varepsilon}$$

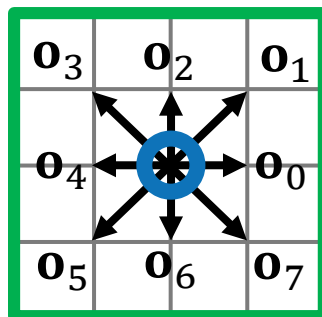
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16 × 16 Input patch  $\mathbf{P}$



 16 spatial bin centers

8 spatial orientations



4 × 4 Pixel block  $\mathcal{B}^s$

$$r_{kl}^s = \sum_{p \in \mathcal{B}_l^s} \max(0, \langle \mathbf{o}_k | \mathbf{g}_p^s \rangle)$$

$k = 0 \dots 7, l = 0 \dots 15$

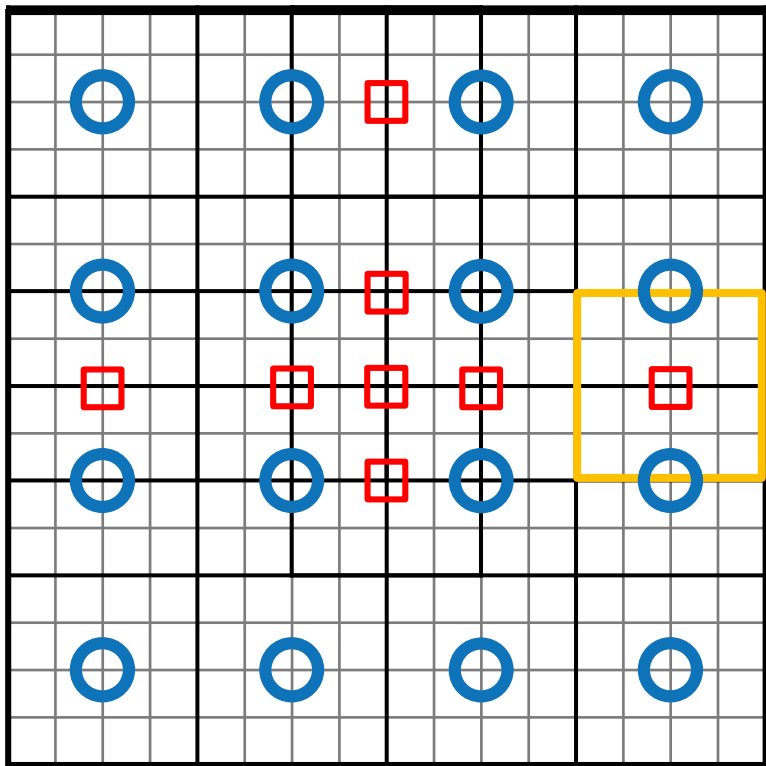
$$s^s = \sum_{p \in \mathcal{P}} \alpha \cdot \|\mathbf{g}_p^s\|_{\infty} + \beta \cdot \|\mathbf{g}_p^s\|_1$$

$$b_{kl}^s = \begin{cases} r_{kl}^s \cdot \theta_0^s > s^s, & \text{if } k \text{ is even} \\ r_{kl}^s \cdot \theta_1^s > s^s, & \text{if } k \text{ is odd} \end{cases}$$

16 × 8 = 128 bits

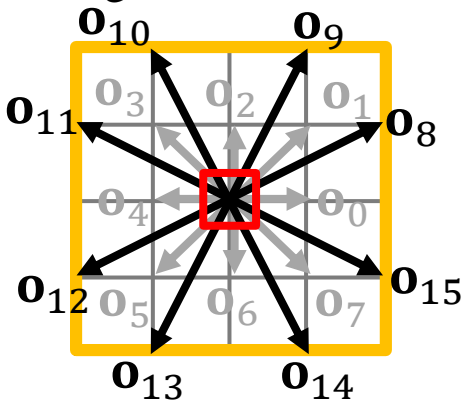
# SABOM

16 × 16 Input patch  $\mathbf{P}$



8 angular bin centers

16 angular orientations



4 × 4 Pixel block  $\mathcal{B}^a$

$$r_{kl}^a = \sum_{p \in \mathcal{B}_l^a} \max(0, \langle \mathbf{o}_k | \mathbf{g}_p^a \rangle)$$

$k = 0 \dots 15, l = 0 \dots 7$

$$s^a = \sum_{p \in \mathcal{P}} \alpha \cdot \|\mathbf{g}_p^a\|_\infty + \beta \cdot \|\mathbf{g}_p^a\|_1$$

$$b_{kl}^a = \begin{cases} r_{kl}^a \cdot \theta_0^a > s^a, & \text{if } k \text{ is even and } k < 8 \\ r_{kl}^a \cdot \theta_1^a > s^a, & \text{if } k \text{ is odd and } k < 8 \\ r_{kl}^a \cdot \theta_2^a > s^a, & \text{if } k \geq 8 \end{cases}$$

8 × 16 = 128 bits



# SABOM

## Implementation details

- **C++ implementation based on OpenCV ([github.com/V-Sense](https://github.com/V-Sense))**
- **CPU parallelisation over light views with OpenMP**
- **Use of integer arithmetic, no division**
- **Efficient dense scan implementation, re-using gradient response computed for neighbours, gradient is precomputed for the whole image**
- **Distance between descriptor computed using `__popcount64` instruction**
- **Binarisation implemented with bitshift**
- **$\alpha = 6, \beta = 2$**
- **$\theta_0^s = 512, \theta_1^s = 256, \theta_0^a = \theta_1^a = \theta_2^a = 512$**

# SABOM

Application to disparity map estimation: LF CPM + PF

Schaffner, Michael, et al. "Towards edge-aware spatio-temporal filtering in real-time." *IEEE Transactions on Image Processing* 27.1 (2017): 265-280.

Chen, Yang, Martin Alain, and Aljosa Smolic. "Fast and accurate optical flow based depth map estimation from light fields." *Irish Machine Vision and Image Processing Conference (IMVIP)*. 2017.

# SABOM

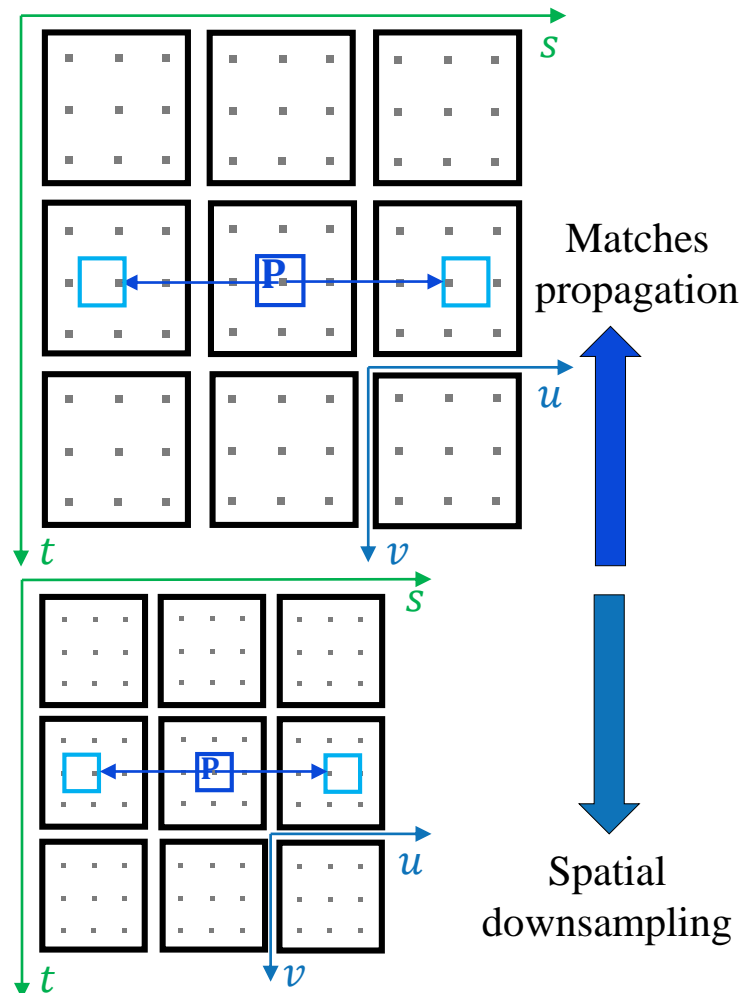
Application to disparity map estimation: LF CPM + PF

- **Gaussian pyramid computed for all views**
- **Starting from coarse level, matches between descriptors are computed on a sparse grid**
- **Estimates are propagated to the next level**
- **At finest level, forward backward check is performed**
- **Check threshold  $\theta_{desc}$  is used to remove outliers**

- **Apply permeability filter to obtain dense disparity map from sparse matches**

CPM

PF



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# Datasets

- **HCI benchmark - [lightfield-analysis.uni-konstanz.de](http://lightfield-analysis.uni-konstanz.de)**
  - 24 light fields
  - Disparity range  $[-3.6, 3.5]$
- **INRIA Dense - [clim.inria.fr/Datasets/InriaSynLF/index.html](http://clim.inria.fr/Datasets/InriaSynLF/index.html)**
  - 39 light fields
  - Disparity range  $[-5.5, 5.3]$
- **Stanford Gantry - [lightfield.stanford.edu](http://lightfield.stanford.edu)**
  - 12 light fields

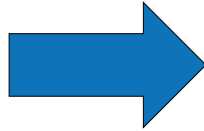
# ROC performance

## Procedure

- Split all light fields views in 32x32 patches and add AWGN with  $\sigma_n = 0.1$
- Create pair of ground truth matches using ground truth disparity maps
- Create a non-match pair for each match by randomly sampling a patch within a 5 pixels radius around the ground truth match



patch



match



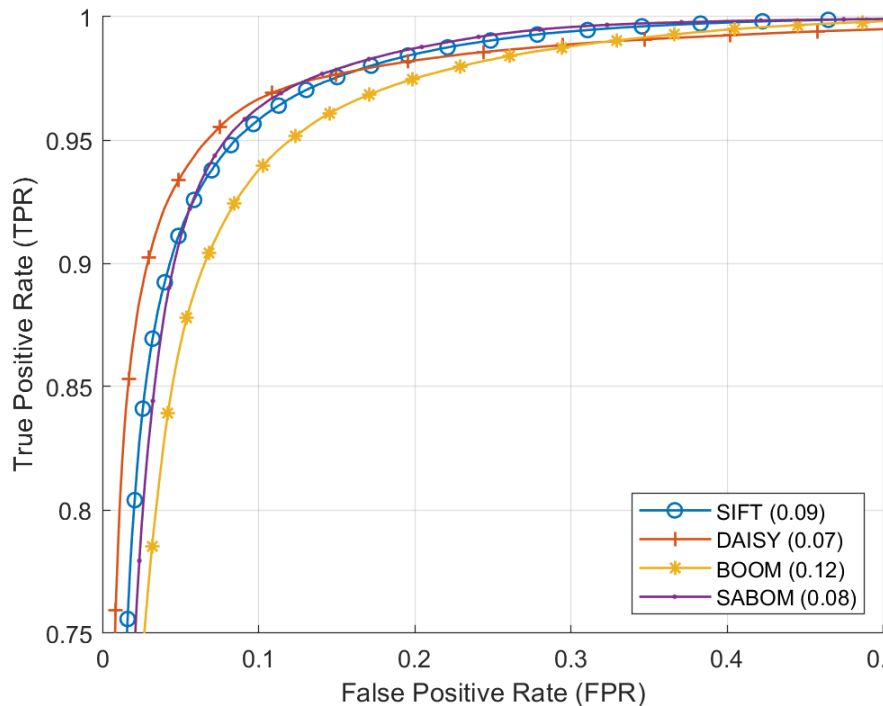
non-match

- Compute descriptor distance between pairs of patches
- Use a sweeping threshold to compute ROC performance curve

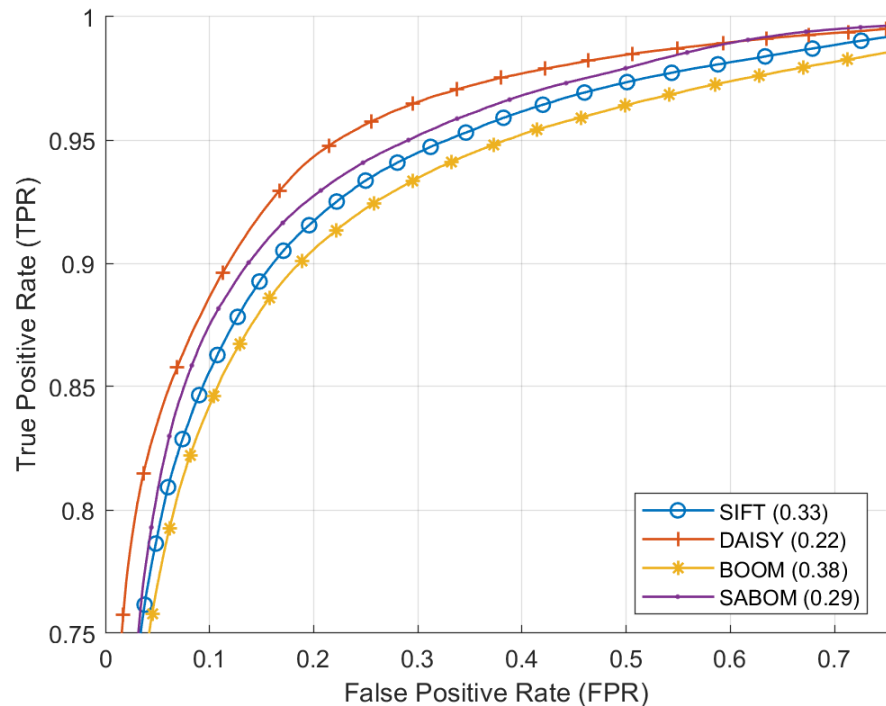
# ROC performance

ROC curves

HCI dataset



INRIA Dense dataset



# ROC performance

Computation time

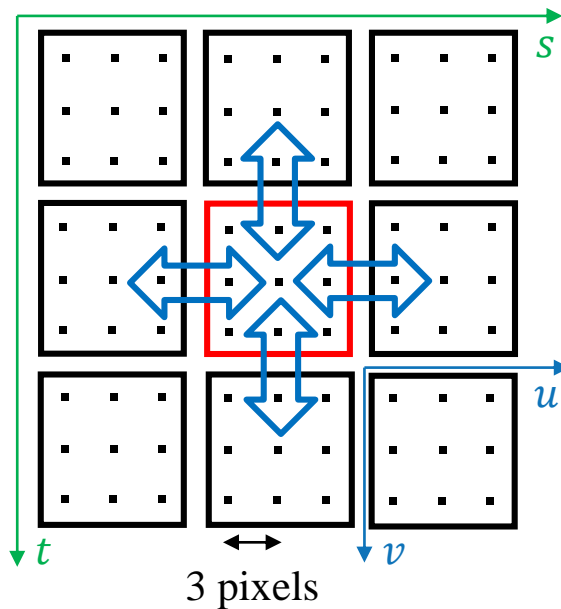
Descriptor	Single descriptor ( $\mu s$ )	Patch distance (ms)
SIFT	0.635	0.152
DAISY	0.254	0.132
BOOM	0.130	0.020
SABOM (proposed)	0.176	0.021



# Disparity map estimation

## Procedure

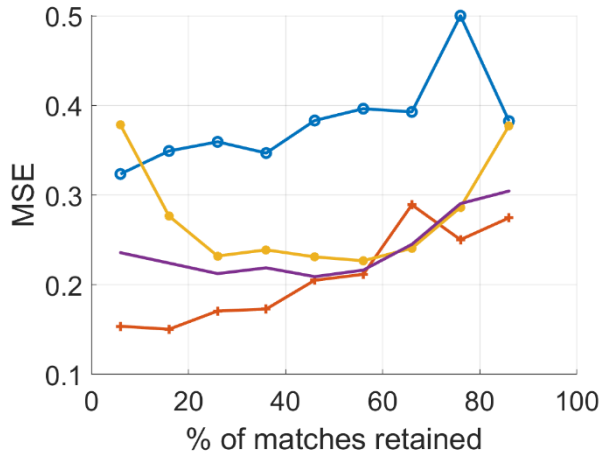
- Estimated on the center view, using 3x3 surrounding views
- CPM grid spacing: 3 pixels
- Sweep  $\theta_{desc}$  range so that matching density ranges from 0 to 100%
- Quality evaluated using MSE



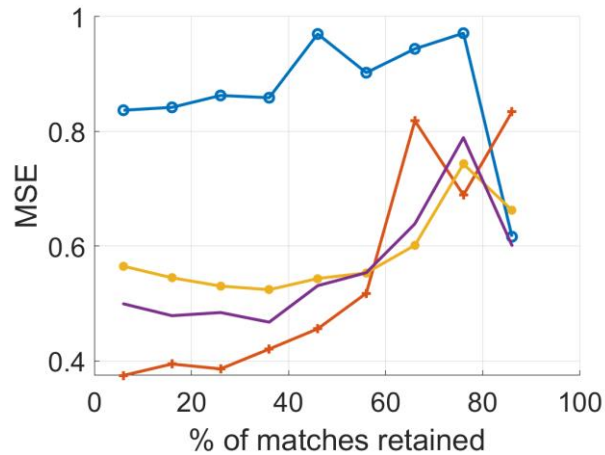
# Disparity map estimation

Accuracy

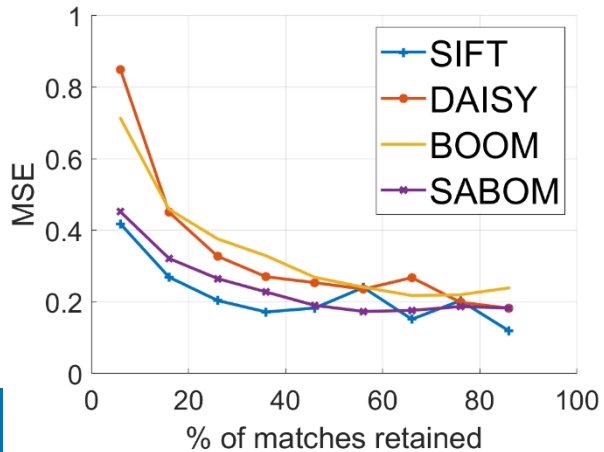
HCI CPM



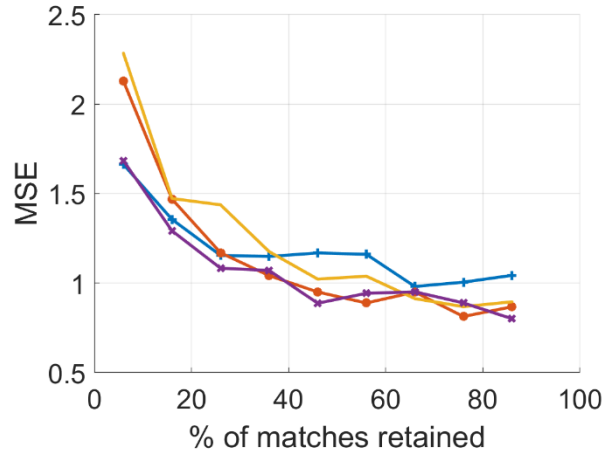
INRIA CPM



HCI CPM + PF



INRIA CPM + PF



# Disparity map estimation

Computation time (in seconds)

	CPM Pyramid Descriptor				CPM Matching	PF
	Level 2	Level 1	Level 0	Total		
SIFT	0.131	0.496	1.952	2.579	0.527	0.569
DAISY	0.054	0.219	0.843	1.116	0.423	
BOOM	0.025	0.099	0.382	0.520	0.325	
SABOM	0.059	0.236	0.564	0.893	0.367	

# Disparity map estimation

Qualitative results

**More results online:**

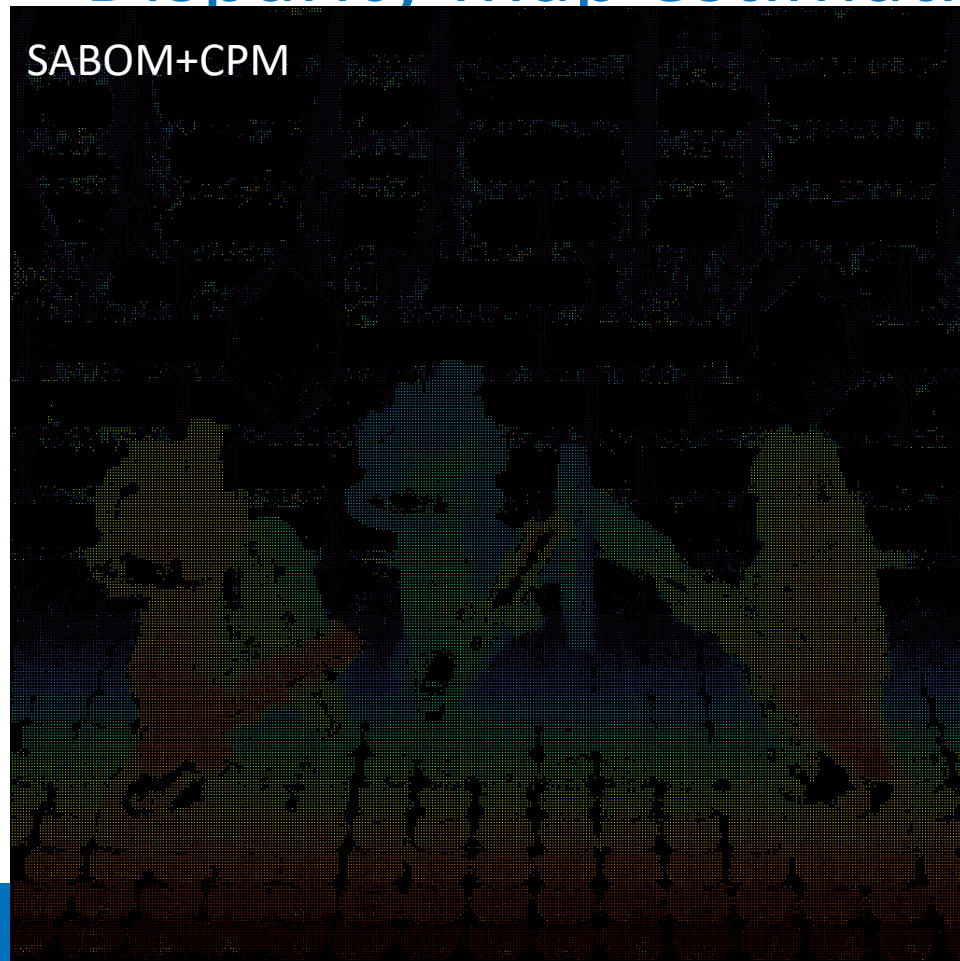
**[v-sense.scss.tcd.ie/research/a-spatio-angular-binary-descriptor-for-fast-light-field-inter-view-matching](http://v-sense.scss.tcd.ie/research/a-spatio-angular-binary-descriptor-for-fast-light-field-inter-view-matching)**

**Code available:**

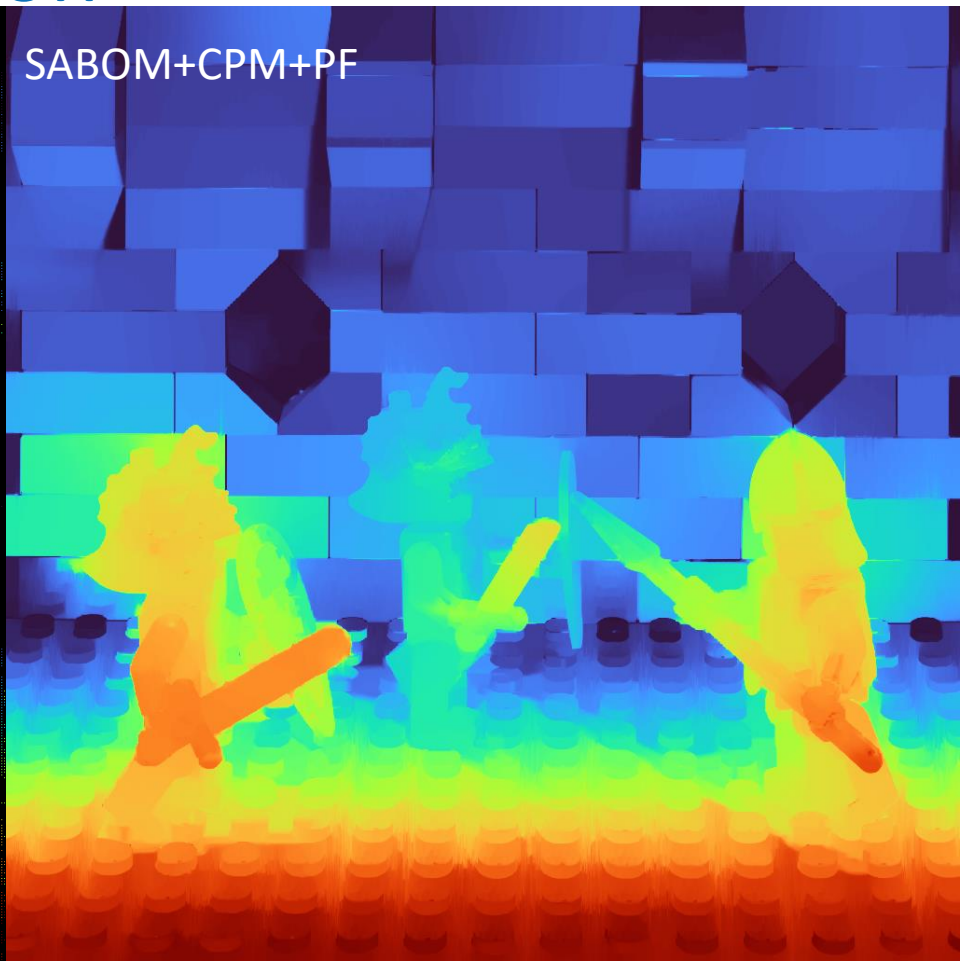
**[github.com/V-Sense](https://github.com/V-Sense)**

# Disparity map estimation

SABOM+CPM



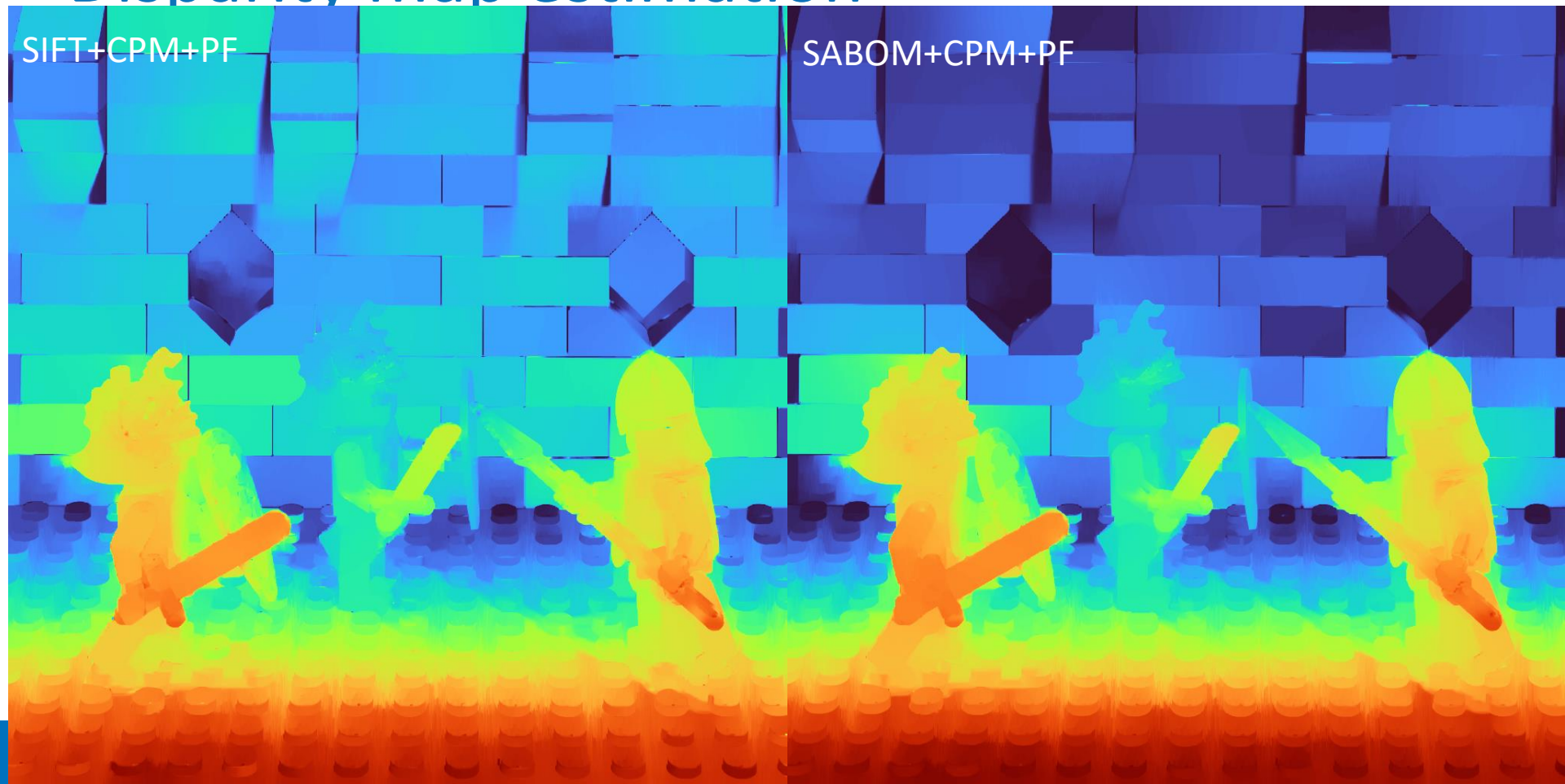
SABOM+CPM+PF



# Disparity map estimation

SIFT+CPM+PF

SABOM+CPM+PF



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# Conclusion

## Summary

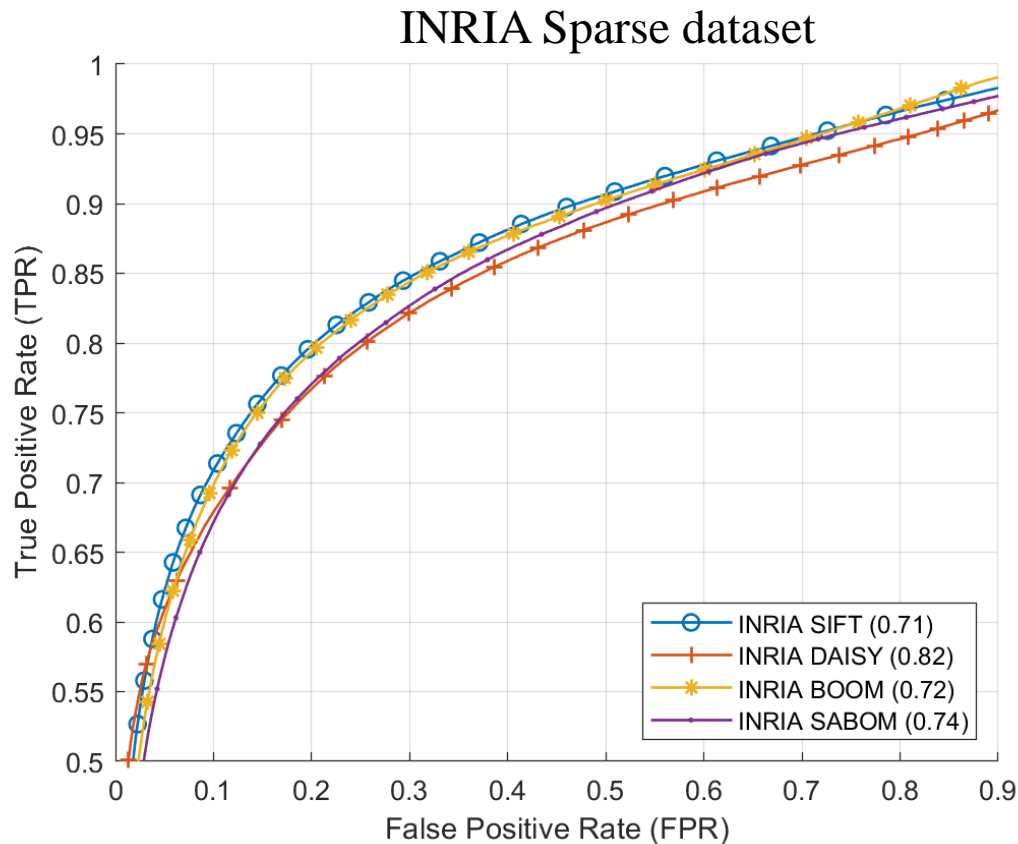
- A novel binary descriptor is introduced for fast and accurate light field inter view matching
- The light field spatial and angular gradients are exploited
- Experiments demonstrate improved performance compared to existing descriptors in terms of ROC performance and when applied to disparity estimation



# Conclusion

## Limitations

- Not suited for sparsely sampled light fields



# Conclusion

## Future work

- Address limitations
- Use in existing framework for more accurate disparity estimation
- Include in other applications
  - LFBM5D
  - Light field colour correction

Dąbala, Łukasz, et al. "Efficient Multi-image Correspondences for On-line Light Field Video Processing." *Computer Graphics Forum*. Vol. 35. No. 7. 2016.

Alain, Martin, and Aljosa Smolic. "Light field denoising by sparse 5D transform domain collaborative filtering." *2017 IEEE 19th International Workshop on Multimedia Signal Processing (MMSP)*. IEEE, 2017.

Alain, Martin, and Aljosa Smolic. "Light field super-resolution via LFBM5D sparse coding." *2018 25th IEEE international conference on image processing (ICIP)*. IEEE, 2018.

Grogan, Mairead and A. Smolic, "L2 based colour correction for light field arrays", in Proc. ACM CVMP, 2019.

Matysiak, Pierre, et al. "High quality light field extraction and post-processing for raw plenoptic data." *IEEE Transactions on Image Processing* 29 (2020): 4188-4203.



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# V-SENSE

## Many Thanks!

- [alainm@tcd.ie](mailto:alainm@tcd.ie)
- <https://v-sense.scss.tcd.ie/research/light-field-imaging/>
- <https://v-sense.scss.tcd.ie/research/a-spatio-angular-binary-descriptor-for-fast-light-field-inter-view-matching/>
- <https://github.com/V-Sense/>