



DE LA RECHERCHE À L'INDUSTRIE

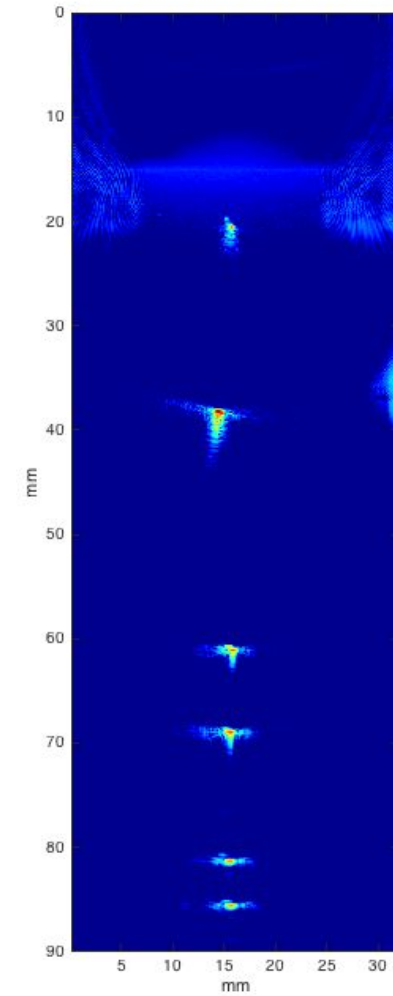
# Realize an adaptive sampling algorithm for TFM imaging on GPU

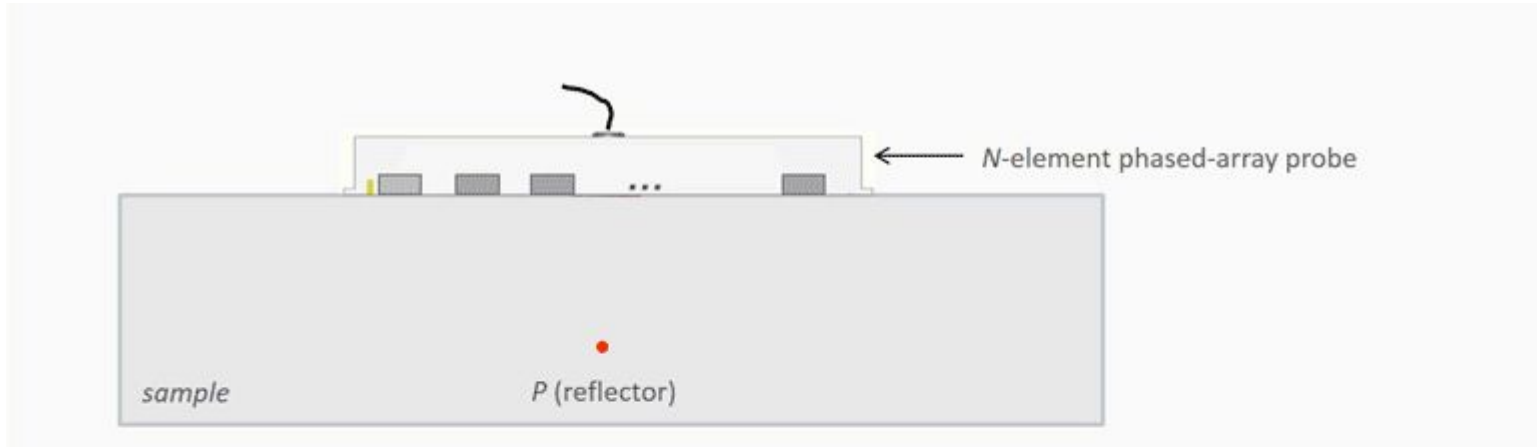
05 juillet 2019

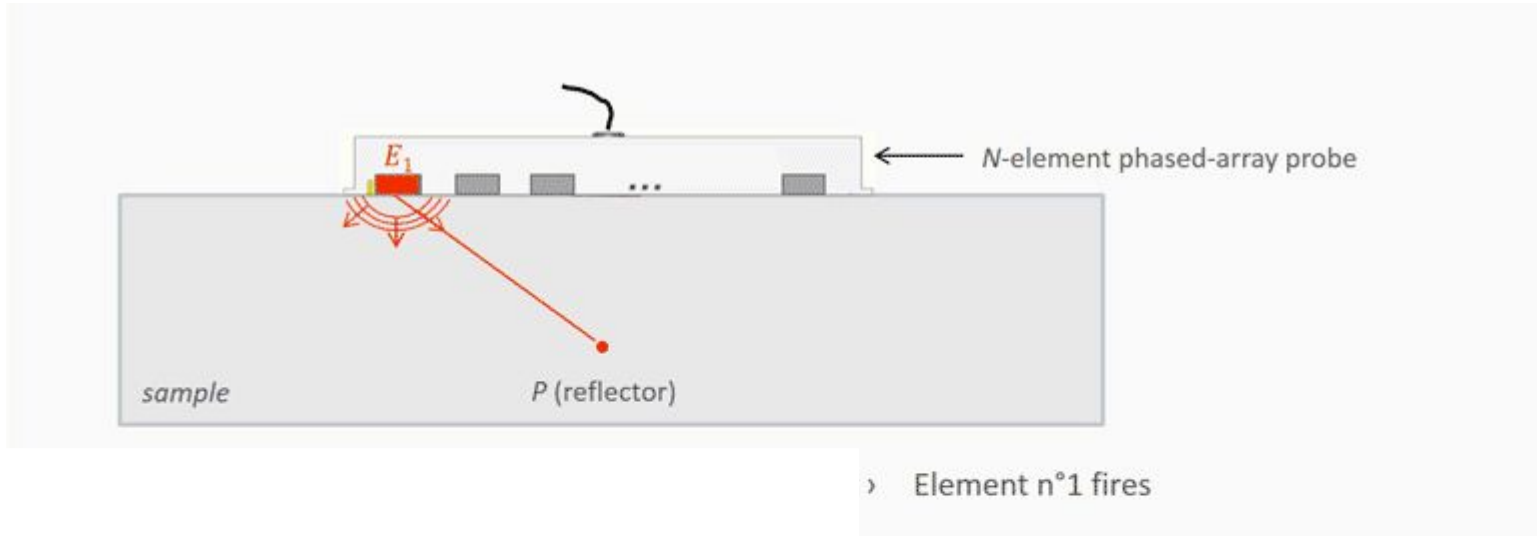
DAVID Jean-François

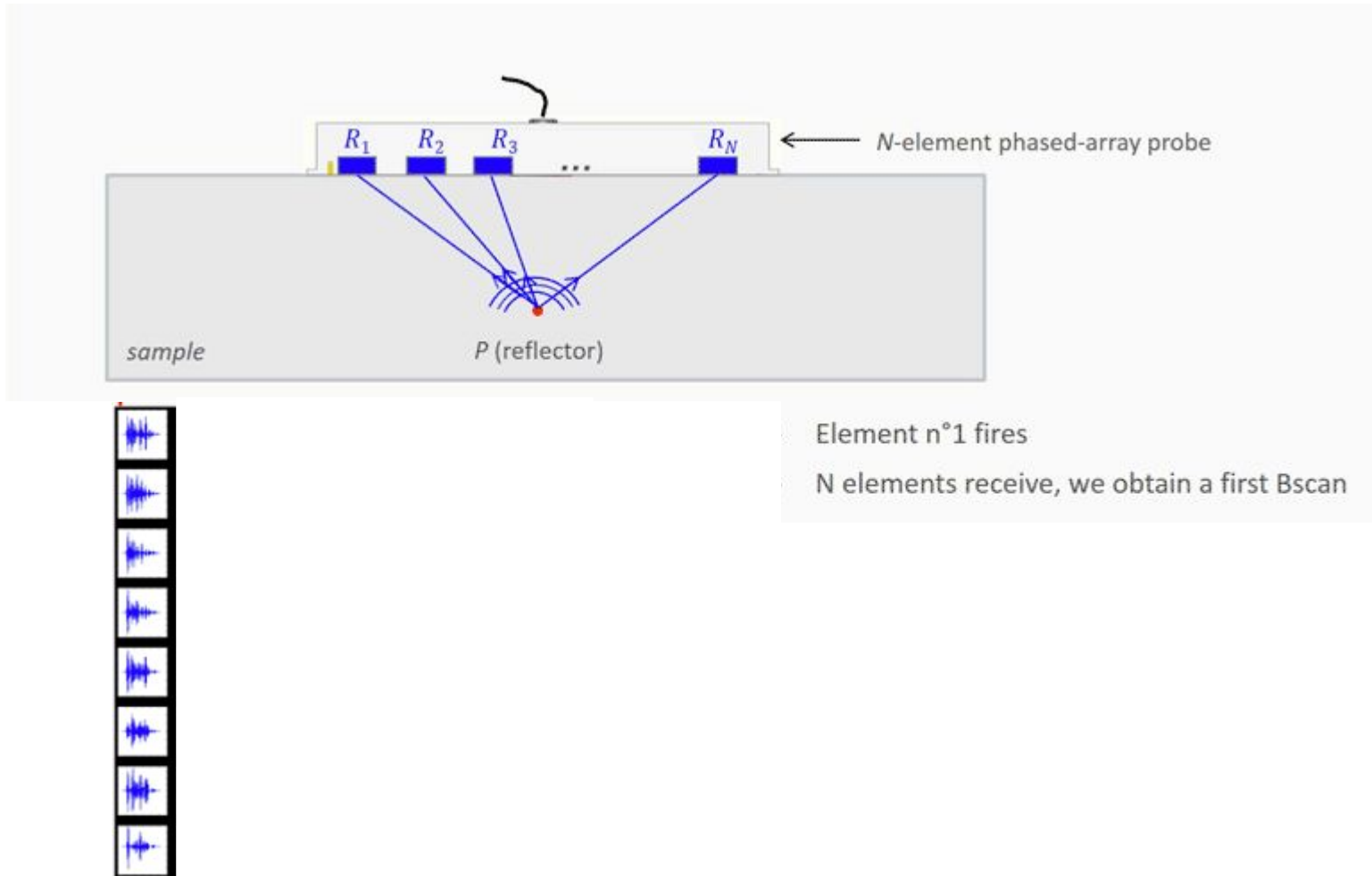
Supervisors :

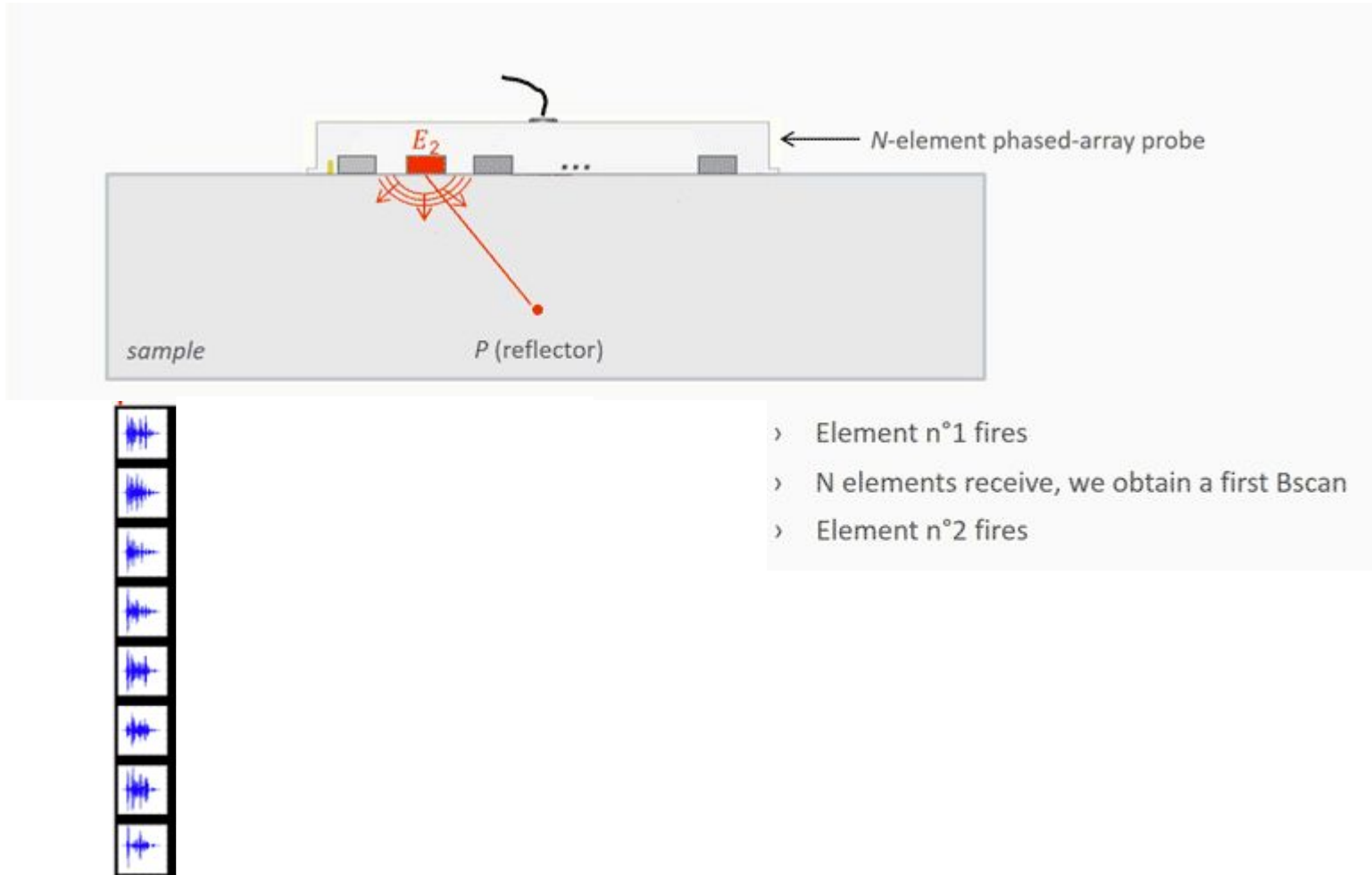
- ▶ M. CHOUEH Hamza
- ▶ M. BERGEAUD Vincent

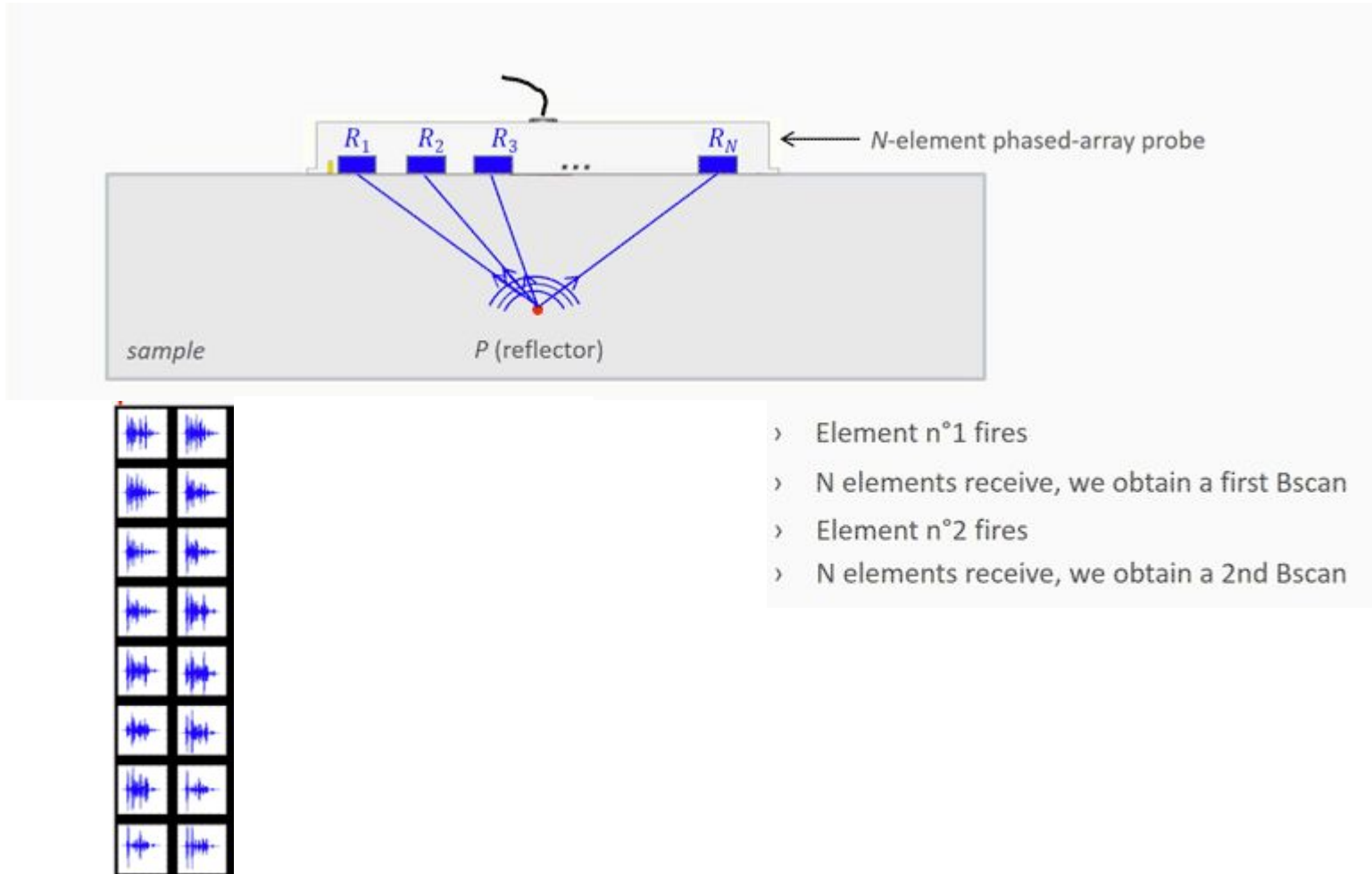


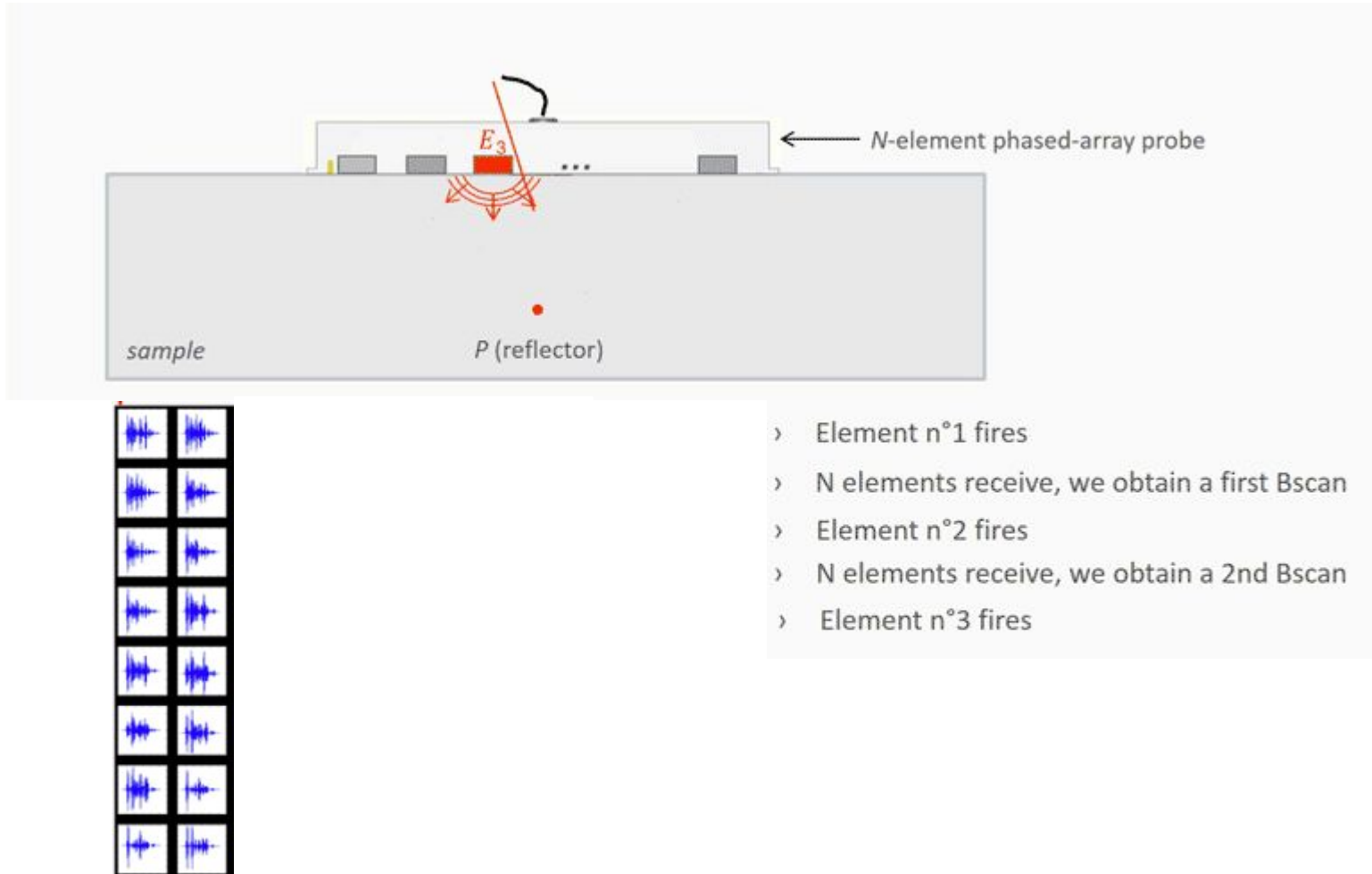




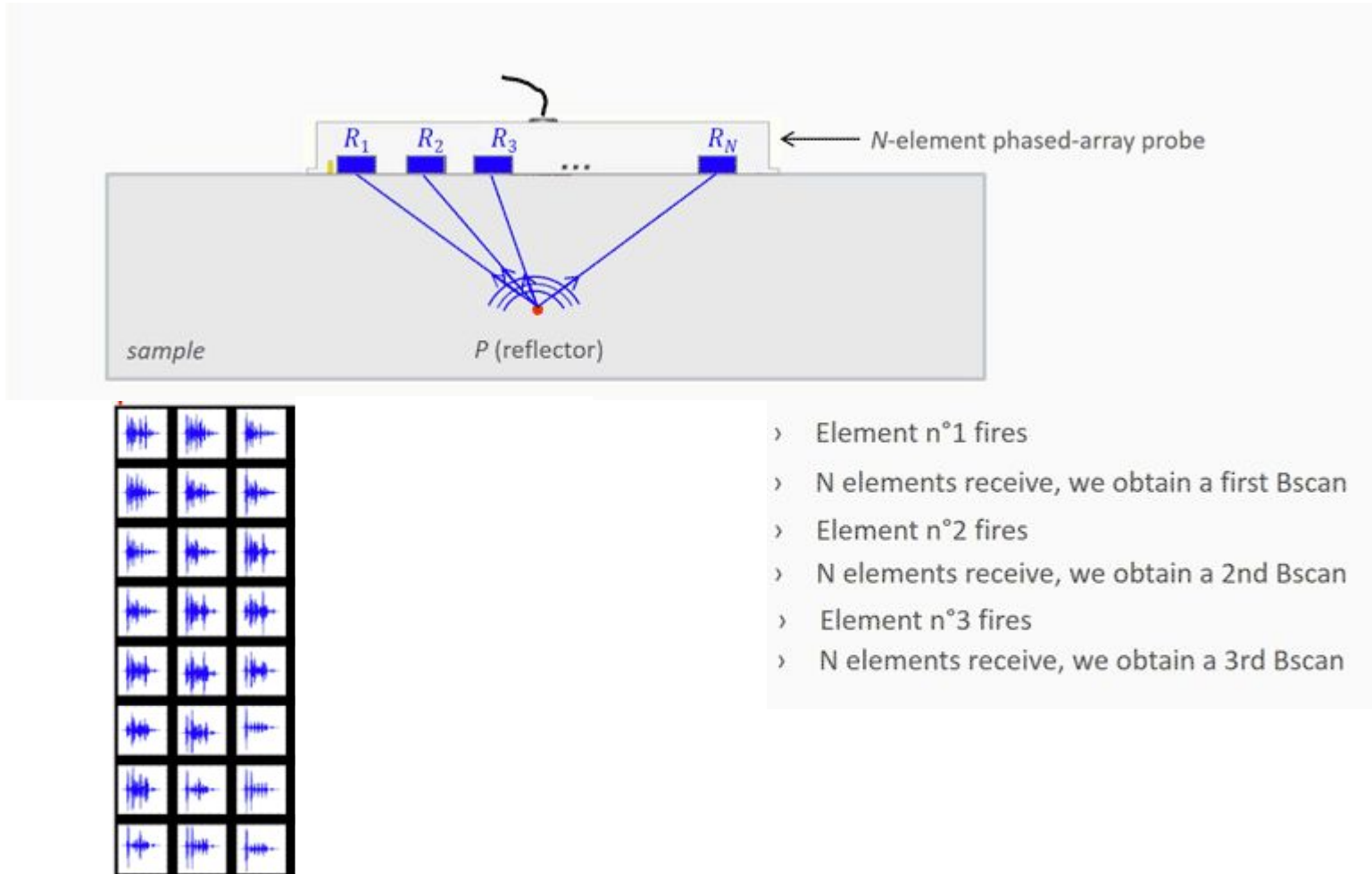


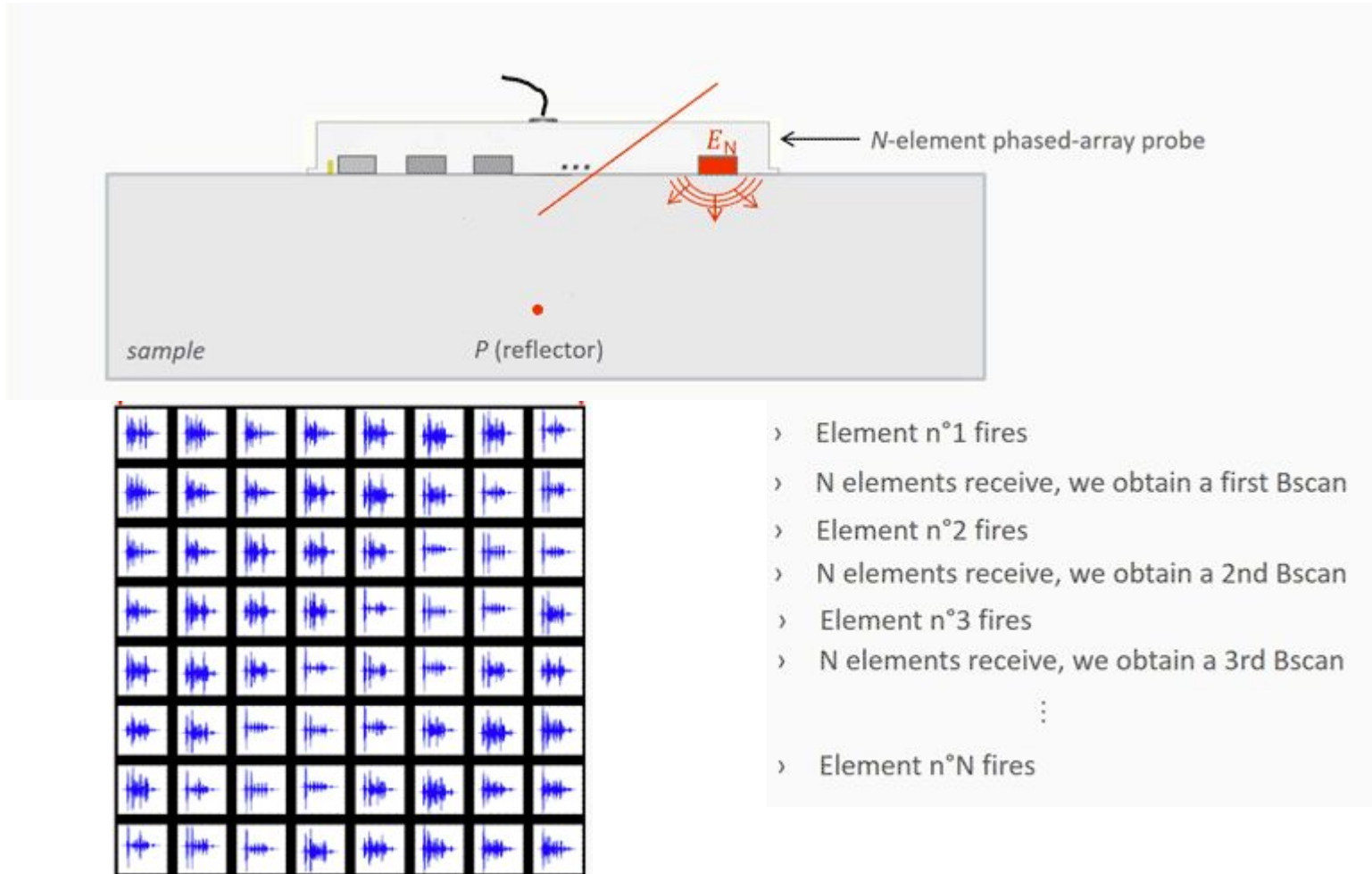


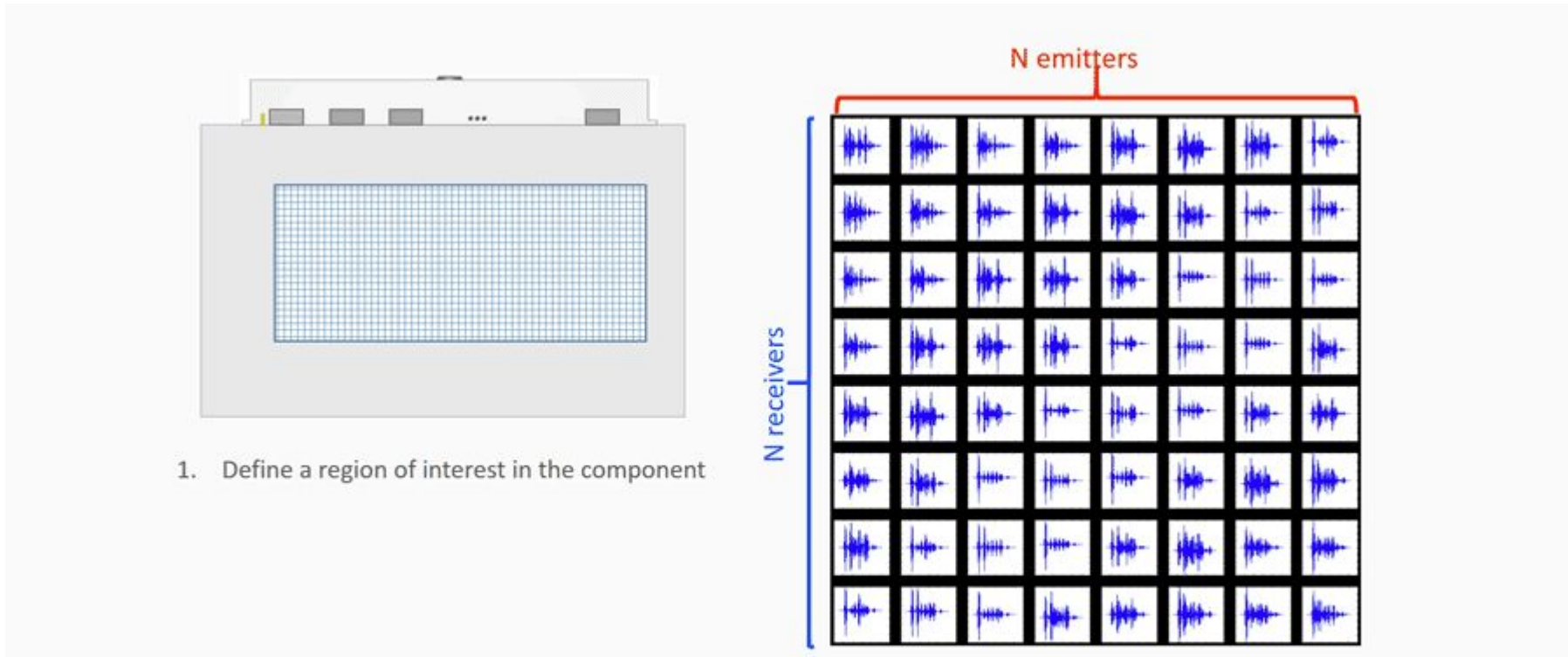


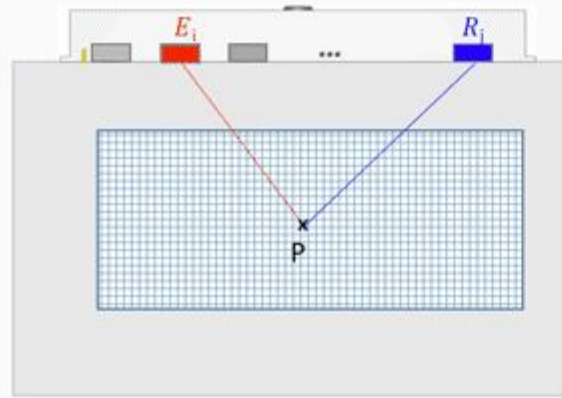




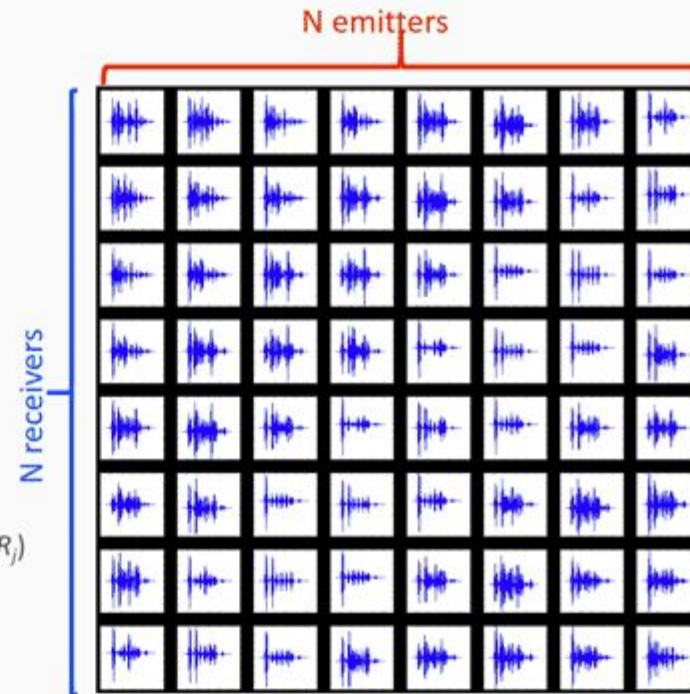


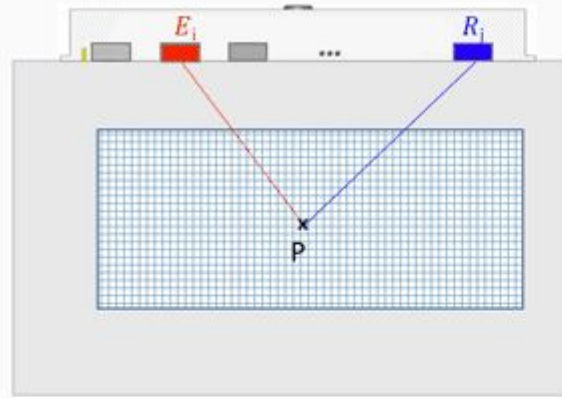




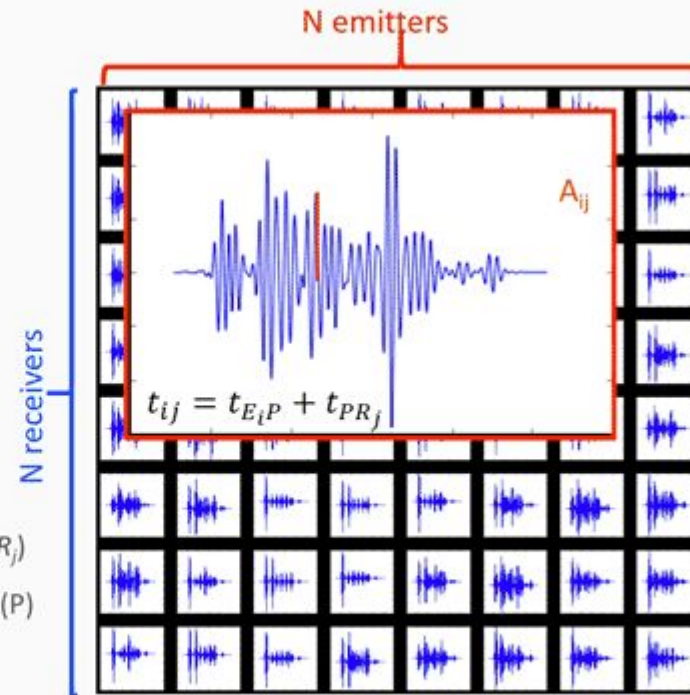


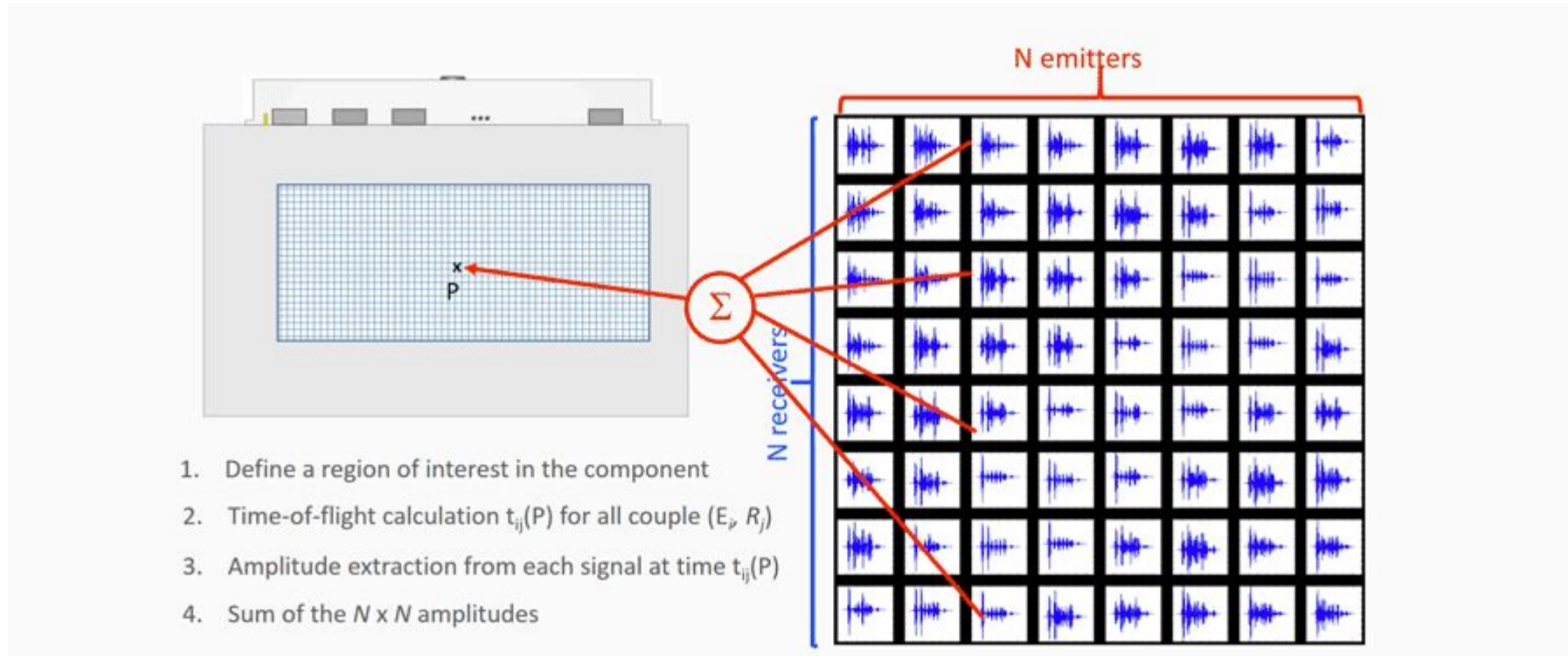
1. Define a region of interest in the component
2. Time-of-flight calculation  $t_{ij}(P)$  for all couple  $(E_i, R_j)$

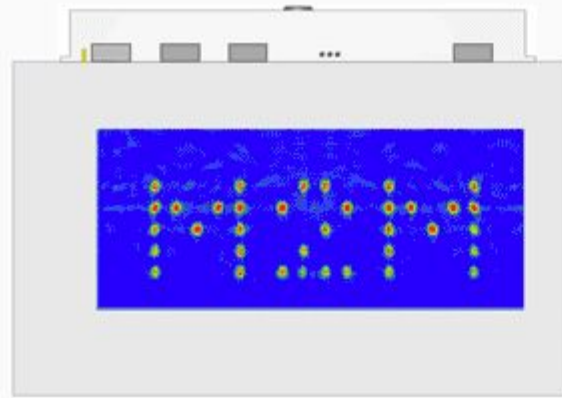




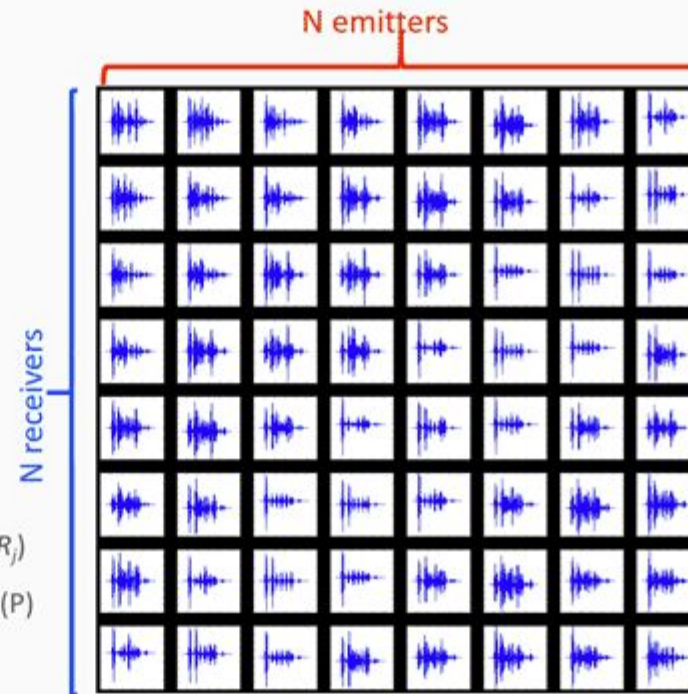
1. Define a region of interest in the component
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3. Amplitude extraction from each signal at time  $t_{ij}(P)$







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2. Time-of-flight calculation  $t_{ij}(P)$  for all couple  $(E_i, R_j)$
3. Amplitude extraction from each signal at time  $t_{ij}(P)$
4. Sum of the  $N \times N$  amplitudes
5. Perform the steps above for all pixels



## Context

- ▶ Work on the thesis of Mr. CHOUH Hamza
  - Approach on CPU (Quadtree and B-Splines)
  - Perspectives on GPU
  
- ▶ Computing complexity TFM: **PN<sup>2</sup>**
  - P: Number of pixels in the TFM image
  - N: Number of elements
  
- ▶ Computing constraints TFM:
  - Real time
  - On-board devices
  - On graphic processor (GPU)
  - Need accurate defect imaging



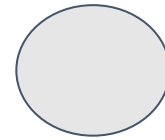
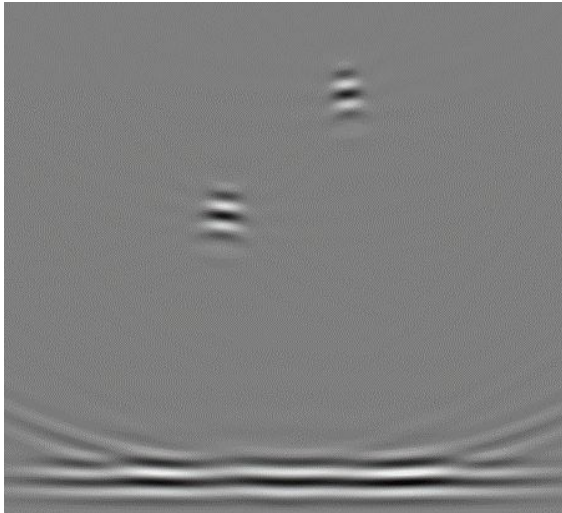
### The constraints :

- The SIMD architecture is not very conducive to algorithms with divergent instructions (if/else, while, ect)
- No dynamic allocation, no queue, stack, etc
- Problems of concurrency, writing and reading
- Limited local memory size for threads, which greatly affects computing speed
- Dynamic and recursive programming possible but very slow

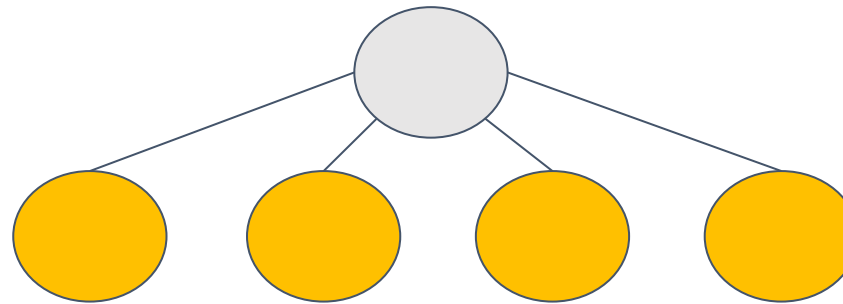
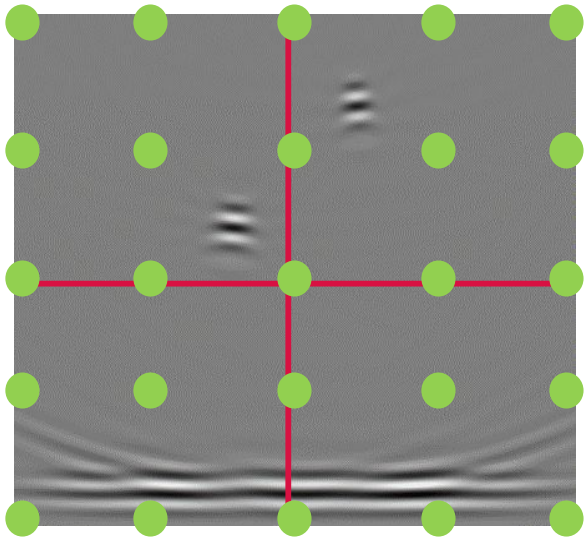
### The challenge :

- Develop an adaptive algorithm on GPU
- Find a reconstruction method for non-regular sampling

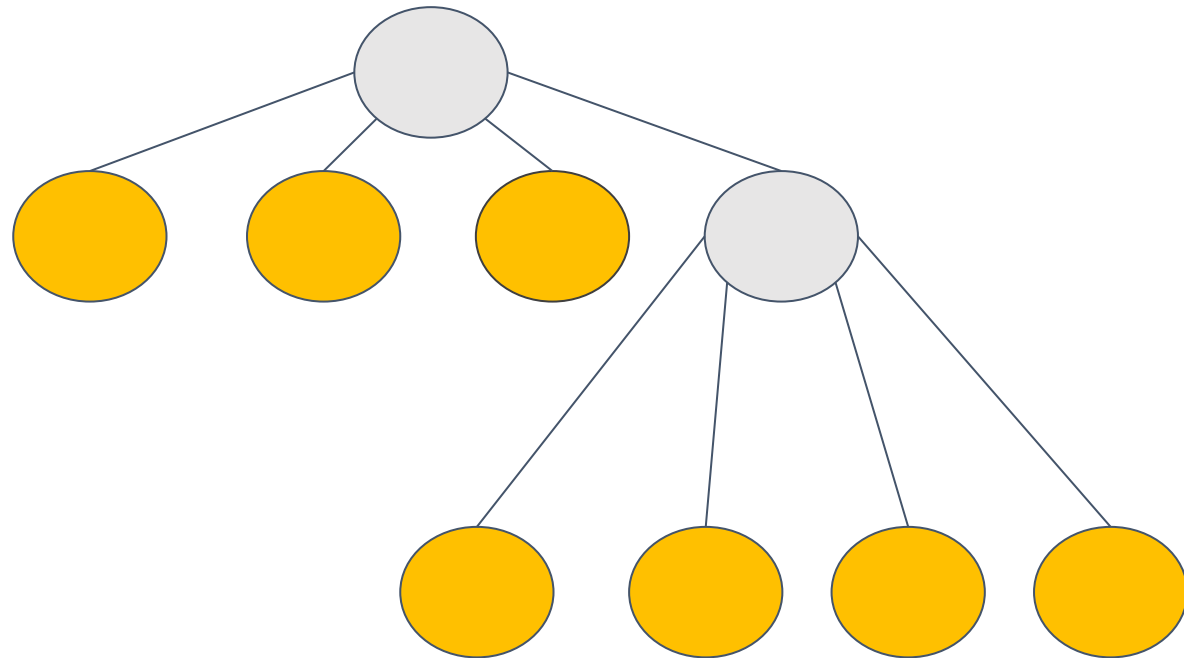
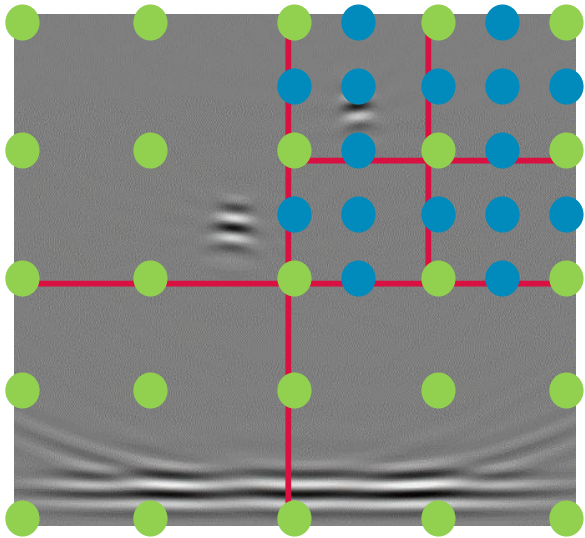
- Is a tree-like data structure in which each node has four children.
- It's often used to partition a two-dimensional space.



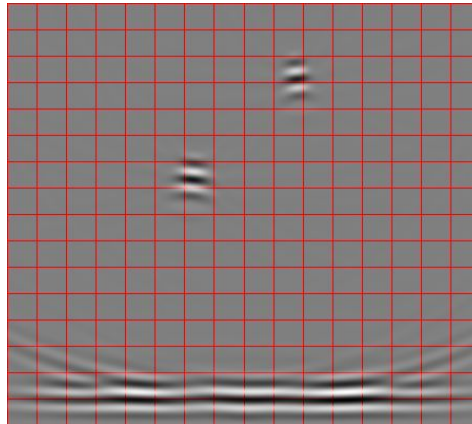
- Is a tree-like data structure in which each node has four children.
- It's often used to partition a two-dimensional space.



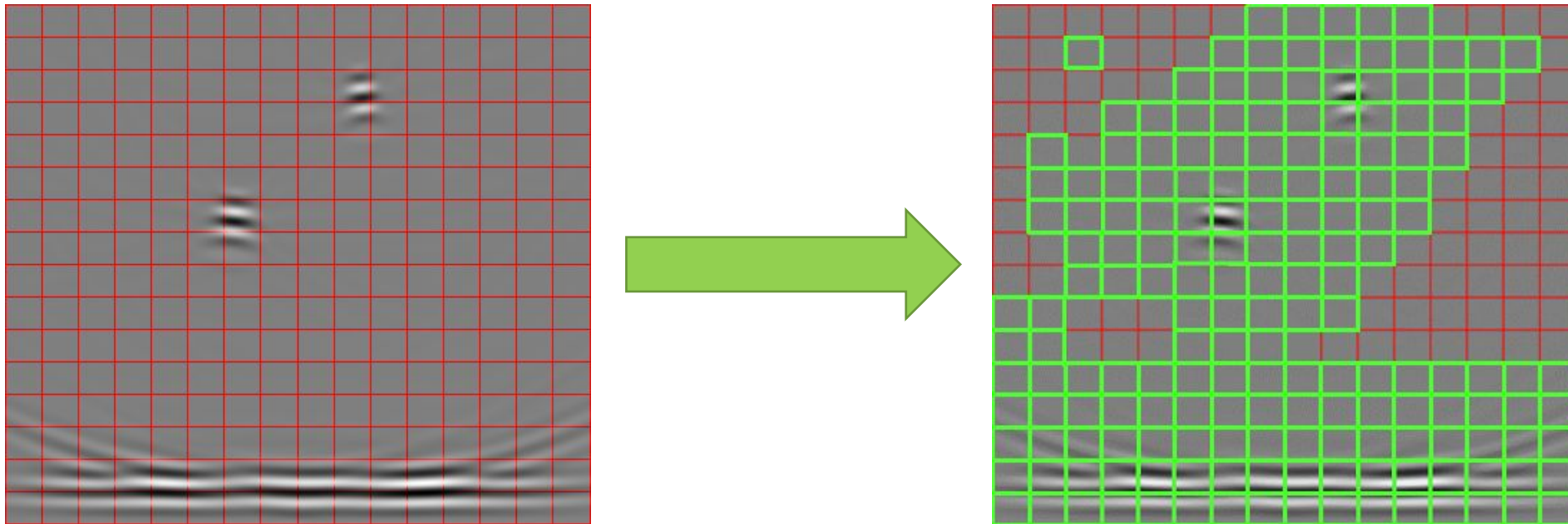
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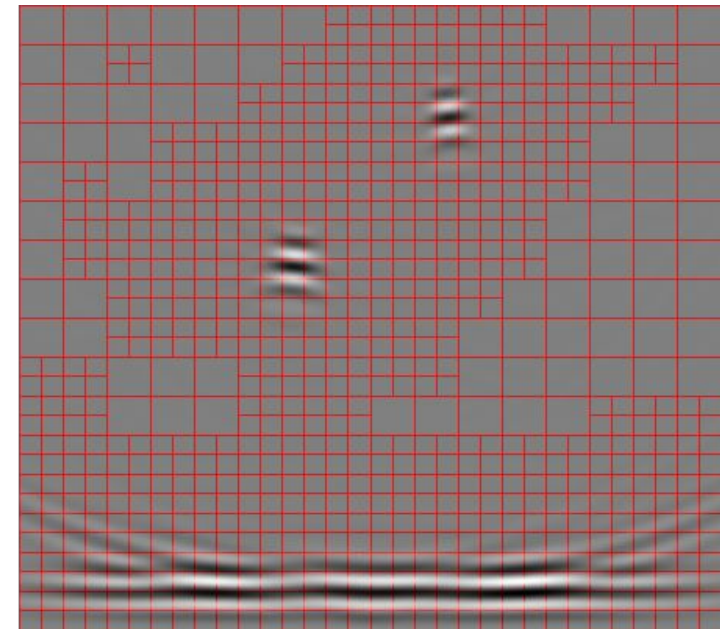
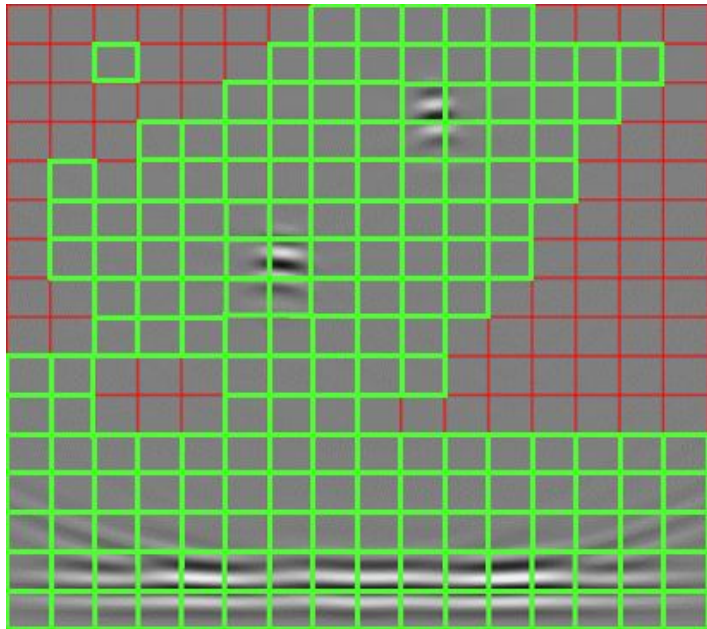


- The array containing the future nodes is pre-allocated, as well as an array of booleans allowing to know the TFM points that have been calculated.
- The root node is initialized, then it is subdivided, as well as its children, until it reaches a certain depth to load the GPU as much as possible.

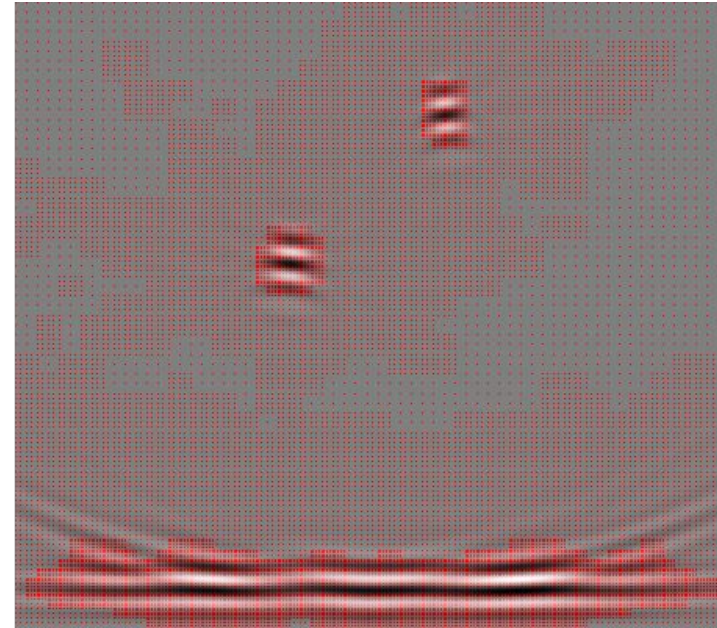
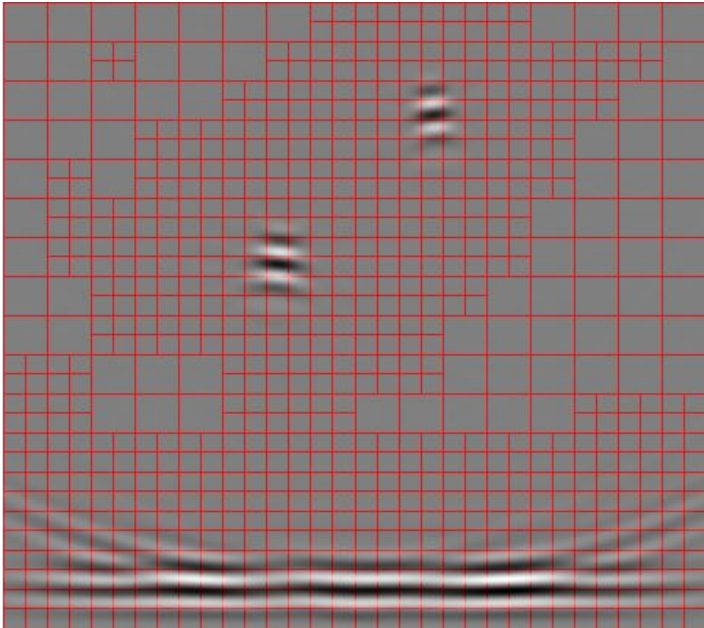


- The score of each node is calculated according to its best amplitude and depth in the quadtree.
- The nodes are sorted by their score in a table.
- A fixed number of nodes with the highest score are selected from the table. These are the nodes that will be subdivided.



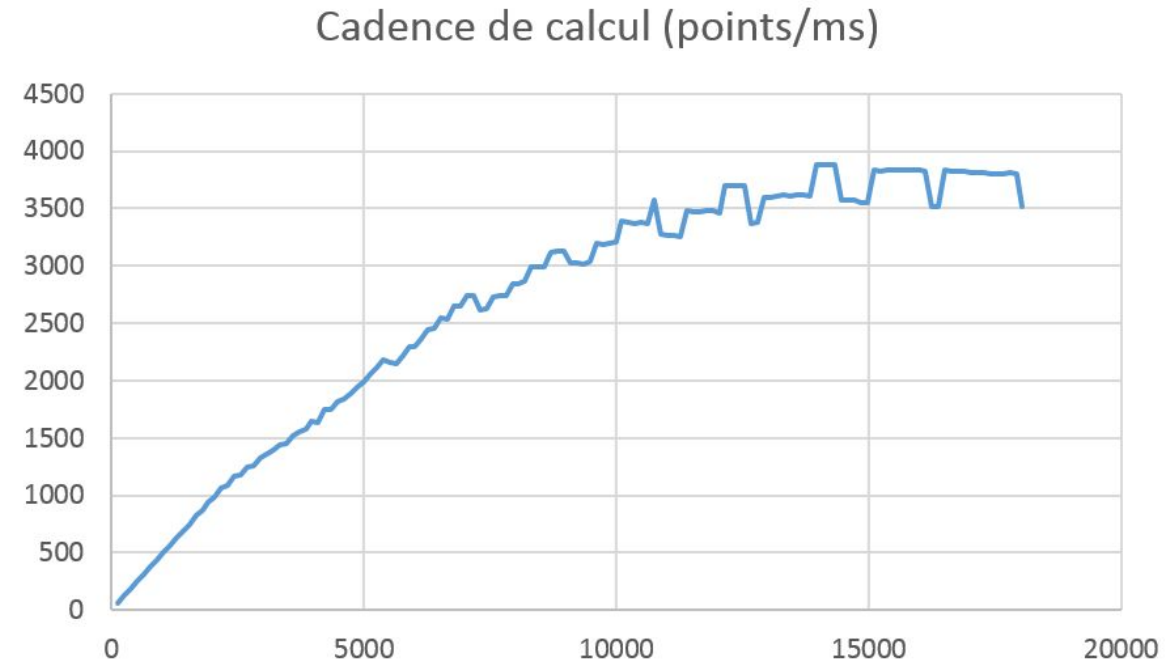


- The amplitude value of the TFM is calculated for the pixels corresponding to each of the cardinal points of the nodes.

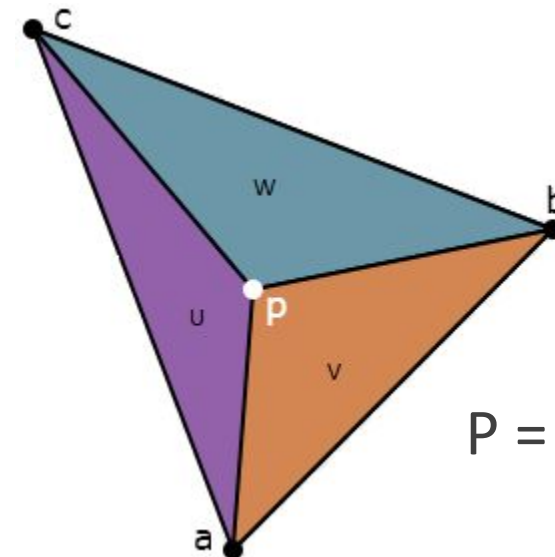
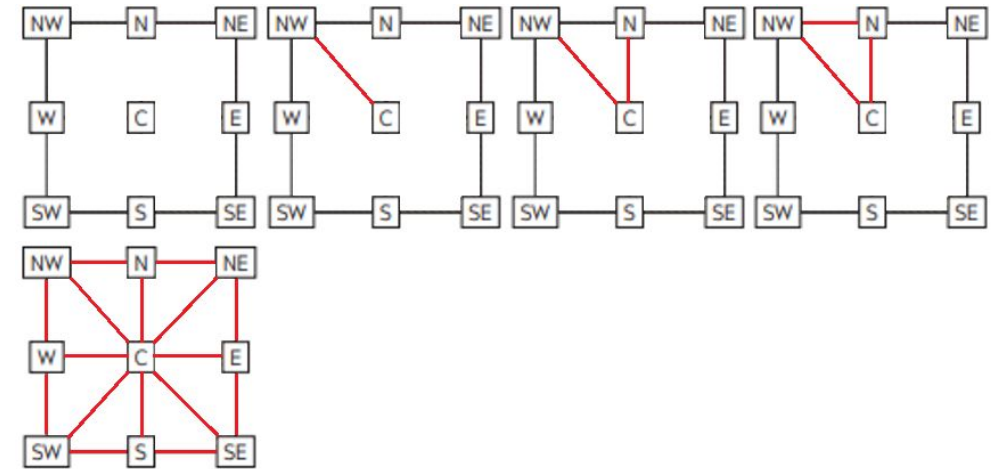




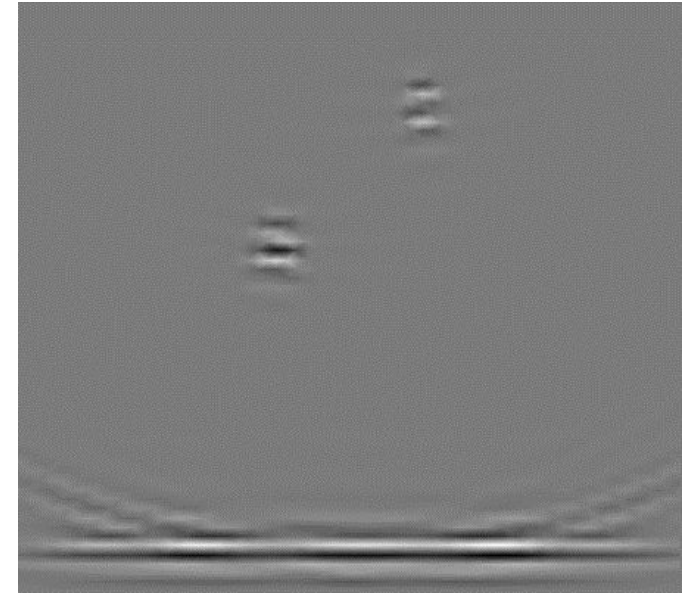
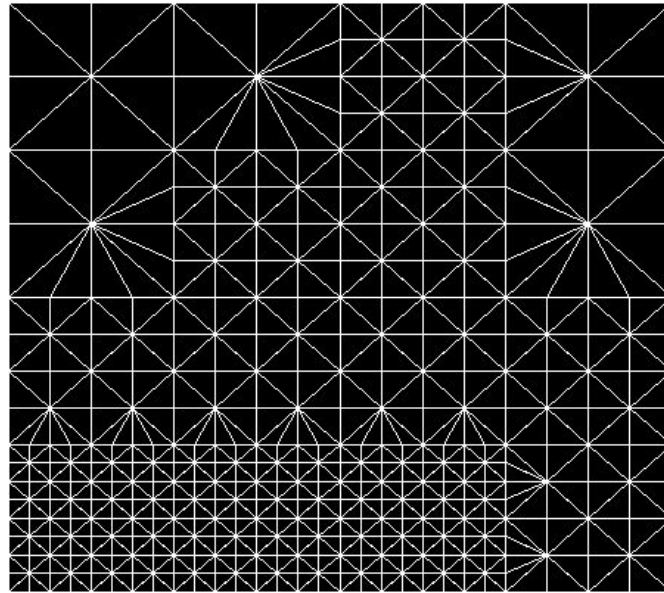
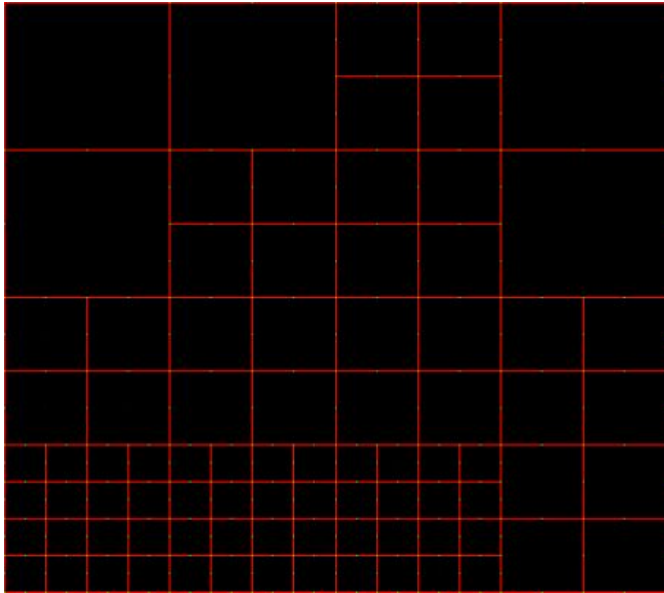
- The user sets a target frame rate (e.g. 50 fps)
- We proceed to a first estimation of the calculation rate which allows us to determine the "budget" of points to calculate per image
  - Measurement of the computation time for a group of pixels at each iteration
  - We give a maximum time for the calculation of an image
- We look for the lowest possible granularity value that maximizes the rate.
  - It must load the GPU
  - Must be located on a peak rate



- The reconstruction of the image is done in two phases: triangulation (mesh of the irregular grid of pixels) and interpolation
- The triangulation is done on each leaf node of the quadtree, each vertex of the triangles corresponds to a TFM point calculated beforehand
- The interpolation used is a barycentric interpolation



$$P = aw + bu + cv$$

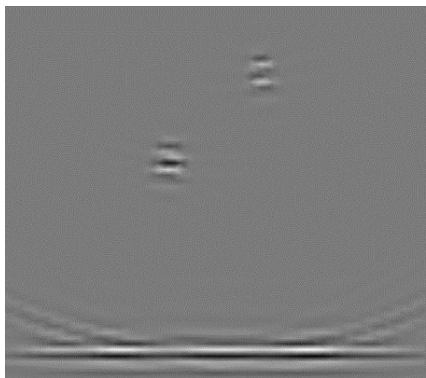


# Adaptive refinement

image size 451 x 401 = 180 851 pixels

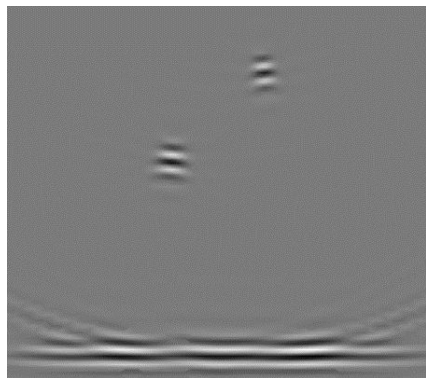
3 002 samples

8 ms



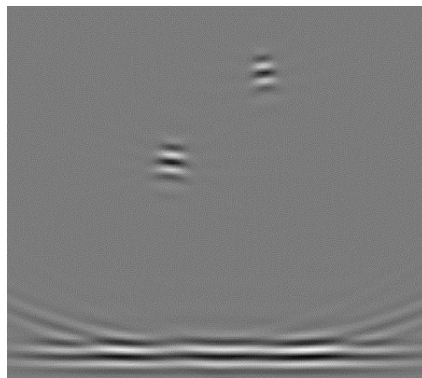
5 968 samples

10 ms



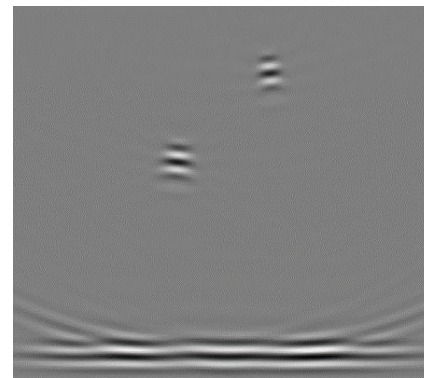
13 393 samples

18 ms



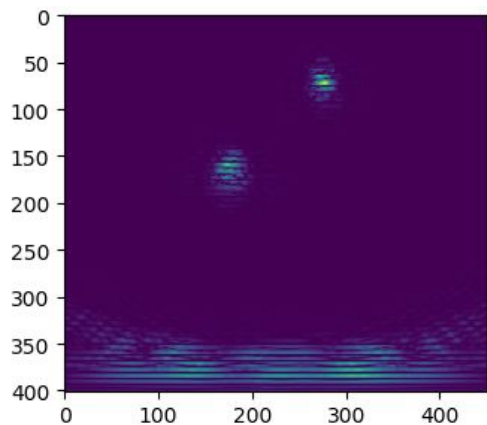
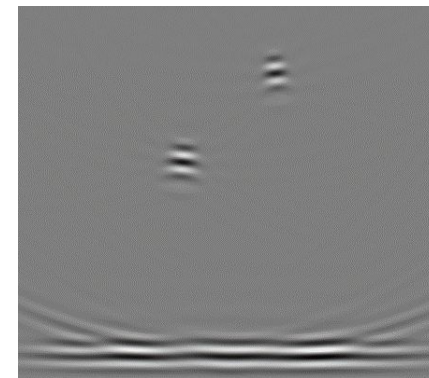
19 314 samples

22 ms

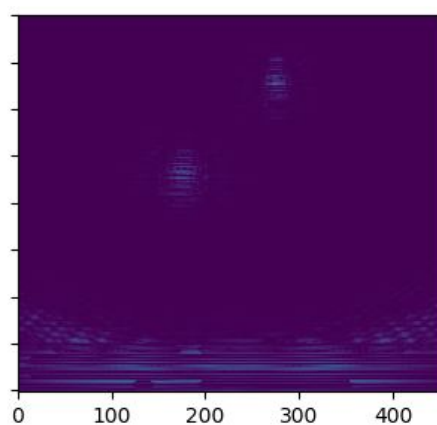


Idéale

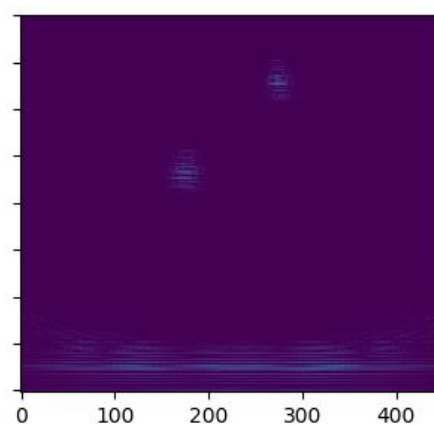
70 ms (without spl)



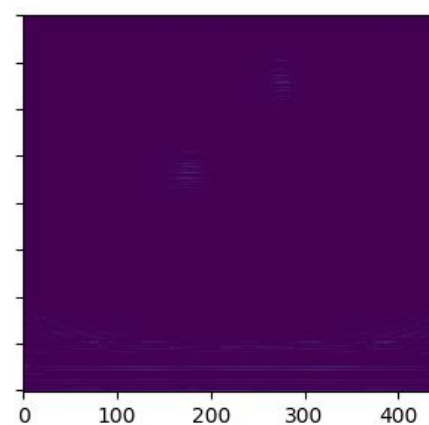
Error max : 65%



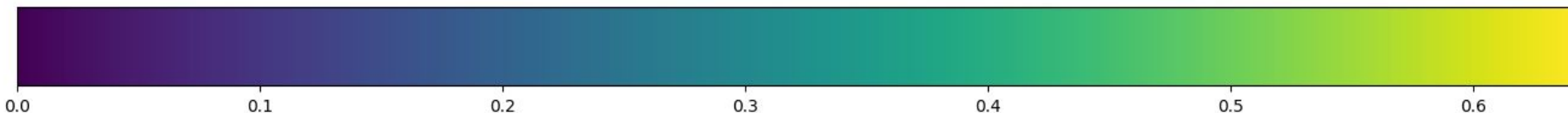
Error max : 46%



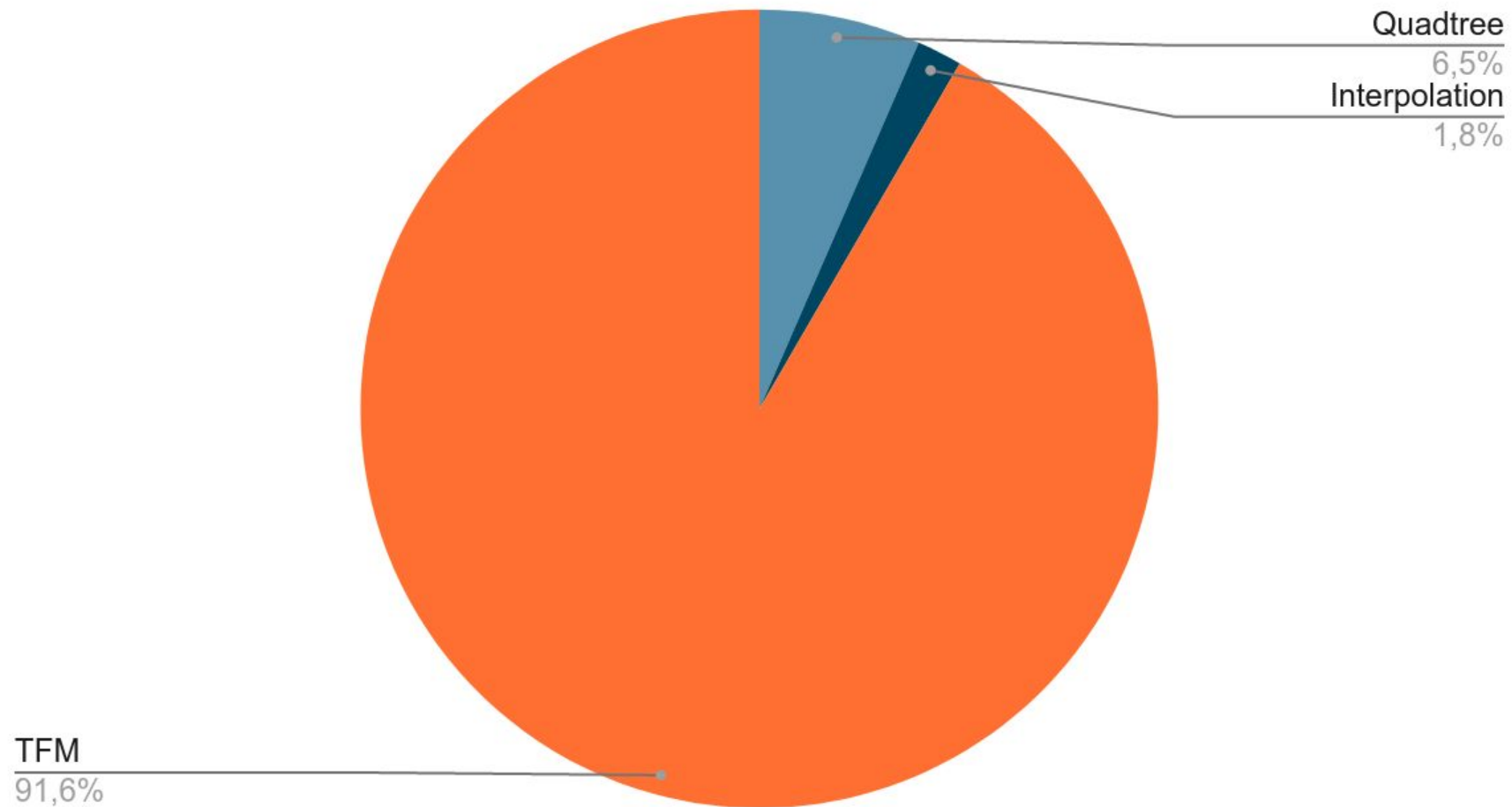
Error max : 24%



Error max : 8%



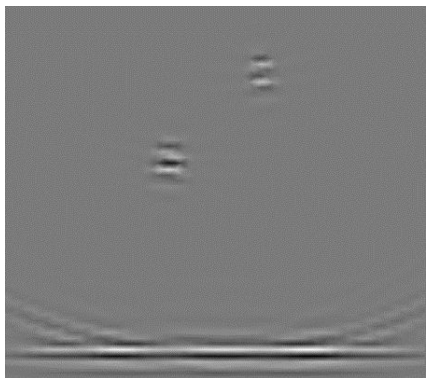
## Computing part for 11% of recovery



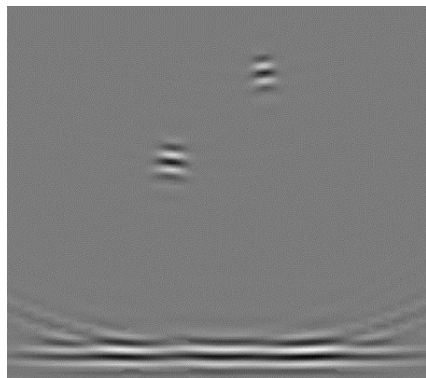
# Raffinement adaptatif

image de taille 451 x 401 = 180 851 pixels

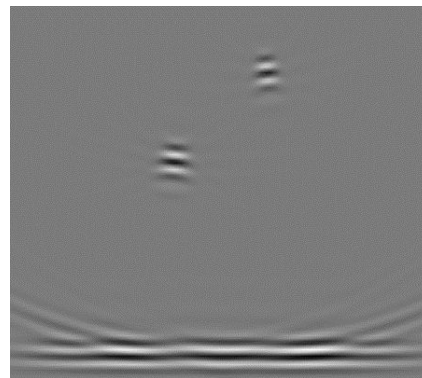
3 002 samples  
10 ms



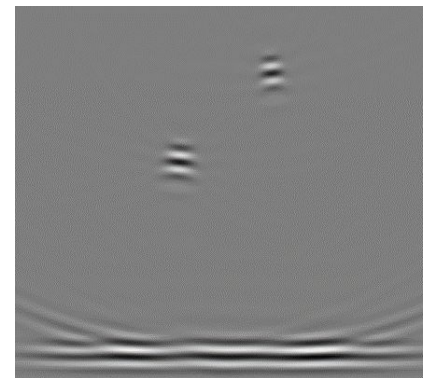
5 968 samples  
14 ms



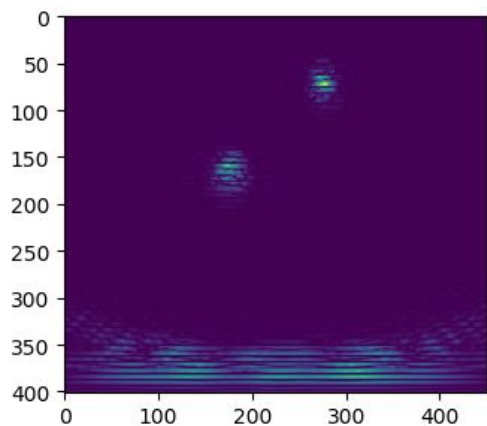
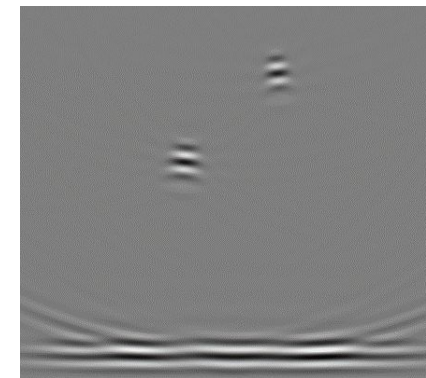
13 393 samples  
18 ms



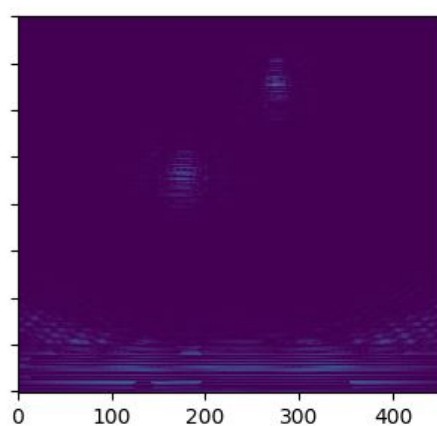
19 314 samples  
22 ms



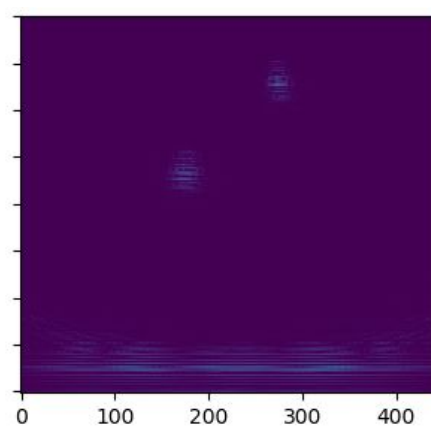
Idéale  
70 ms



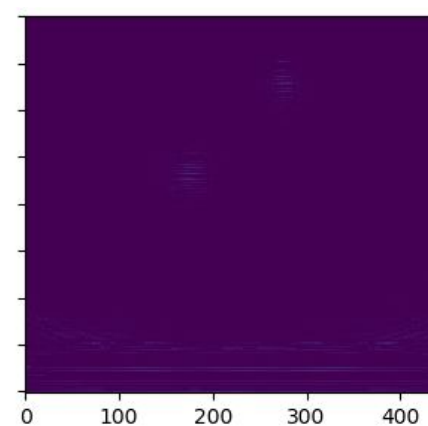
Error max : 65%



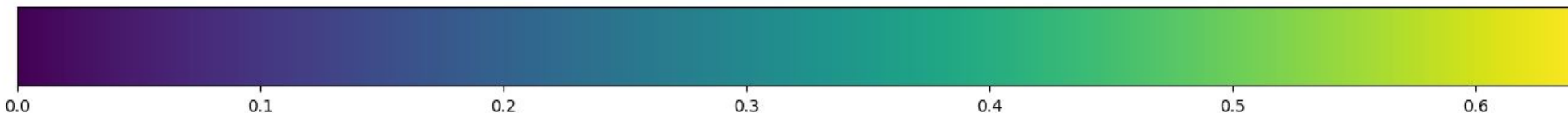
Error max : 46%



Error max : 24%

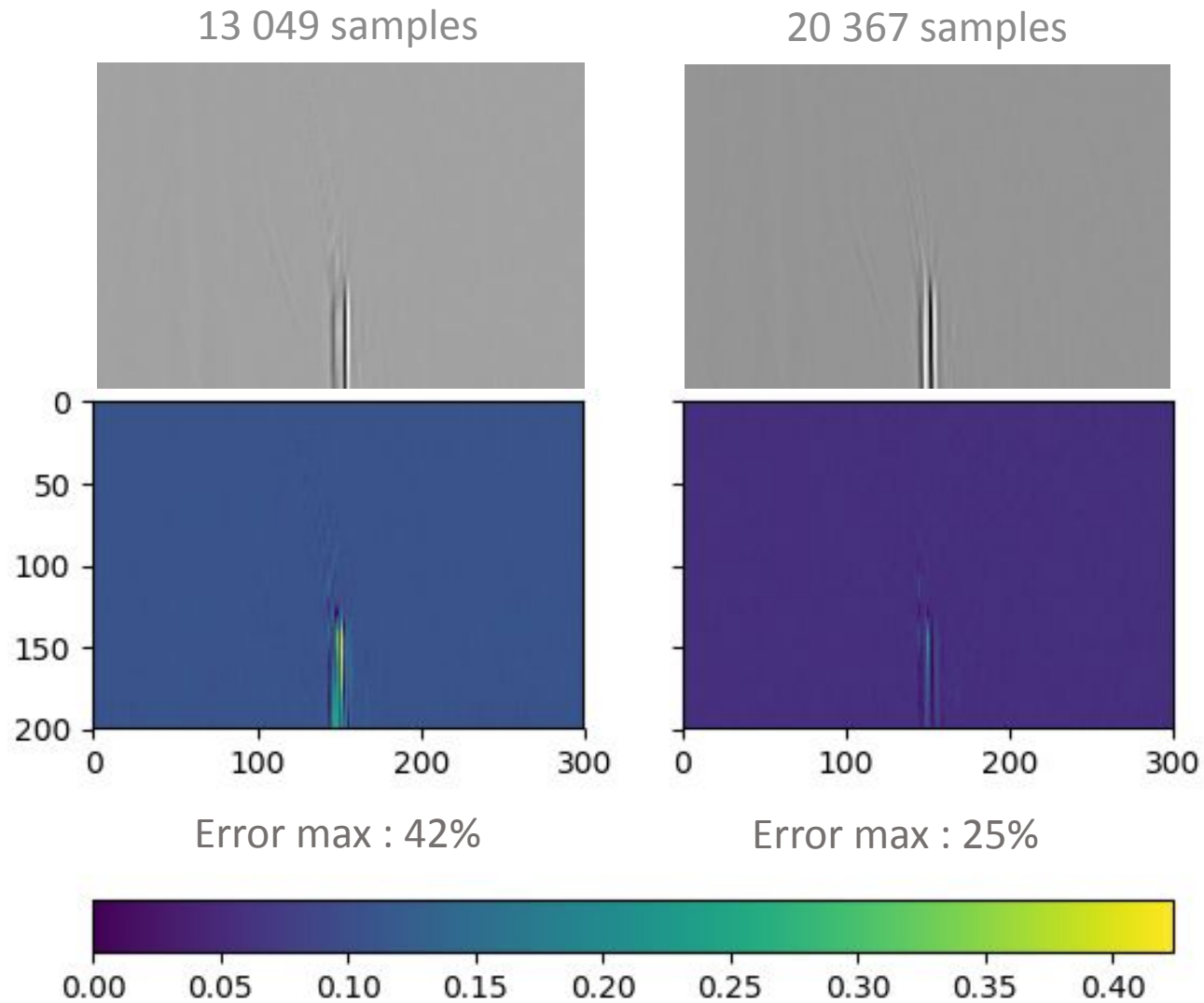


Error max : 8%



# Other exemple

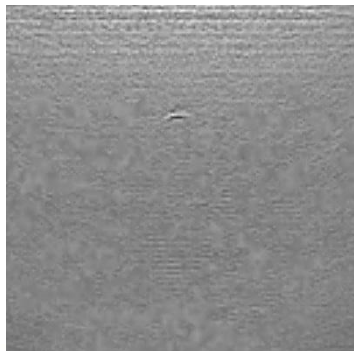
image size 301 x 201 = 60 501 pixels



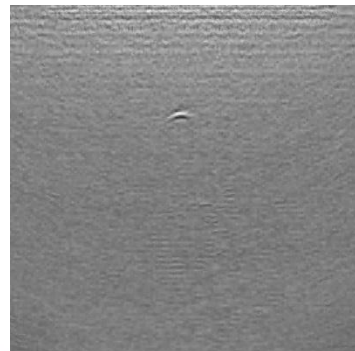
# Other exemple

image size 401 x 401 = 160 801 pixels

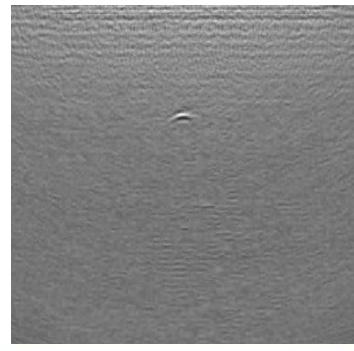
13 335 samples



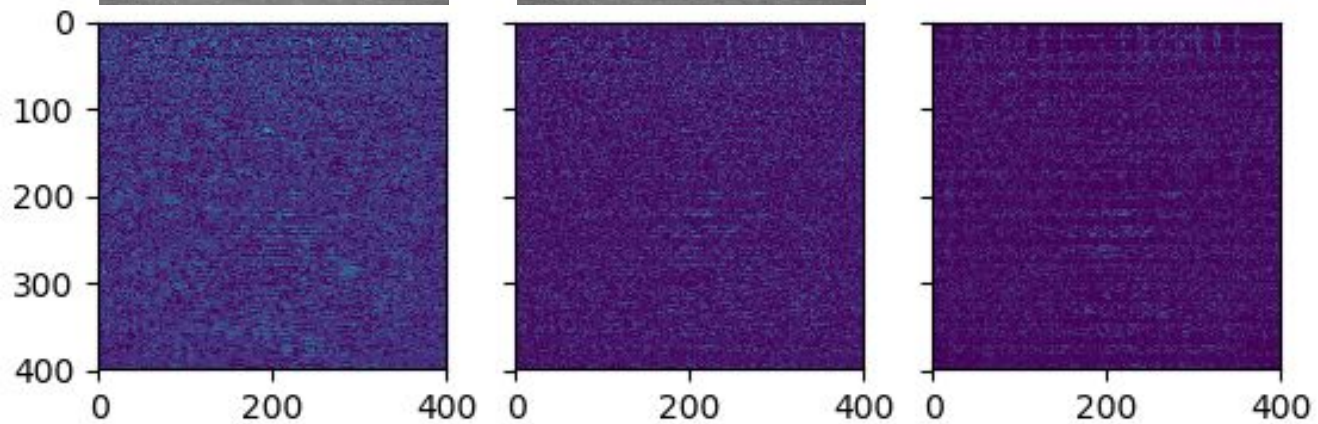
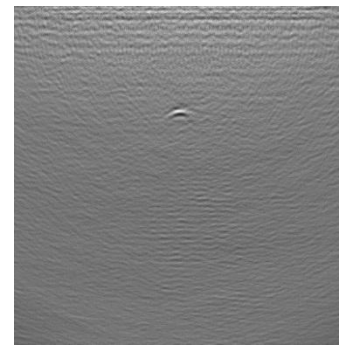
21 197 samples



34 811 samples



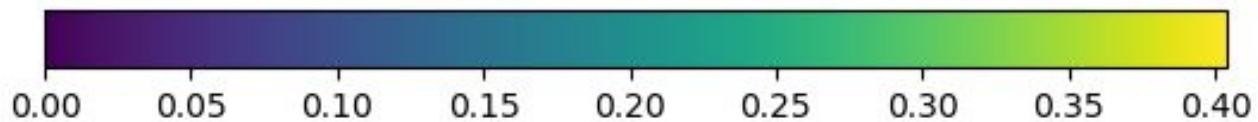
Idéale



Error max : 45%

Error max : 25%

Error max : 22%





- Implementation of an adaptive sampling algorithm, and TFM imaging reconstruction on GPU
- For a small number of samples, correct images are reconstructed
- The construction of the tree and its refinement represent a negligible additional cost compared to TFM

- 1 Pour chaque element en emission  $iE$
- 2     Pour chaque element en reception  $iR$
- 3         Pour chaque pixel de l'image  $P$
- 4             Ajouter la contribution du tir  $(iE, iR)$  sur le pixel  $P$