

Creatis



Mean-Shift scale parameters optimization on the EGEE grid

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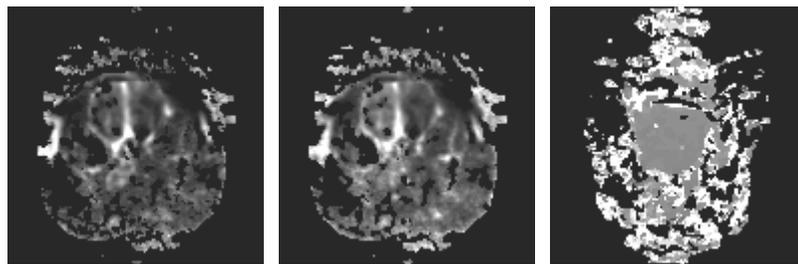
29/06/2010



Medical Context

--- clustering is a powerful tool to analyze multi-dimensional data

- Images and parameters
 - Multi-modality: Ultrasound, PET, MRI, X-ray, etc.



cbf peak ttp
MRI parameters for penumbra segmentation
in ischemia [M.Wiart]

→
clustering

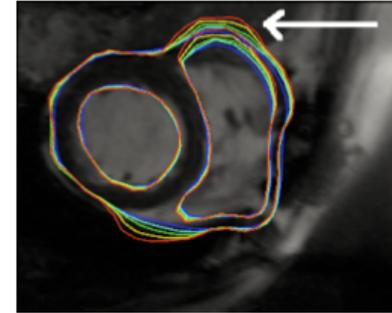
- Characterize pathology
- Segmentation
- Detect disease

- Clustering
 - Filtering: noise removing, edges preserving, etc.
 - Segmentation: cluster or separation of clusters

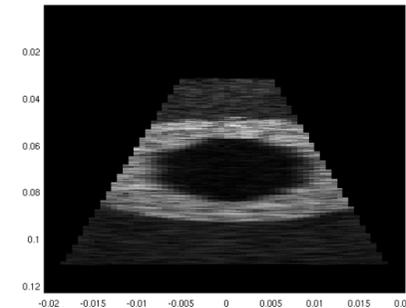
Grid computing for medical imaging

--- used in production for a wide range of applications

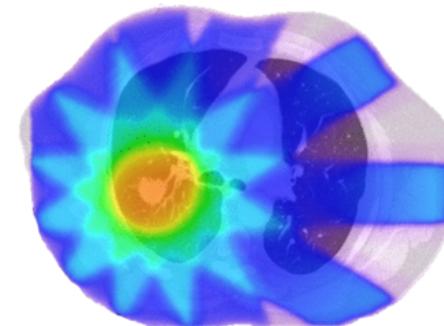
- Parameter sweeps
 - Functional MRI [Olabarriaga *et al*, Biogrid'07]
 - Image segmentation [Maheshwari *et al*, HG'09]
[Specovius *et al*, HG'10]
- Data parallelization
 - On image lines / slices / volumes / sequences
 - Pharmacokinetic modelling [Blanquer *et al*, HG'08]
 - Image simulation
- Monte-Carlo simulation
 - Radiotherapy [Maigne *et al*, PPL 2004]
 - Image simulation (PET, CT)



Sweep on segmentation parameters
[Maheshwari *et al*, HG'09]



Ultrasonic 2D+t simulation



GATE simulation

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OUTLINE

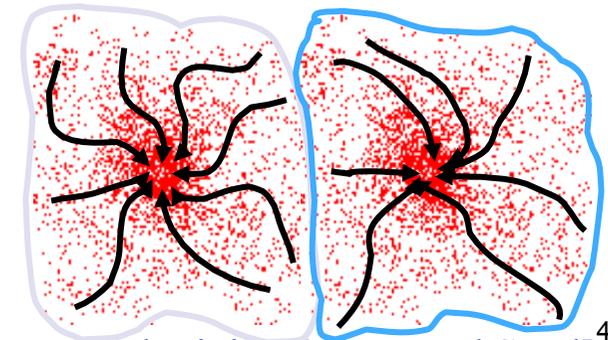
1. Context

2. Mean-Shift presentation

3. Experiments and results on bandwidth optimization

4. Grid deployment and execution

5. Conclusion

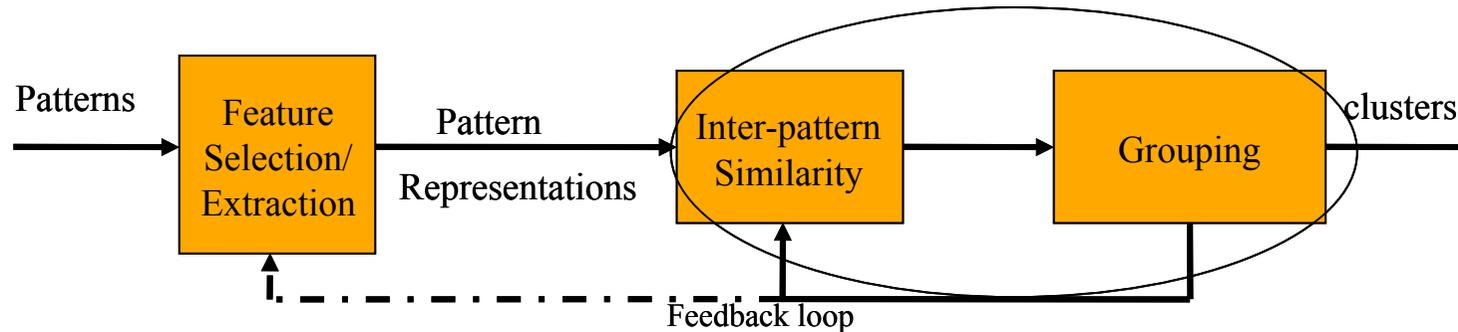


[Yaron Ukrainitz & Bernard Sarel]⁴

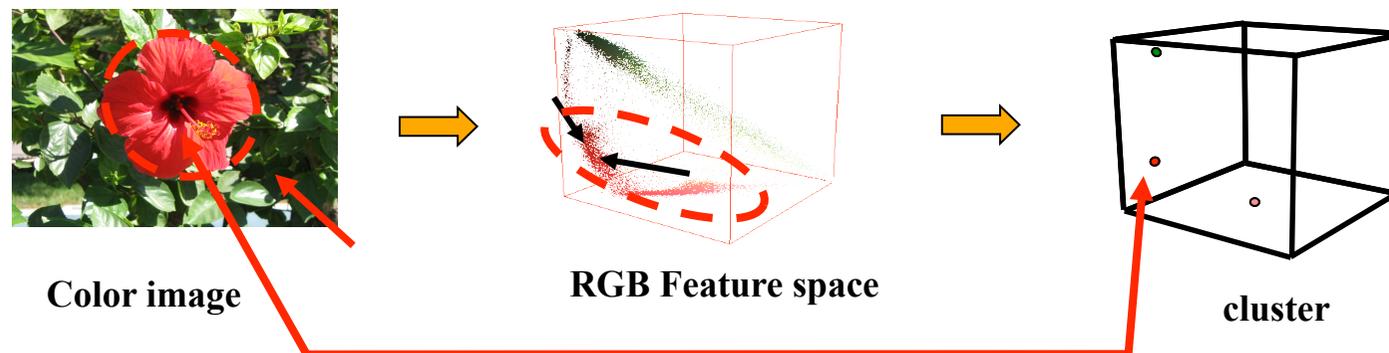
General Clustering Framework

---It is also the mean-shift framework

- General Framework



- Representation in Feature Space



Mean-Shift presentation

---Why we use mean-shift as our clustering approach?

- General Principle

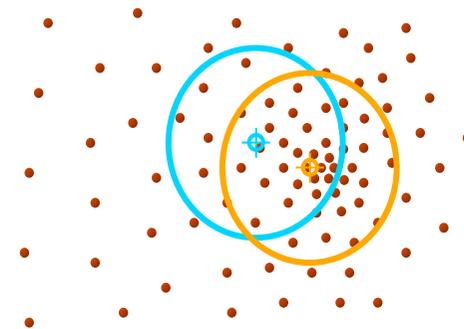
- Goal: finding modes in a set of data samples
- Method: probability density gradient of non-parametric estimation

- Characteristics

- Can handle arbitrary feature spaces
- Does not assume any prior shape (e.g. elliptical) on data clusters
- Iterative hill climbing algorithm

- Challenges

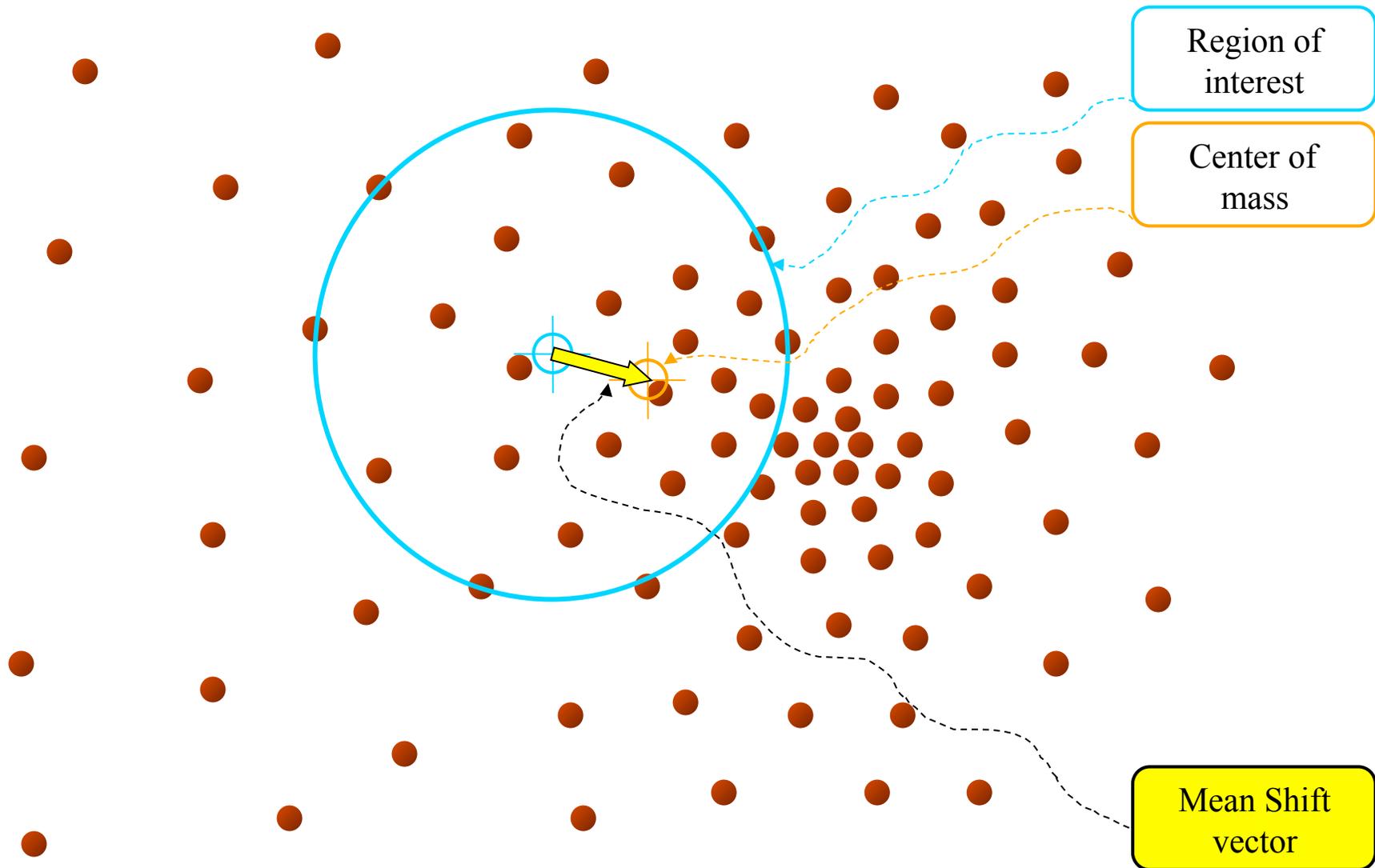
- Finding local optimum (mode)
- Setting scale parameters H (window size)
- Time consuming



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[Yaron Ukrainitz & Bernard Sarel]

Intuitive Description

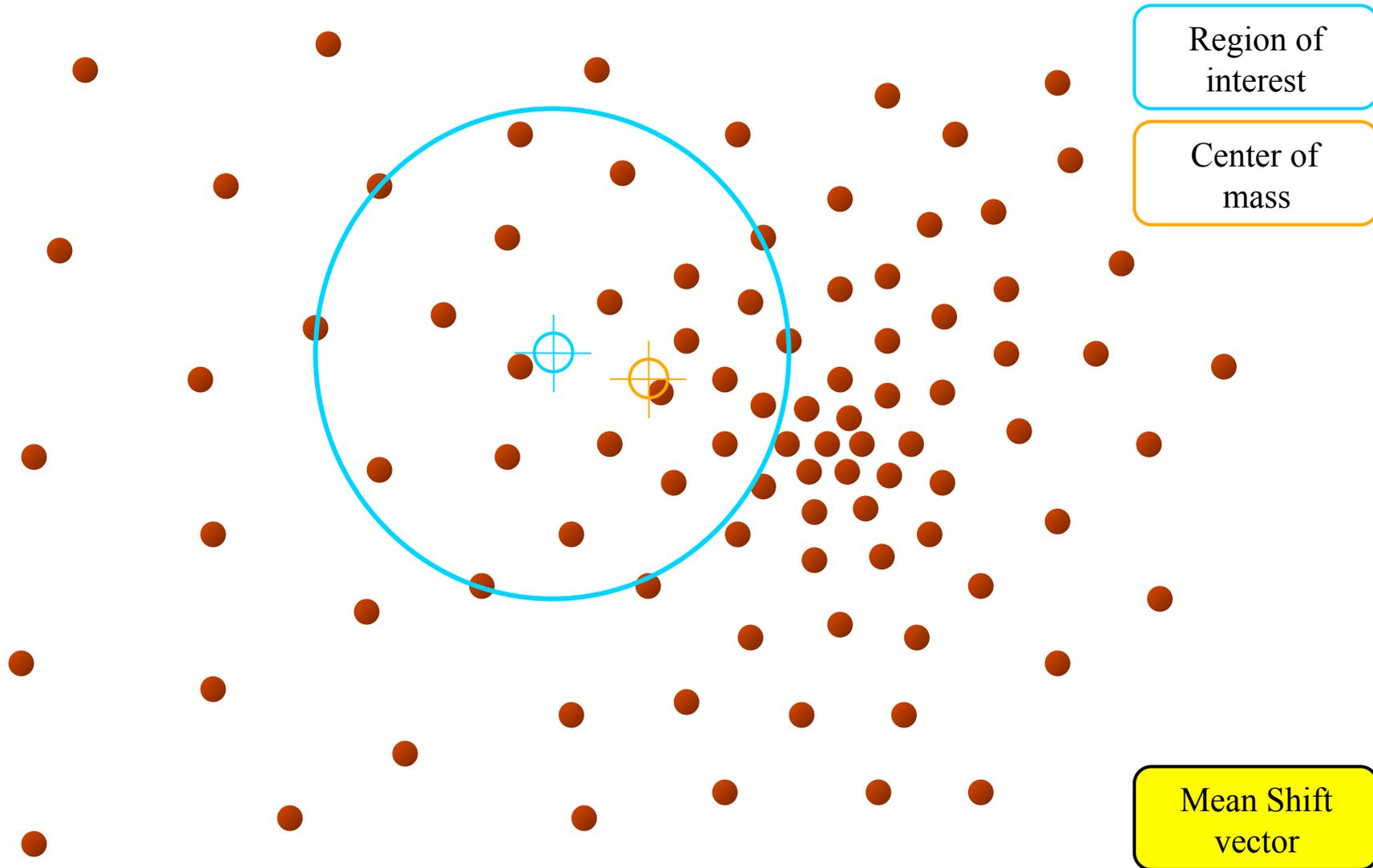


Objective : Find the densest region

Distribution of identical points

7

Intuitive Description

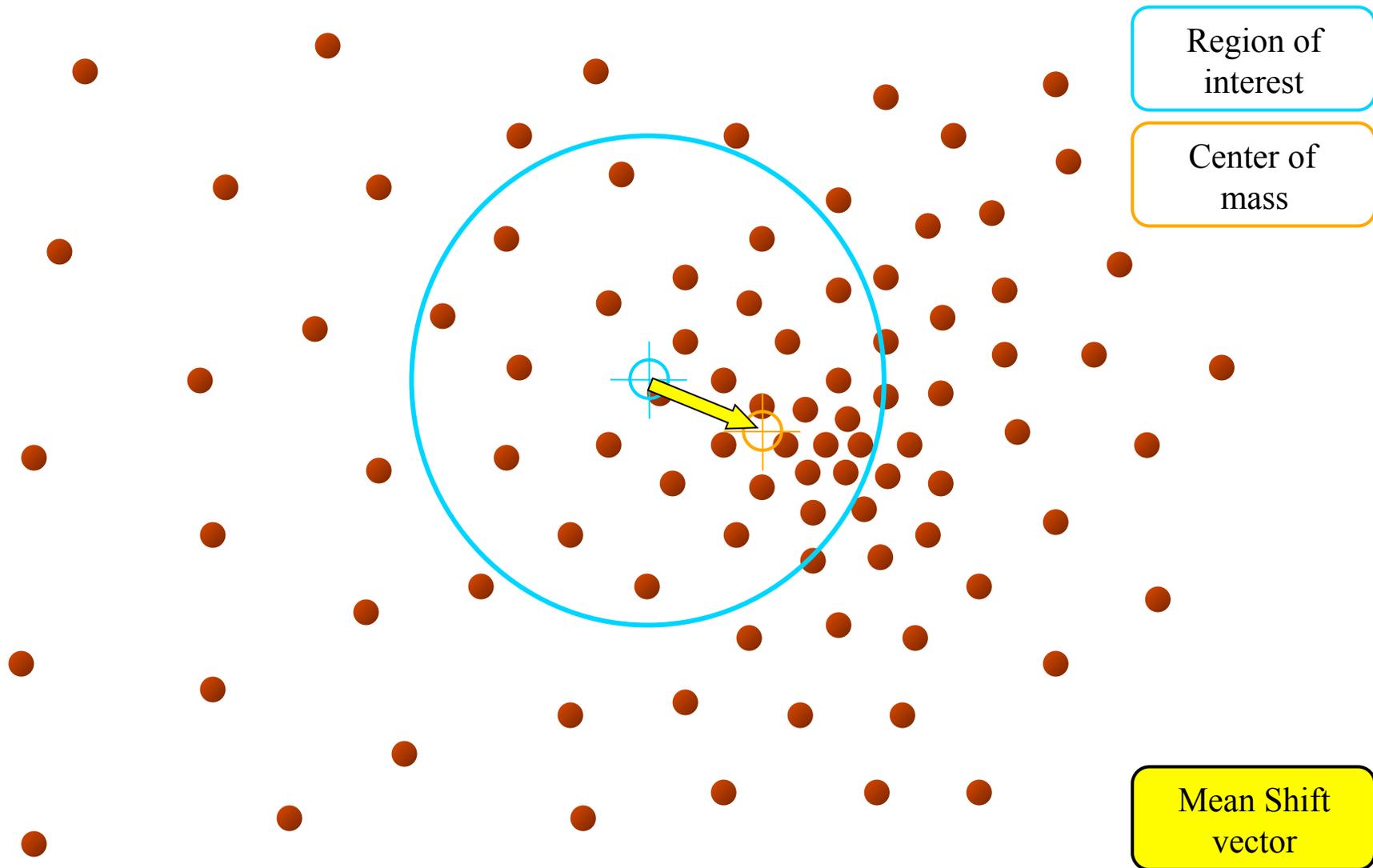


Objective : Find the densest region

Distribution of identical points

8

Intuitive Description

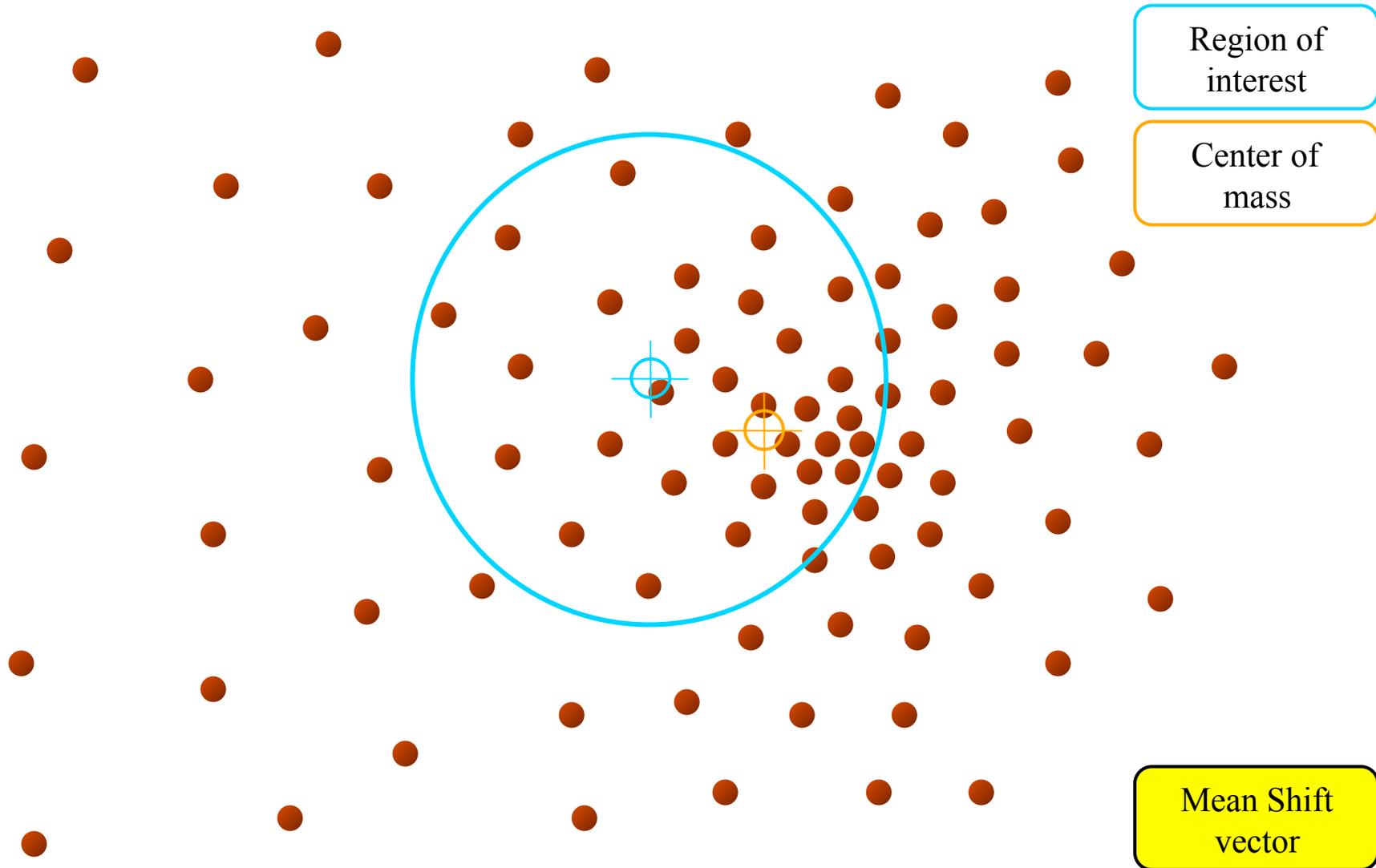


Objective : Find the densest region

Distribution of identical points

9

Intuitive Description

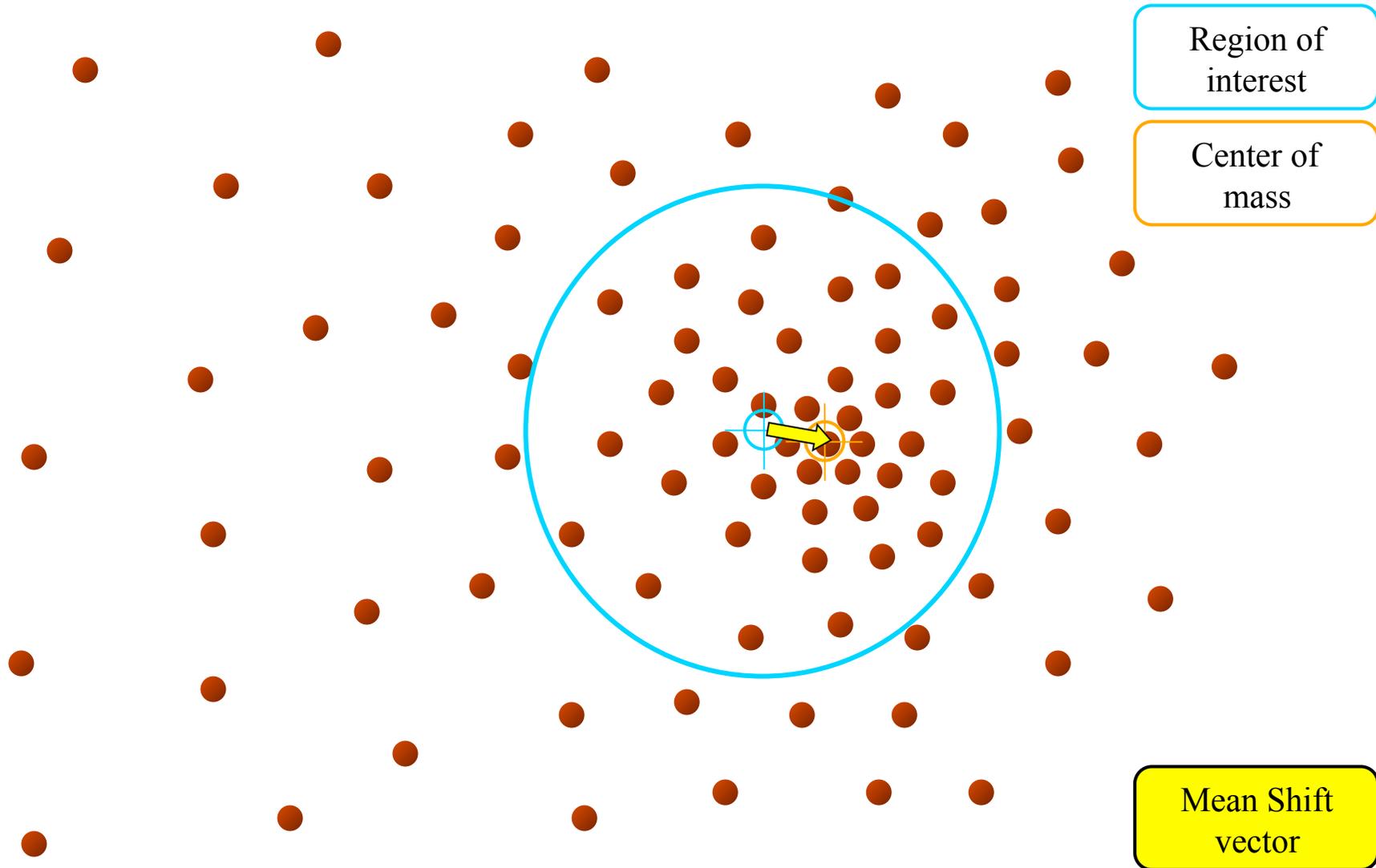


Objective : Find the densest region

Distribution of identical points

10

Intuitive Description

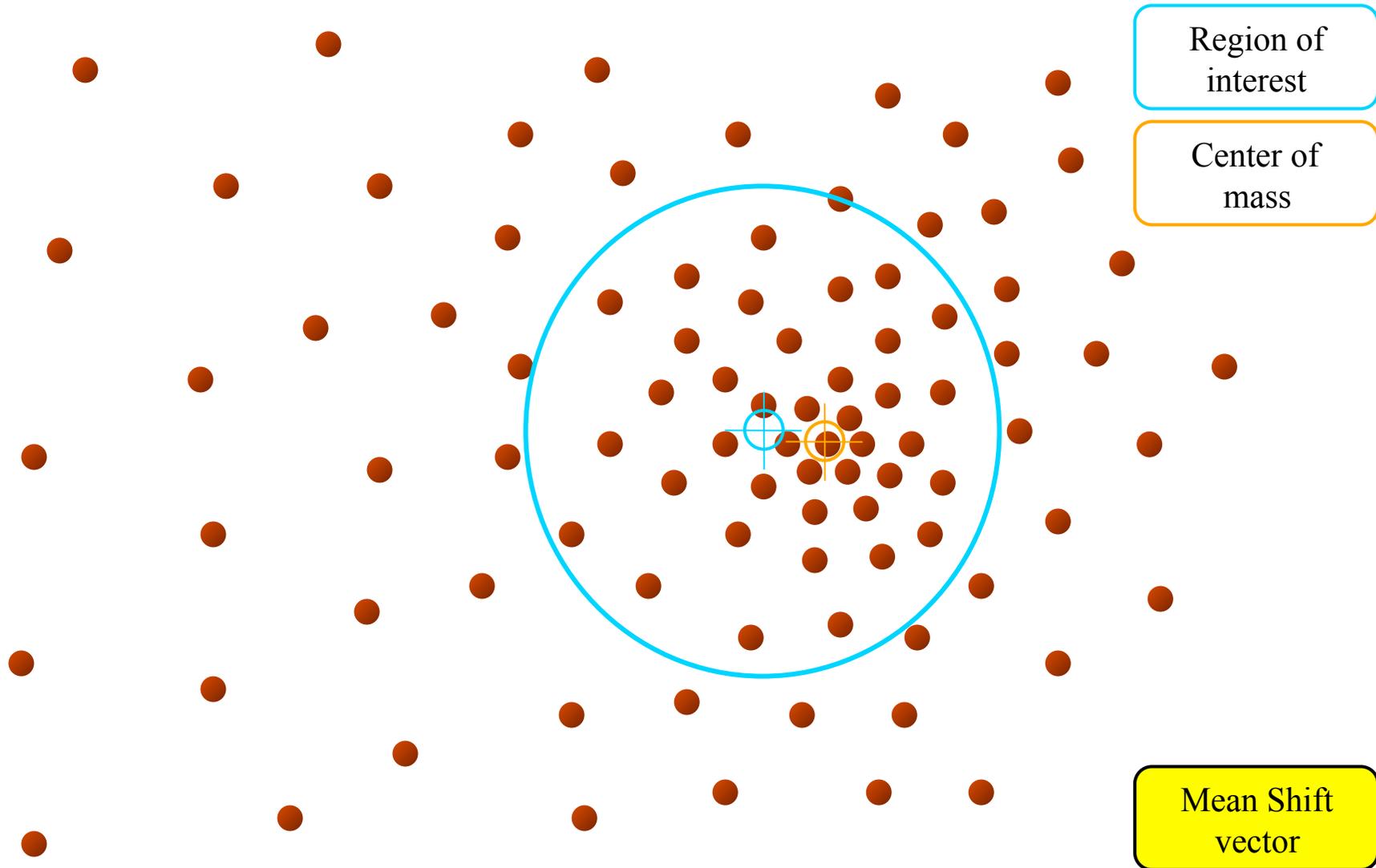


Objective : Find the densest region

Distribution of identical points

11

Intuitive Description

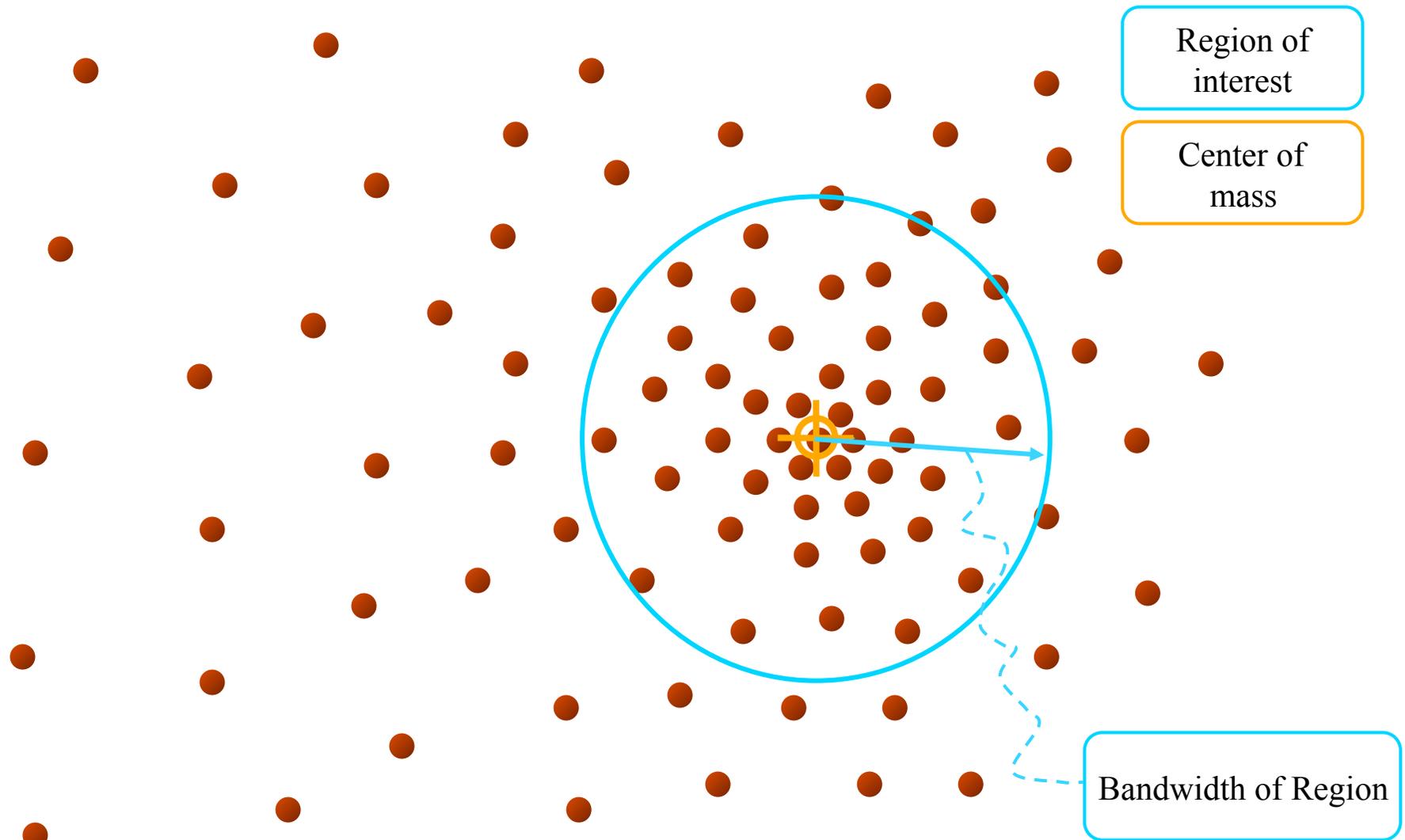


Objective : Find the densest region

Distribution of identical points

12

Intuitive Description



Objective : Find the densest region

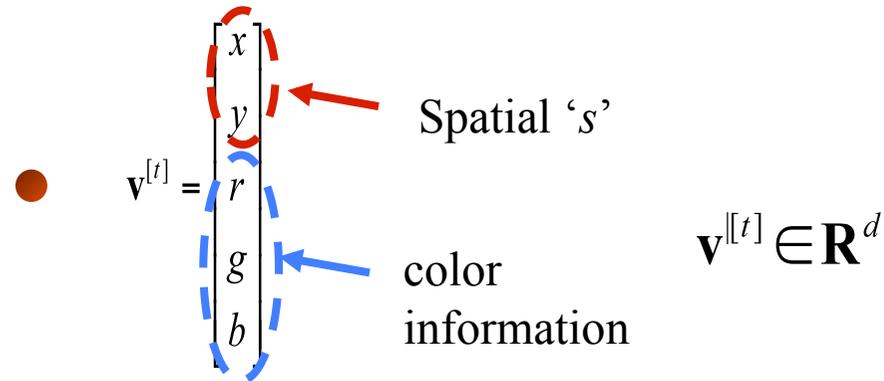
Distribution of identical points

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Mean-Shift Filtering Procedures

[Comaniciu 99']

- 1. For a given pixel



- 2. Initialize

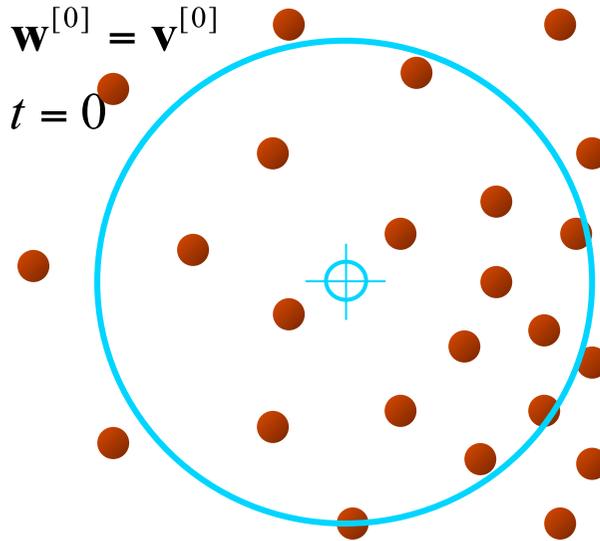
$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_{red}^2 & 0 & 0 \\ 0 & 0 & 0 & h_{green}^2 & 0 \\ 0 & 0 & 0 & 0 & h_{blue}^2 \end{bmatrix}$$

and

$$\mathbf{w}^{[0]} = \mathbf{v}^{[0]}$$

$t = 0$

h_s spatial scale
 h_r, h_g, h_b color scales



Mean-Shift Filtering Procedures

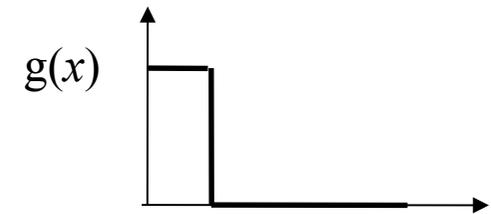
[Comaniciu 99']

- 3. Compute the centre of mass \mathbf{w}

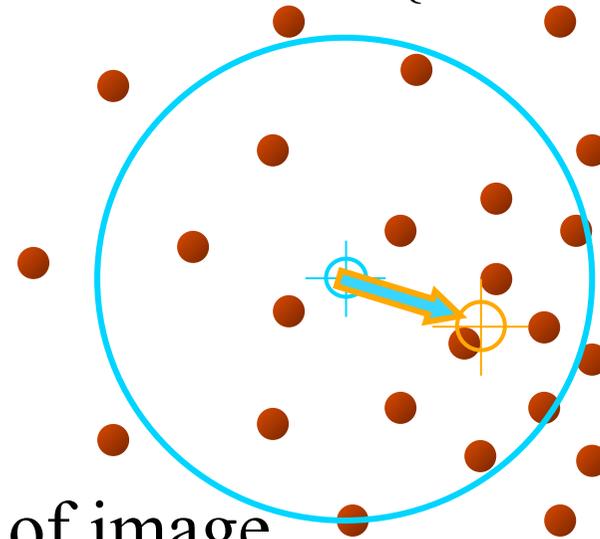
$$\mathbf{w}^{[t+1]} = \frac{\sum_{i=1}^n g(d^2(\mathbf{w}^{[t]}, \mathbf{v}_i \cdot \mathbf{H})) \mathbf{v}_i}{\sum_{i=1}^n g(d^2(\mathbf{w}^{[t]}, \mathbf{v}_i \cdot \mathbf{H}))}$$

until convergence

$d(\cdot)$: Mahalanobis distance



$$g(x) = \begin{cases} 1 & 0 < x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



- 4. Make $\mathbf{v}^{[t]} = \mathbf{w}^{[mode]}$
- 5. Repeat steps 2-4 for each pixel of image

Our Problem: optimize bandwidth \mathbf{H}

- Two questions:
 - 1. Can we use gradient ascent to optimize \mathbf{H} ?
 - 2. Is it worth using an anisotropic bandwidth \mathbf{H} ?

$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_r^2 & 0 & 0 \\ 0 & 0 & 0 & h_r^2 & 0 \\ 0 & 0 & 0 & 0 & h_r^2 \end{bmatrix}$$

Isotropic H


$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_{red}^2 & 0 & 0 \\ 0 & 0 & 0 & h_{green}^2 & 0 \\ 0 & 0 & 0 & 0 & h_{blue}^2 \end{bmatrix}$$

Anisotropic H

Description of Experiments

- Tested image



Original Image



Noisy Image

$$\sigma^2 = 30$$

- Isotropic Bandwidth \mathbf{H}

$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_r^2 & 0 & 0 \\ 0 & 0 & 0 & h_r^2 & 0 \\ 0 & 0 & 0 & 0 & h_r^2 \end{bmatrix} \begin{matrix} h_s \in [1, 5] \\ h_r \in [20, 240] \end{matrix}$$

- Anisotropic Bandwidth \mathbf{H}

$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_{red}^2 & 0 & 0 \\ 0 & 0 & 0 & h_{green}^2 & 0 \\ 0 & 0 & 0 & 0 & h_{blue}^2 \end{bmatrix} \begin{matrix} h_s \in [1, 5] \\ h_{red} \in [20, 240] \\ h_{green} \in [20, 240] \\ h_{blue} \in [20, 240] \end{matrix}$$

Results isotropic \mathbf{H} VS anisotropic \mathbf{H}

* Results Validation: PSNR

- PSNR results for anisotropic \mathbf{H} is a little better



(a) Original Image

$$\sigma^2 = 30$$



(b) Noisy Image

19.71dB(ref)



(c) Optimal for isotropic \mathbf{H}

28.93dB(3,160)



(d) Optimal for anisotropic parameters

29.01dB

(3,240,140,200)

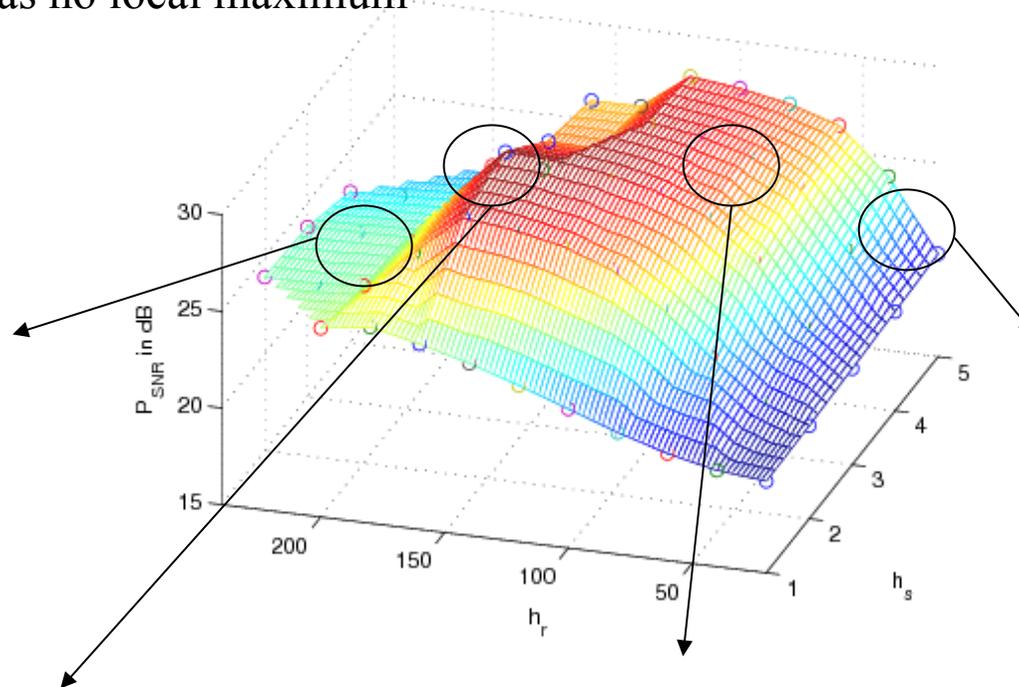
- Gradient Ascent (GA) can optimize bandwidth \mathbf{H}
 - For isotropic \mathbf{H} , GA is robust for searching the optimization.
 - For anisotropic \mathbf{H} , GA can be used if the neighborhood radius is > 2.5

Results (PSNR evolution example against h_s & h_r of isotropic H)

- Gradient Ascent algorithm
 - PSNR has no local maximum



24.17dB(2,200)



20.32dB(5,20)



28.93dB(3,160)



27.96dB(4,100)

Perspectives

- Mean-Shift filtering experiments on more data
 - Synthesized data & RGB images & Medical data

- Choice of bandwidth \mathbf{H}

$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_r^2 & 0 & 0 \\ 0 & 0 & 0 & h_r^2 & 0 \\ 0 & 0 & 0 & 0 & h_r^2 \end{bmatrix}$$

Isotropic H

$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_{red}^2 & 0 & 0 \\ 0 & 0 & 0 & h_{green}^2 & 0 \\ 0 & 0 & 0 & 0 & h_{blue}^2 \end{bmatrix}$$

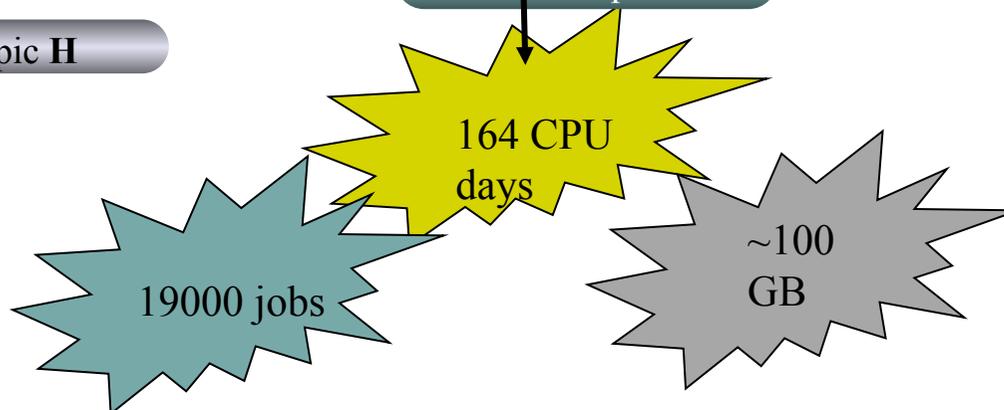
Anisotropic H

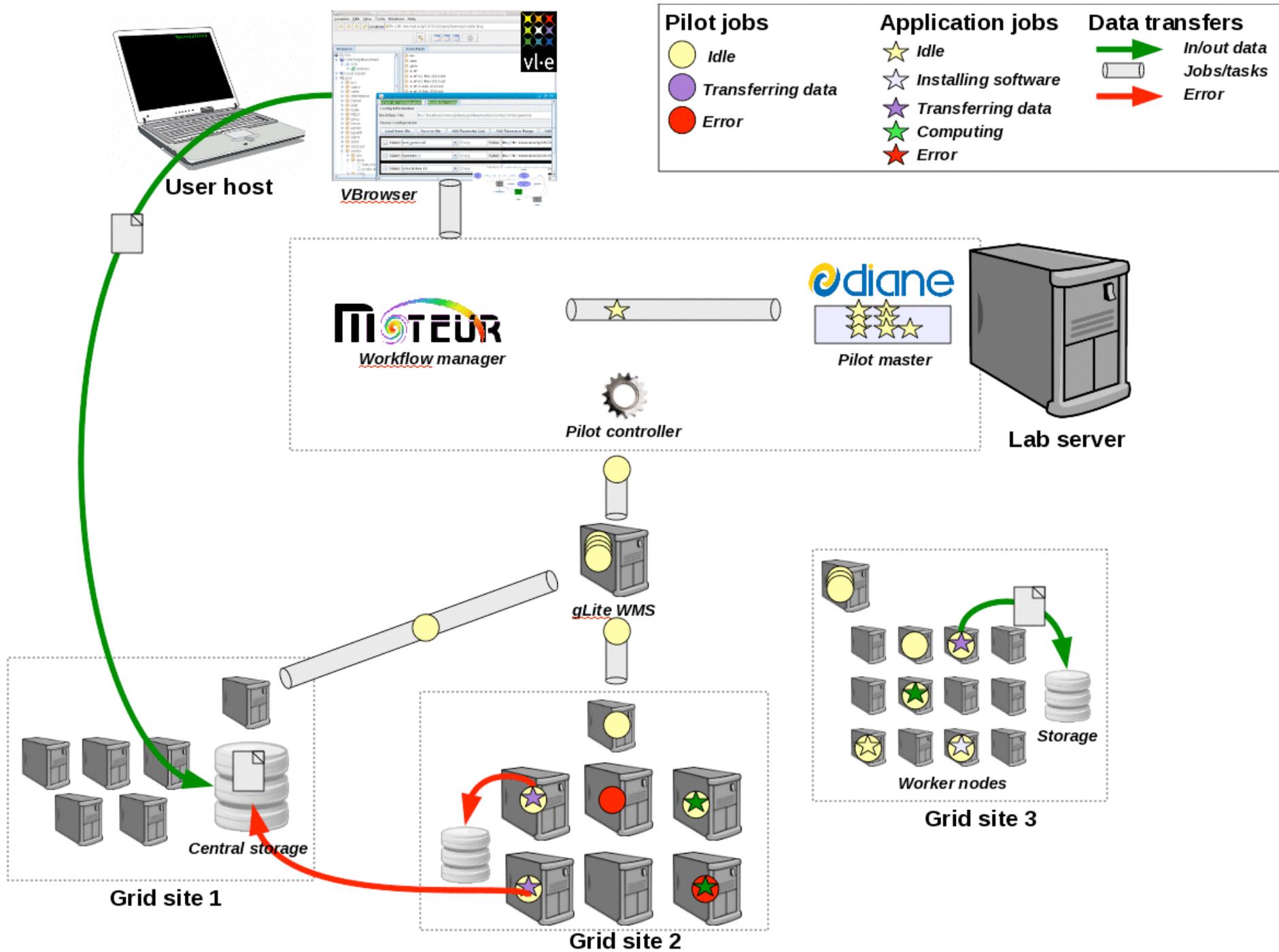
Move here

$$\mathbf{H} = \begin{bmatrix} h_s^2 & 0 & 0 & 0 & 0 \\ 0 & h_s^2 & 0 & 0 & 0 \\ 0 & 0 & h_{red}^2 & h_{rg}^2 & h_{rb}^2 \\ 0 & 0 & h_{rg}^2 & h_{green}^2 & h_{gb}^2 \\ 0 & 0 & h_{rb}^2 & h_{gb}^2 & h_{blue}^2 \end{bmatrix}$$

Range Correlated H

- Need for grid !

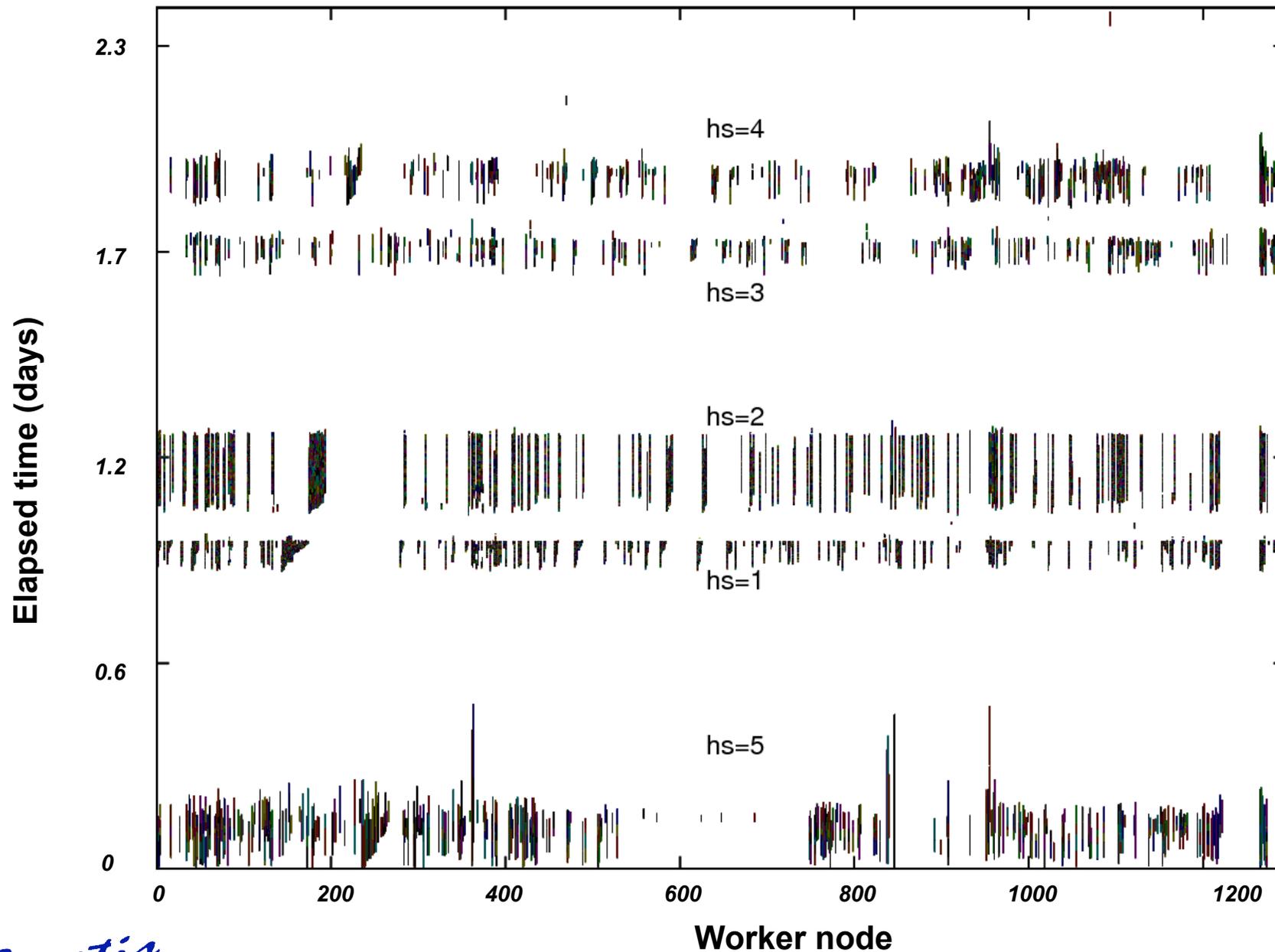


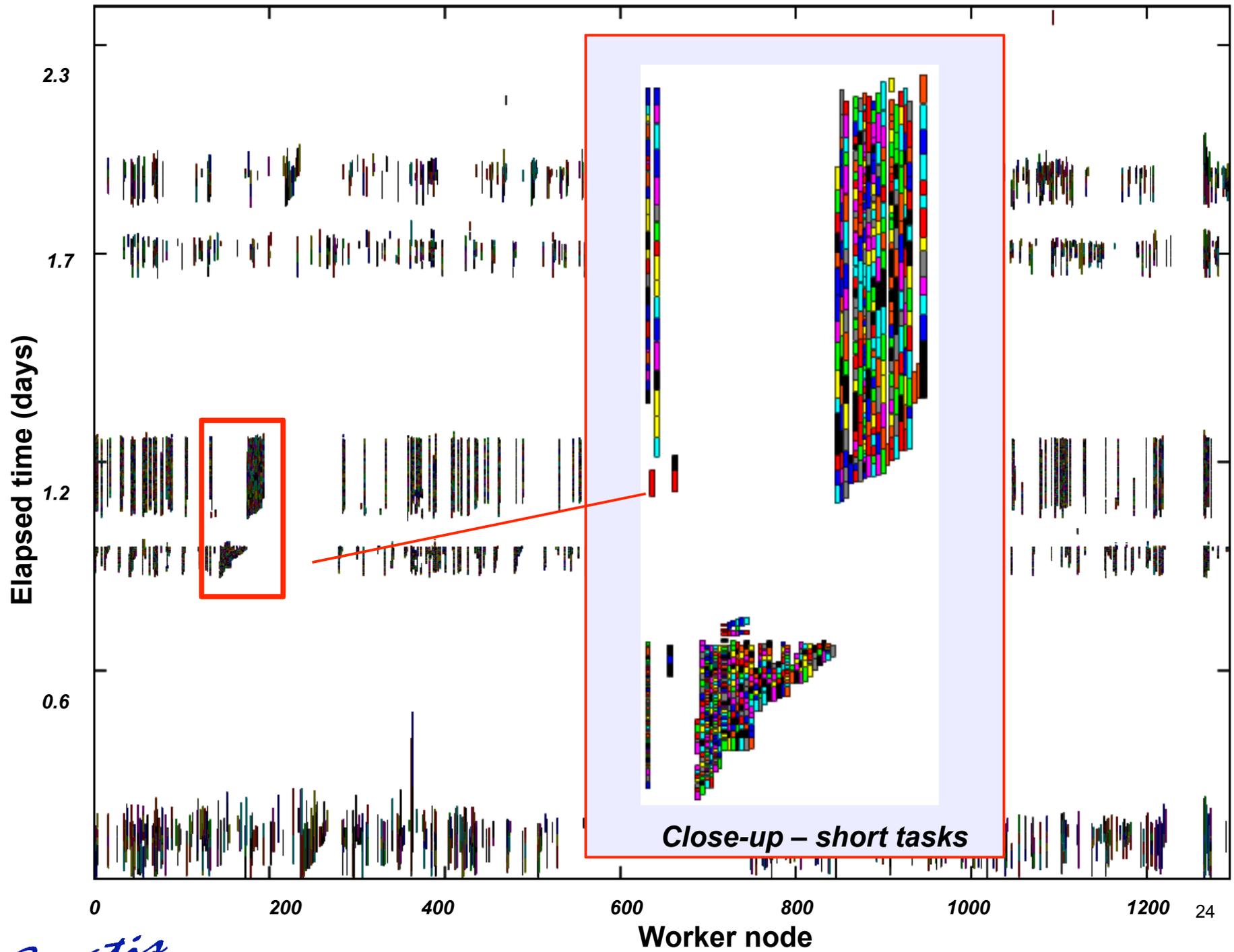


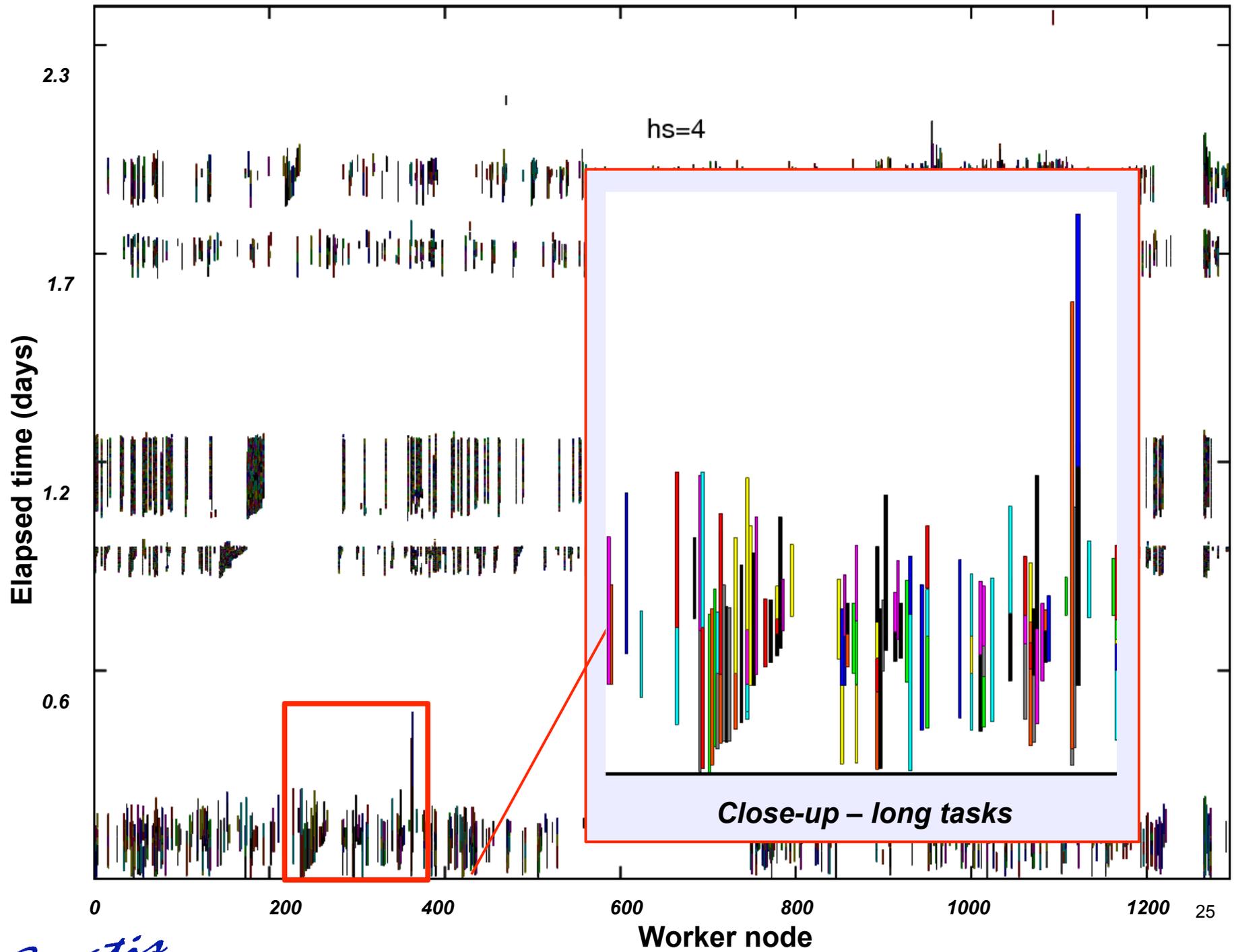
Grid Setup

- Application porting
 - Compiled Matlab for Centos5, 64bits
 - Application described as a workflow
- Infrastructure
 - EGEE, biomed Virtual Organisation
 - ~120 sites
- Handling long jobs
 - Proxy uploaded to myproxy
 - Jobs have watchdog scripts to fetch fresh proxy
- Data transfer reliability
 - Data cached in pilots
 - Results written where produced (fall-back on central)
 - Upload test before execution

Gantt diagram of the experiment







Performance analysis

h_s value	Total CPU time	Elapsed time	Speed-up	Total data transfer time	Produced data	Successful tasks	Total tasks	Error ratio
1	13.0 days	3h25min	92	89.8h	43GB	8,000	8,106	1.3%
2	50.4 days	18h36min	65	63.2h	41GB	8,000	8,929	10.4%
3	17.3 days	13h01min	32	34.8h	5GB	1,000	1,317	24%
4	29.0 days	13h23min	52	44.7h	5GB	1,000	1,089	8.2%
5	54.3 days	11h09min	117	51.9h	5.1GB	1,000	1,179	15.2%

- Error causes

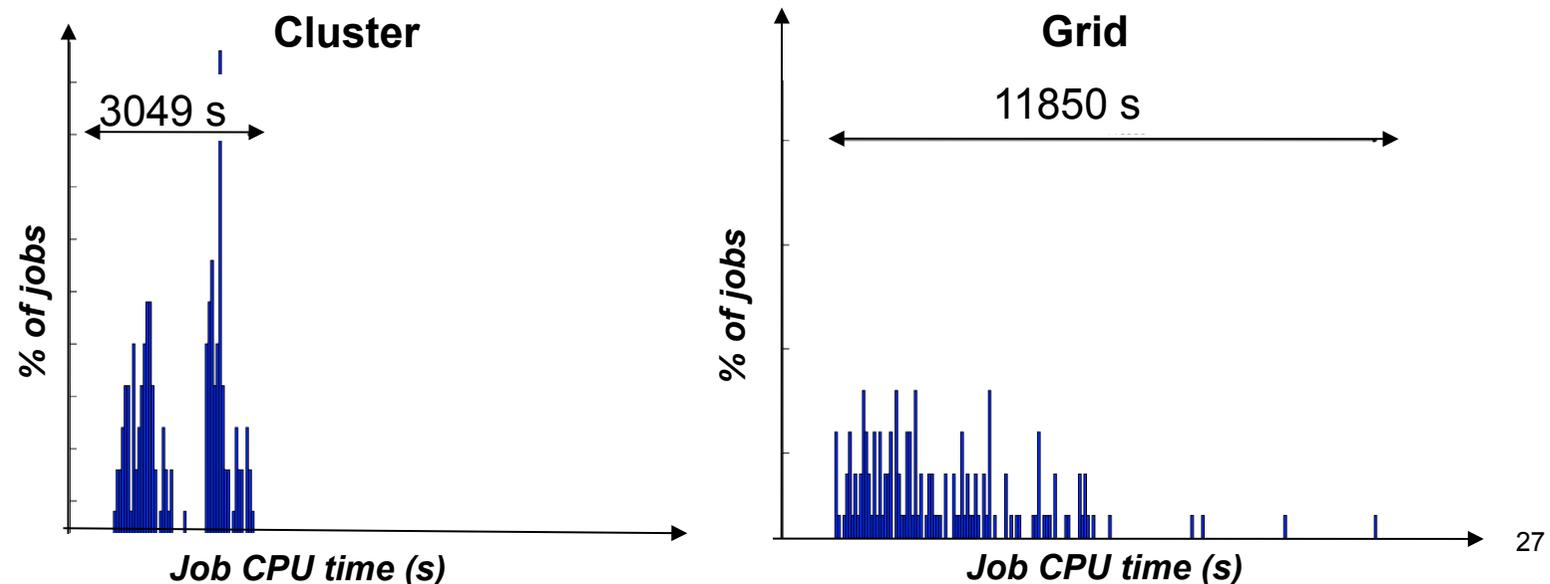
- Cannot download input (76.7%) ; cannot upload output (9.9%)
- Pilot/master communication problem (8.8%)
- Application error (4.8%)

- Time facts

- Total CPU time: 5.8 months
- Average IO / CPU: 7%

Comparison with cluster execution

- Local cluster
 - Six 8-core Intel Xeon L5430 2.66 Ghz with PBS
 - $(\text{average bogomips cluster}) / (\text{average bogomips grid}) = 1.14$
- CPU time comparison
 - $(\text{Total CPU time grid}) / (\text{Total CPU time cluster}) = 1.5$
 - Increased variability on grid (factor 4):



Conclusion (on Mean-Shift side)

- Mean-Shift results
 - Worth using bandwidth with 4 parameters (anisotropic bandwidth \mathbf{H})
 - Optimization with gradient ascent is possible
- Future work on Mean-shift
 - Validate Mean-Shift results on more images (including medical images)
 - Test bandwidth with 7 parameters
 - Try other quality measurement metrics (besides PSNR)

Conclusion (on grid side)

- Grid achievements
 - Significant speed-up (164 days => 2.5 days)
 - Autonomous usage from end-user
 - Some observations on scheduling and data transfers (see paper)
- Future work on the grid setup
 - Improve experiment planning on grids
 - Make data retrieval more robust

Credits



EGEE-III Life-Science cluster

Institut des Grilles du CNRS

Creatis

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MOTEUR

Johan Montagnat, Alban Gaignard ; CNRS I3S

VL-e medical software

Silvia D. Olabarriaga ; *AMC Amsterdam*

Piter T. de Boer ; Universiteit Van Amsterdam

Spiros Koulouzis ; Universiteit Van Amsterdam

Pilot jobs (DIANE)

Jakub T. Moscicki ; *CERN*

Grid support

China Scholarship Counsel (CSC)

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Thank You for your attention!