

Real Time Disparity Estimation

with Variational Methods

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I. Introduction

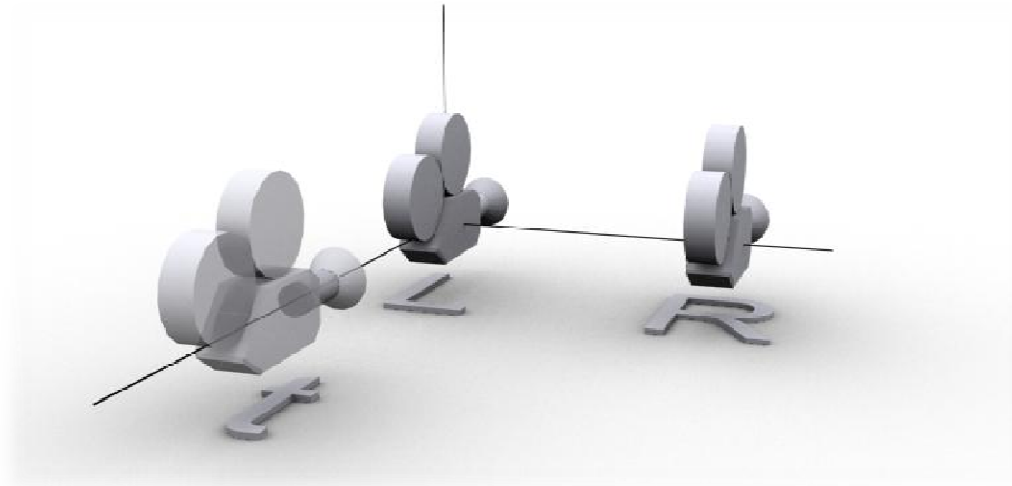
- Disparity Estimation
- Methods Overview

II. Real-Time Variational Approach

- Euler-Lagrange Equation
- Full Multi Grid
- Multi Level Adaptation Technique

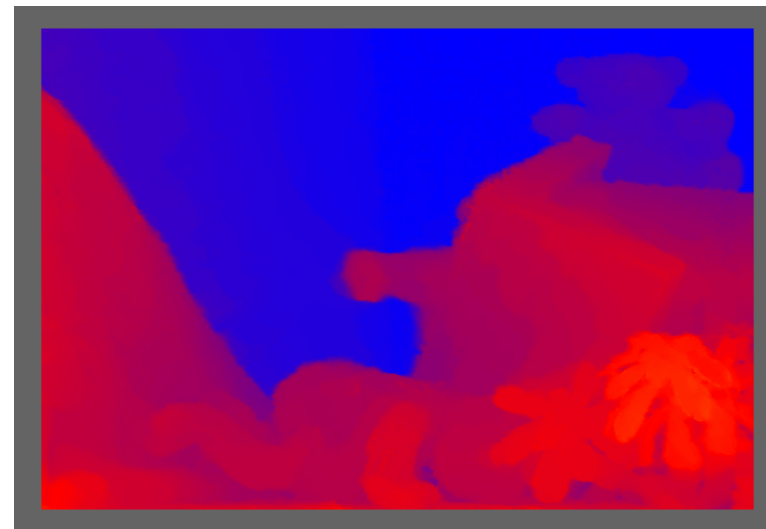
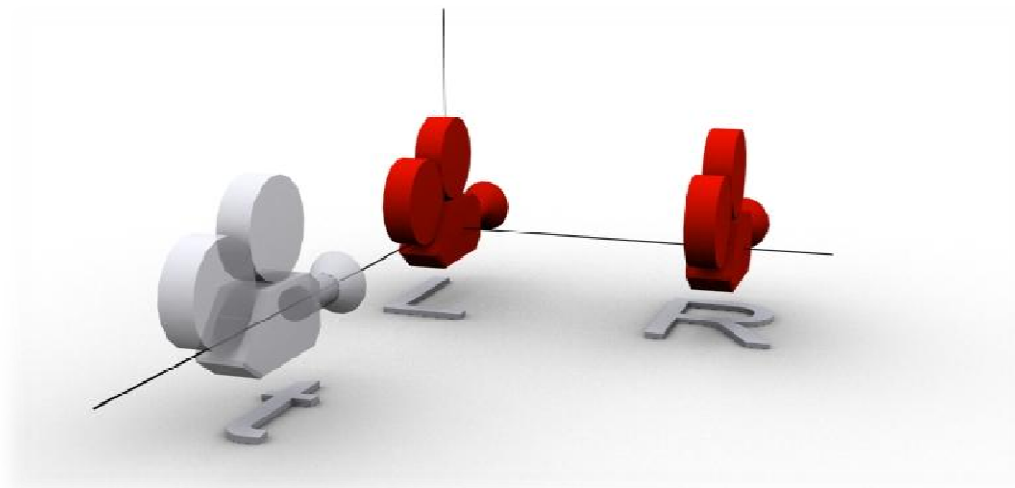
III. Results

- Speed / Quality Discussion
- Future Work



- **Disparity field** is a vector field of pattern of apparent **motion** of objects, surfaces, and edges in a visual scene caused by the relative motion between an observer (**eyes** or **cameras**) and the scene.

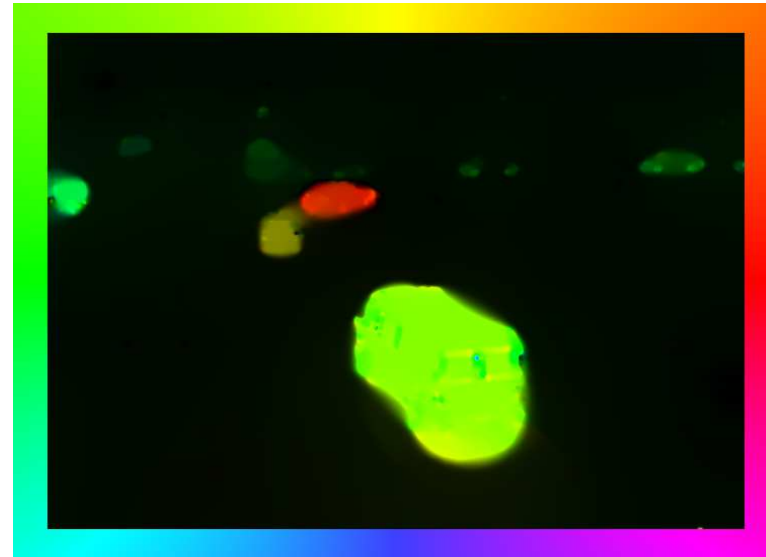
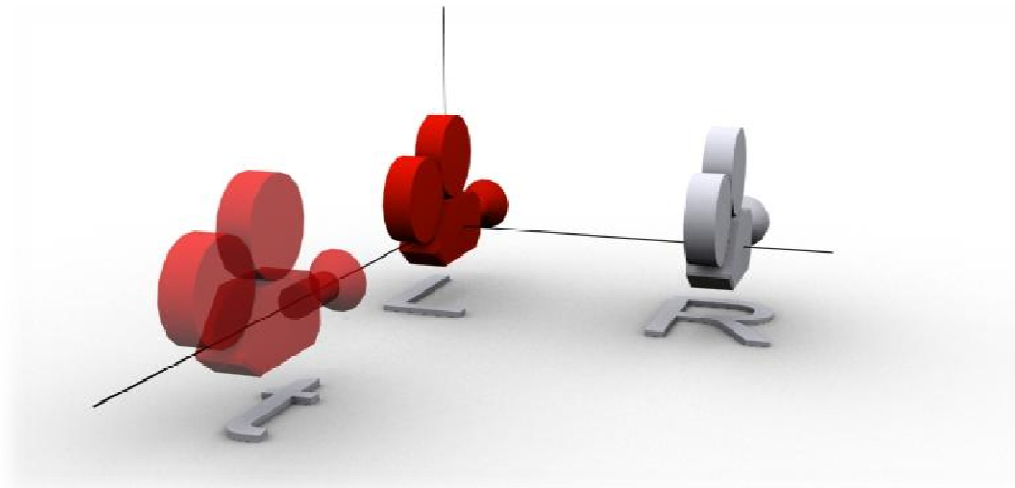
Disparity Field



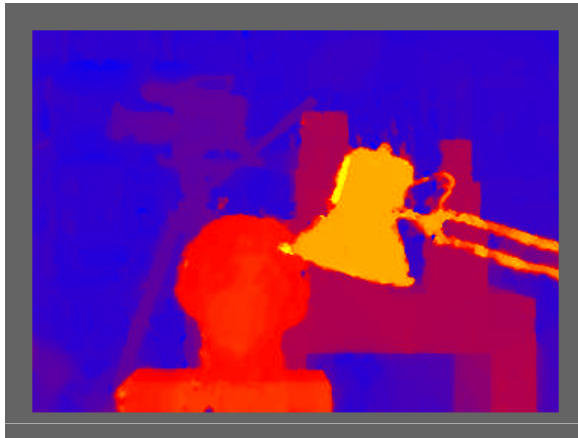
3px far <<

>> near 16px

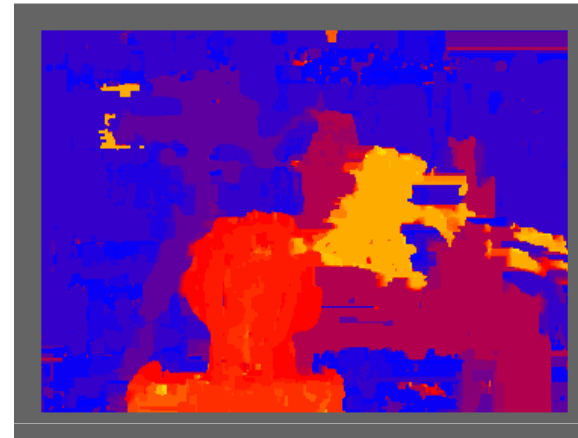
Optical Flow Field



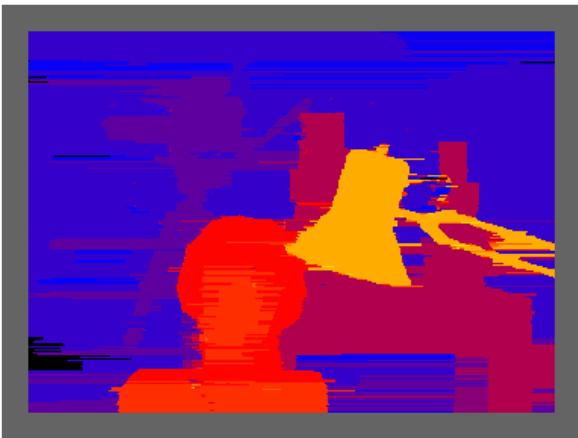
Methods Overview



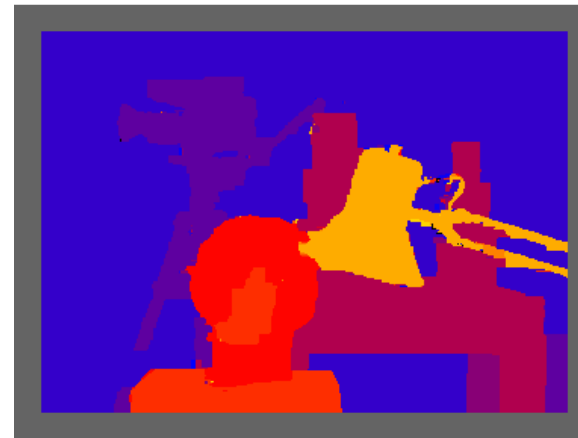
Ours: Variational Method



Infection Method



Belief Propagation Method



Graph Cuts Method

Euler-Lagrange Equation

- *Constancy assumption* on the grey value:

$$I_1(x + z(x, y), y) = I_2(x, y)$$

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- Data term:

$$\text{Arg min } \|I_1(x + z(x, y), y) - I_2(x, y)\|_2$$

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- Energy Functional:

$$E(z(x, y)) = \iint_{\Omega} \|I_1(x + z(x, y), y) - I_2(x, y)\|^2 dS$$

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- Preserving discontinuities:

$$E(z(x, y)) = \iint_{\Omega} \|I_1(x + z(x, y), y) - I_2(x, y)\|^2 + \varphi \cdot \Psi(|\nabla z(x, y)|^2) dS$$

- Where $\Psi(s^2) = \lambda \sqrt{\lambda^2 + s^2} - \lambda^2$ is non-linear regularization function

Euler-Lagrange Equation

- The mathematicians *Leonard Euler* (1707-1783) and *Joseph-Louis Lagrange* (1736-1813) belong to the founders of the calculus of variations.
- Steady state of a function:
 $f = f(x)$
- Satisfies the equation:
 $f_x(x) = 0$
 - where $f_x(x) \equiv \frac{df(x)}{dx}$
- Solution:
 $x = a$
- Steady state of a functional
 $E(f(x)) = \int_l F(x, f(x), f_x(x)) dx$
- Satisfies the equation:
 $\frac{\partial F}{\partial f} - \frac{d}{dx} \frac{\partial F}{\partial f_x} = 0$
- Solution:
 $f(x) = \varphi(x)$

Euler-Lagrange Equation

- A smooth function $z(x, y)$ that minimizes the energy functional:

$$E(z(x, y)) = \iint_{\Omega} F\left(x, y, z(x, y), \frac{\partial z(x, y)}{\partial x}, \frac{\partial z(x, y)}{\partial y}\right) dS$$

- Satisfies necessary the Euler-Lagrange equation:

$$\frac{\partial F}{\partial z} - \frac{d}{dx} \frac{\partial F}{\partial z_x} - \frac{d}{dy} \frac{\partial F}{\partial z_y} = 0$$

- where $z_u = \frac{\partial z}{\partial u}$ and $z_v = \frac{\partial z}{\partial v}$.

- In our case we can write the Euler-Lagrange equation:

$$I_{1x}(x+z, y) \cdot (I_1(x+z, y) - I_2(x, y)) - \varphi \cdot \operatorname{div}(\Psi(|\nabla z|^2) \nabla z) = 0$$

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- Which could be rewritten in form:

$$A_1 z + A_2 - \varphi \cdot \Lambda(z) = 0$$

- where A_i are linear operators,
- and $\Lambda(z)$ is a *non-linear* differential operator.

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- We reformulate this non-linear equation system:

$$L(x) = f$$

- where x denotes vector $(u^T, v^T)^T$,
- L is a *non-linear* operator,
- f stands for the right hand side.

Full Approximation Scheme

- Multigrid approach was offered by R. Fedorenko in 1961
- Very effective for solving elliptic and parabolic PDE systems

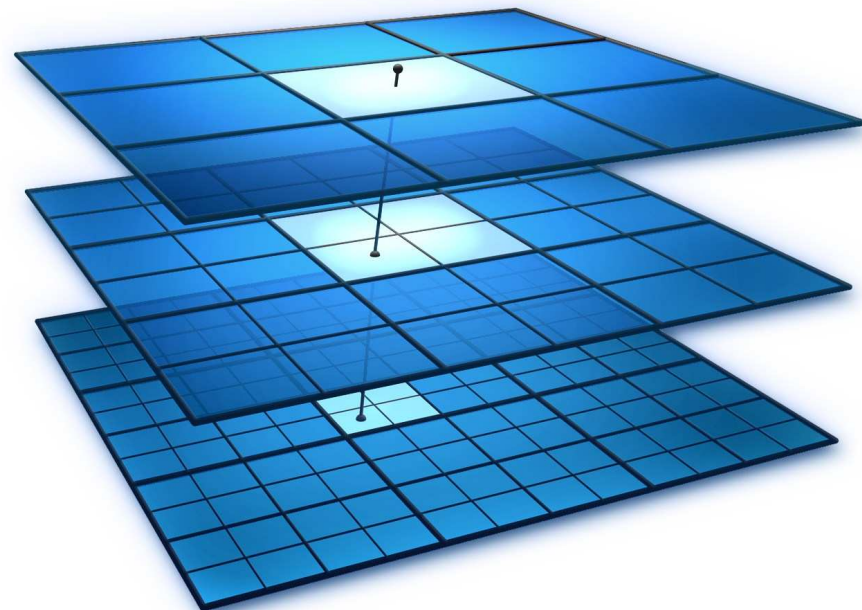
- Uses a set of grids with different step sizes:

$$G(h) = (\Delta h, \Delta h)$$

$$G(H^i) = G(2^i h)$$

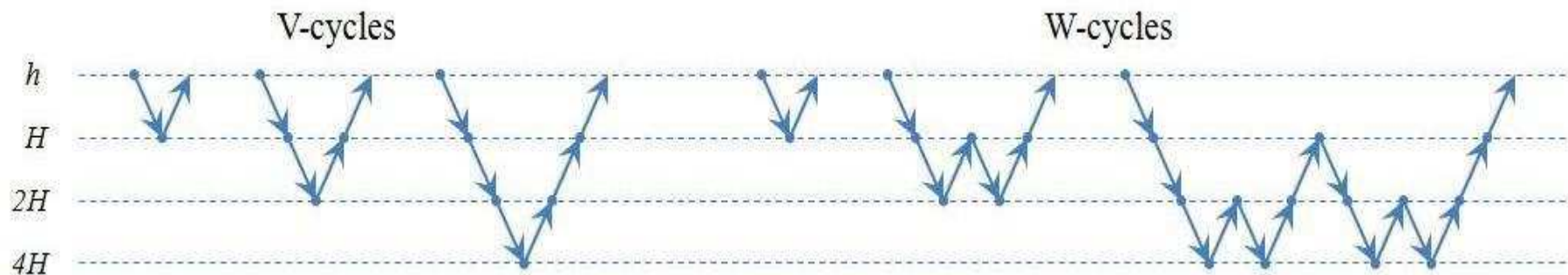
with approximation to a grid

$$L^h x^h = f^h$$

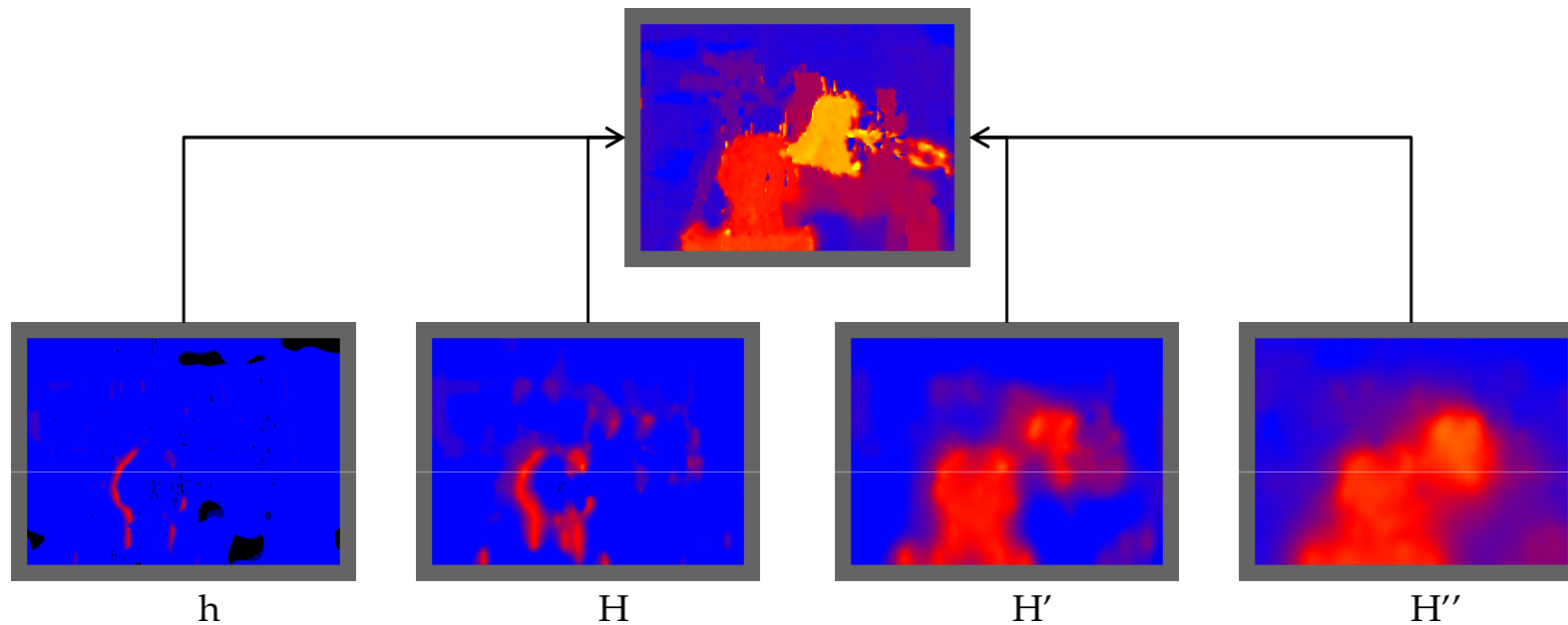


Full Approximation Scheme

- First way
 - Solve problem on a given grid by calling Multigrid on a coarse approximation to get a good guess to refine
- Second way
 - Think of error as a sum of sine curves of different frequencies
 - Same idea as FFT solution, but not explicit in algorithm
 - Each call to Multigrid responsible for suppressing coefficients of sine curves of the lower half of the frequencies in the error

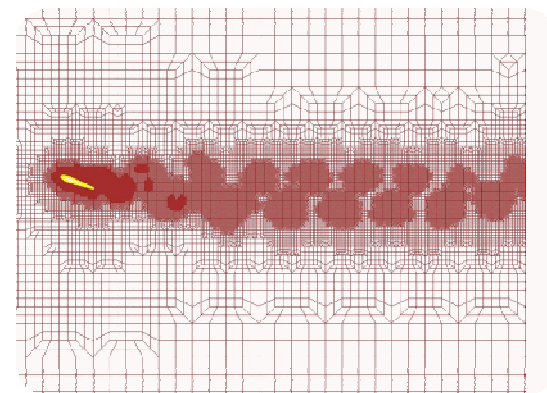
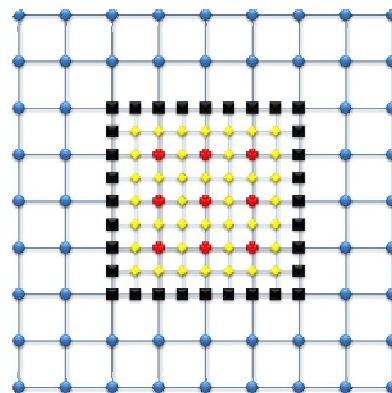
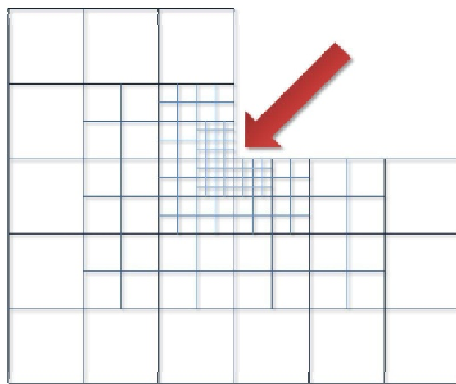


Full Multigrid

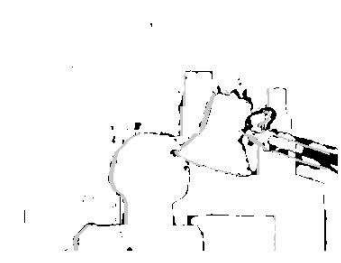


Multilevel Adaptation Technique

- Adaptive grids
 - Increased computational savings over a static grid approach
 - Effective multi grid and adaptive grid combination
 - Easy and effective parallelizable
- Irregular grids
 - Concentrate nodes near to a local peculiarity
 - Simple combination of two grid levels



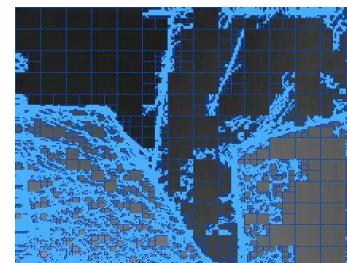
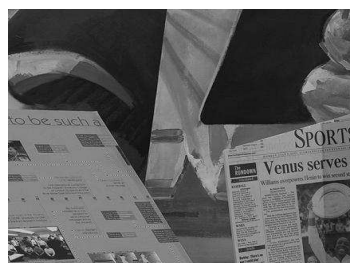
Results



Tsukuba

Bad pixels: 3,33%

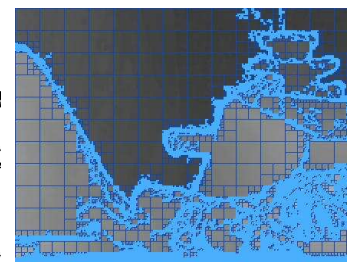
Time: 465 ms



Venus

Bad pixels: 1,15%

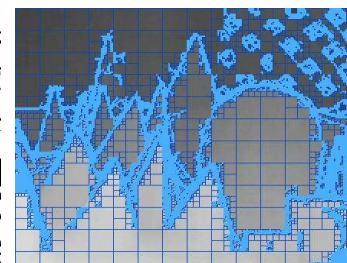
Time: 265 ms



Teddy

Bad pixels: 5,88%

Time: 305 ms



Cones

Bad pixels: 4,61%

Time: 285 ms

Results

- Excerpt from the evaluation table generated by the Middlebury stereo evaluation webpage (error threshold = 2 pixels).

Algorithm	Tsukuba			Venus			Teddy			Cones			Average percent of bad pixels
	nonocc	all	disc	nonocc	all	disc	nonocc	all	disc	nonocc	all	disc	
DoubleBP	0,83	1,24	4,49	0,10	0,35	1,46	1,41	4,13	4,73	1,71	7,20	5,16	2,72
CoopRegion	0,77	1,00	4,14	0,11	0,18	1,53	2,14	3,41	6,61	2,10	5,95	6,24	2,80
...													
Ours: RealTimeVar	1,33	3,13	6,94	0,27	1,07	3,16	1,30	2,30	3,87	2,31	3,43	6,90	3,00
...													
MultiResGC	0,67	1,05	3,64	0,22	0,46	2,97	4,20	7,13	11,6	3,22	8,80	8,07	4,30
RealTimeBP	1,25	3,04	6,66	0,63	1,53	7,68	5,68	8,27	10,2	2,90	9,11	8,27	5,43
RealTimeGPU	1,34	3,27	7,17	1,02	1,90	12,4	3,90	8,65	10,4	4,37	10,8	12,3	6,46
Infection	6,34	7,81	22,8	2,70	3,66	26,0	12,8	18,3	33,5	10,7	16,6	30,1	15,9

- Main contributions so far:
 - The most accurate disparity estimation method among real-time methods (*according to Middlebury University web-site*)
 - Effective combination of multigrids with grid adaptation techniques
 - Multigrid powered by *null*-cycles
- Future work:
 - From the disparity estimation to the pure optical flow reconstruction
 - User recognition and authentication with stereo web-cams
 - Use your body as an input device in interactive games
 - Robots with stereo cameras as eyes

The end

- Thank you for your attention

Ready to answer your questions ☺