

ViSP 2.6.0: Visual servoing platform

# ViSP tracking methods overview

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lagadic

The INRIA logo, consisting of a stylized white 'R' inside a square, followed by the letters "INRIA" in a white, sans-serif font.

# Tracking methods with ViSP

1. Dot tracker
2. KLT point tracker
3. Moving edges tracker
4. 3D model-based tracker



# 1. Dot tracker

A dot :

- A dot is a part of image where the connected pixels have the same level
- Not necessary an ellipsoid (even it is by default)
- Two classes vpDot and vpDot2

The dot in ViSP is defined by :

- The gray level
- The center of gravity (cog)
- The size
- The moments :
  - The surface  $m_{00}$
  - Inertia first order moments along i and j  $m_{01}$  and  $m_{10}$
  - Inertia first second moments along i and j  $m_{02}$  and  $m_{20}$
  - $m_{11}$



# 1. Dot tracker with vpDot class

Tracking method : vpDot

- Initialization : Define the dot cog (generally by clicking in the dot)
- Tracking :
  - Binarisation of the image
  - Recursive method to detect all the neighbour components belonging to the object. Start from the previous coordinates of the center of gravity
  - If the dot is found : Compute the parameters (size, moment, ...)
  - If no dot is found : The tracking fails



# 1. Dot tracker with vpDot2 class

Tracking method : vpDot2

- Initialization : Define the dot cog (generally by clicking in the dot)
- Tracking :
  - Binarisation of the image
  - From the previous position of the cog, goes right to detect the boundary, then follow the boundary in order to compute the Freeman chain
  - Use the Freeman Chain to find the dot characteristics (cog, size, moments)
  - If a dot is found, check if it looks like the previous dot (size, moment)
  - If no dot or not similar, check if the dot is in an image part around



# 1. Dot Tracker

## Advantages :

- Robust : Almost no tracking error if noise and specularities not too strong
- Give information about the tracked objects (cog, moments)
- In vpDot2 : automatic dot detection for initialization and if a dot is lost search a similar dot in a larger ROI

## Limits :

- Speed depends on the size : may be slow if the object is big, especially with vpDot.
- vpDot can not track an object if the displacement is too large
- Due to the recursivity limitation on some OS like windows, vpDot is not able to track huge dots.



## 2. KLT point tracker

KLT : Kanade – Lucas -Tomasi

- The goal is to align a template  $T(x)$  to an input image  $I(x)$
- Could be also a small window in the image
- Based on a gradient method

KLT in ViSP :

- `vpKltOpencv` class that interfaces the KLT implemented in OpenCV
- A patch is defined by :
  - the tracked points in the current image
  - the tracked points in the previous image
- The points lost during the tracking are given if necessary



## 2. KLT point tracker

Tracking method :

- The goal is to move the patch until minimizing the image dissimilarity

$$\sum_x [I(W(x, p) - T(x))]^2$$

- Where  $W(x, p)$  corresponds to a warp which can be more or less complex

For a translation :  $W(x, p) = \begin{bmatrix} x + p_1 \\ y + p_2 \end{bmatrix}$

- Assuming that  $p$  is known and best increment  $\Delta p$  is sought, the problem becomes :

$$\sum_x [I(W(x, p + \Delta p) - T(x))]^2$$





## 2. KLT point tracker

Tracking method :

- The problem is linearized by performing first order Taylor extension

$$\sum_x \left[ I(W(x, p + \Delta p)) + \nabla I \frac{\partial W}{\partial p} \Delta p - T(x) \right]^2$$

- The function is derived and set equal to 0 to find the minimum

$$\Delta p = H^{-1} \left[ \nabla I \frac{\partial W}{\partial p} \right]^T [T(x) - I(W(x, p))] \quad \text{where } H = \sum_x \left[ \nabla I \frac{\partial W}{\partial p} \right]^T \left[ \nabla I \frac{\partial W}{\partial p} \right]$$

- $p$  is updated with this method  $p = p + \Delta p$  until  $\Delta p < \epsilon$



## 2. KLT point tracker

Good features :

- A good point to track :
  - Textured
  - High intensity variations in both x and y axis
- Harris points are used

Advantages :

- Very fast method
- In ViSP, Harris points detection is automatic

Limits :

- Displacement between two images must be small
- In ViSP : use `IpImage` instead of `vplImage` : need conversion
- Few parameters are available.



### 3. Moving edges tracker

Moving edges :

- Based on edge detector with gradient filter
- 3 types : line, ellipse and nurbs

In ViSP:

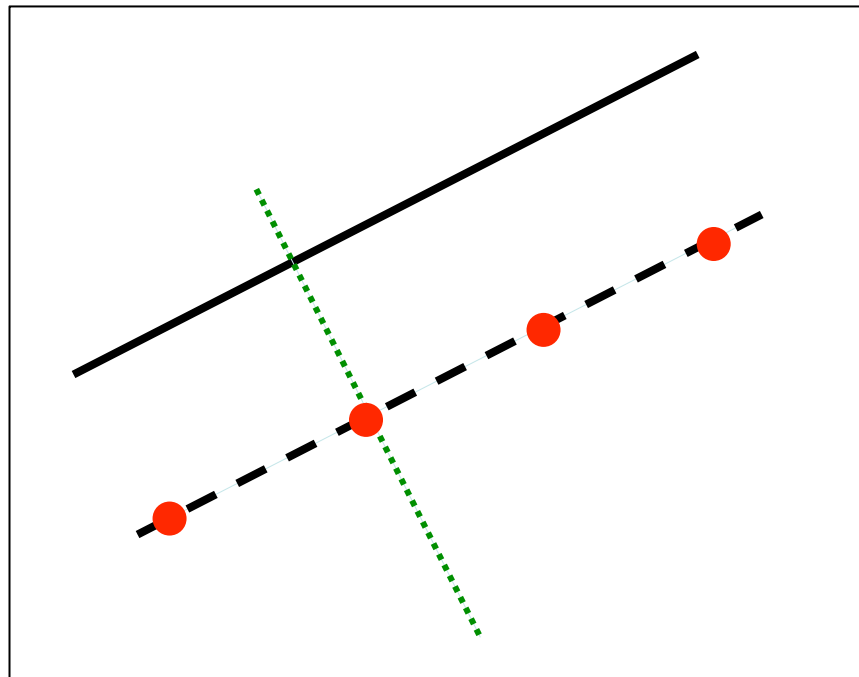
- vpMeLine, vpMeEllipse, vpMeNurbs classes that inherit from vpMeTracker
- vpMeTracker contains a list of vpMeSite
- Each vpMeSite corresponds to one edge point in the image.
- vpMeSite is defined by :
  - A position (i,j)
  - An angle which corresponds to the normal to the edge
  - An history of the previous convolution result



### 3. Moving edges tracker for a line

Method:

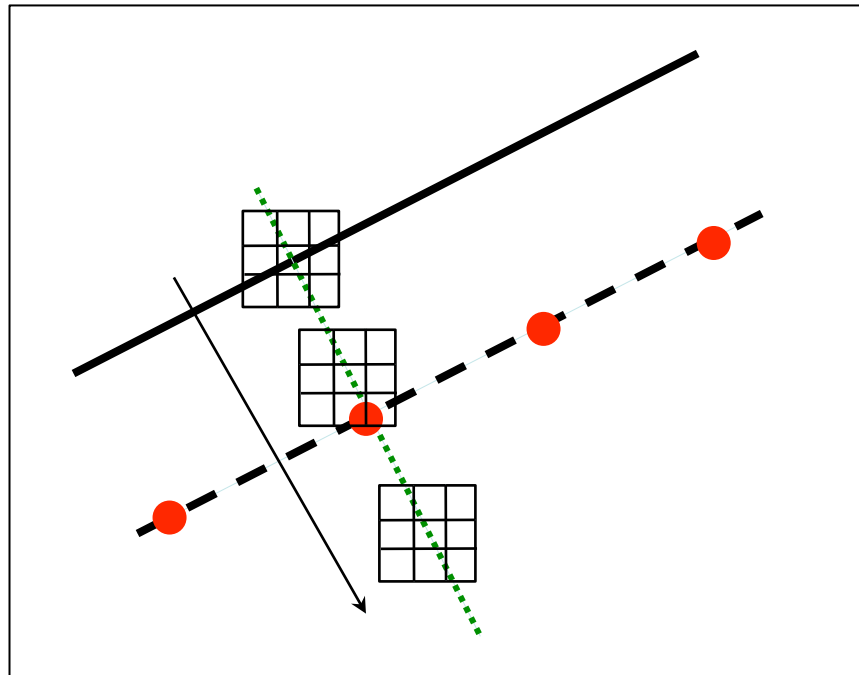
- Capture a new image
- For each vpMeSite : build a list of points along the normal to the edge centered on the edge point previous location



### 3. Moving edges tracker

Method :

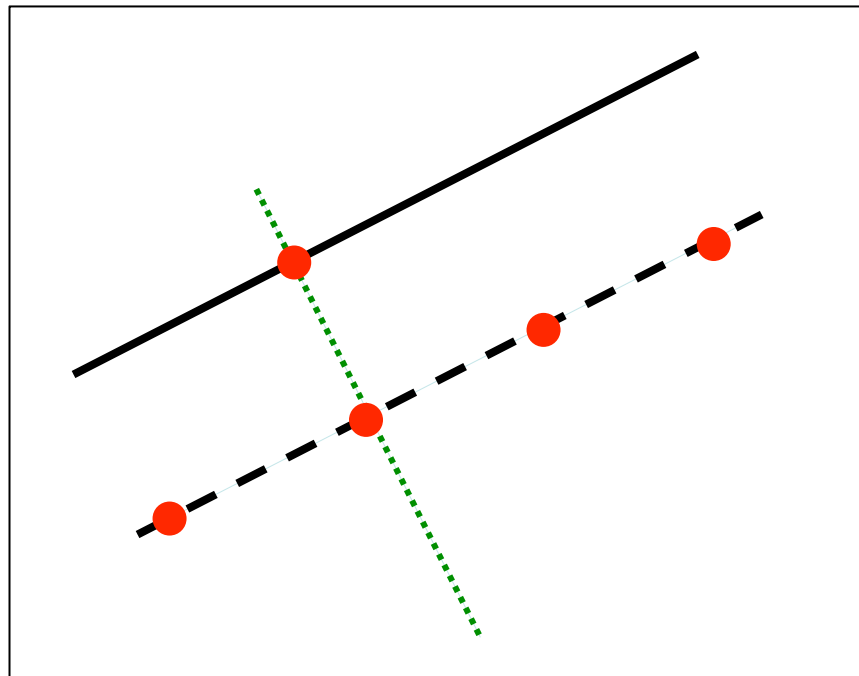
- For each point computes the convolution with a filter optimized to detect edges with an angle near the previous angle



### 3. Moving edges tracker

Method :

- If one point respects the two conditions :
  - The convolution result is close to the previous one
  - The convolution result is high enough
- Then it is considered as the new edge point



### 3. Moving edges tracker

Method :

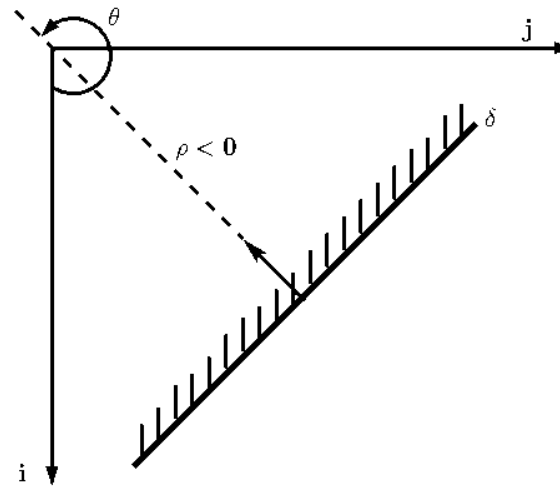
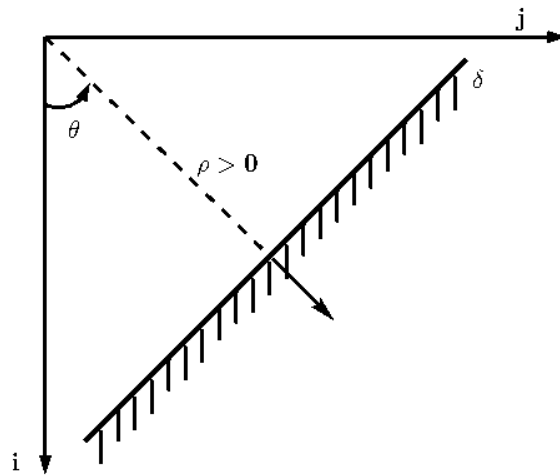
- After all vpMeSite are detected, characteristics of the line, ellipse and nurbs are used to detect outliers
- To suppress outliers : a robust method based on M-Estimators is used

vpMeLine class:

- A line is defined by it's equation

$$ai + bj + c = 0$$

$$i \cos(\theta) + j \sin(\theta) - \rho = 0$$



### 3. Moving edges tracker

vpMeLine class:

- The parameters :
  - a, b and c
  - rho and theta

vpMeEllipse class:

- An ellipse is defined by it's ellipse equation

$$i^2 + K_0 j^2 + K_1 ij + 2K_2 i + 2K_3 j + K_4 = 0$$

- The K parameters are available in vpMeEllipse class





### 3. Moving edges tracker

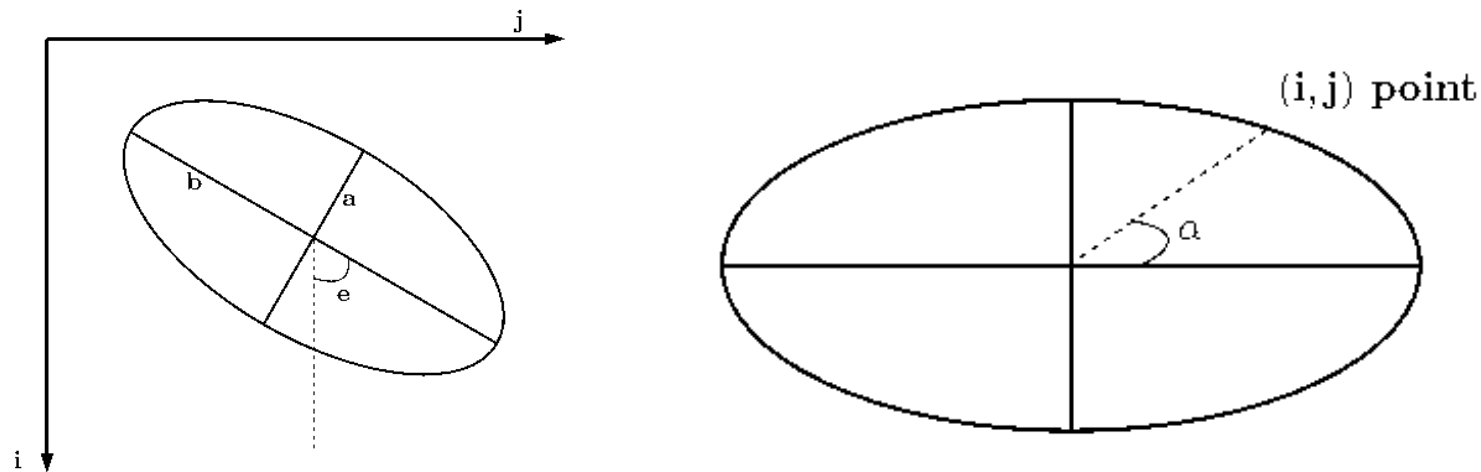
vpMeEllipse class:

- These two equations describe the ellipse points too :

$$i = i_c + b \cos(e) \cos(\alpha) - a \sin(e) \sin(\alpha)$$

$$j = j_c + b \sin(e) \cos(\alpha) - a \cos(e) \sin(\alpha)$$

- Parameters  $i_c$ ,  $j_c$ ,  $a$ ,  $b$  and  $e$  are available in vpMeEllipse.  $\alpha$  is in  $[0, 2\pi]$



### 3. Moving edges tracker

vpMeNurbs class:

- The edge is defined by a parametric curve

$$N_{i,0}(u) = \begin{cases} 1 & \text{si } u_i \leq u < u_{i+1} \\ 0 & \text{sinon} \end{cases}$$

where  $0 \leq u \leq 1$

$$N_{i,p}(u) = \frac{u - u_i}{u_{i+p} - u_i} N_{i,p-1}(u) + \frac{u_{i+p+1} - u}{u_{i+p+1} - u_{i+1}} N_{i+1,p-1}(u)$$

Edge points coordinates  $\rightarrow C(u) = \frac{\sum_{i=0}^n N_{i,p}(u) \omega_i P_i}{\sum_{i=0}^n N_{i,p}(u) \omega_i}$

where  $P_i$  are control points and  $\omega_i$  are weights.

- All the points coordinates are given
- All the derivatives at any points are given too
- All the parameters  $N_{i,p}$ ,  $P_i$  and  $\omega_i$  are available



### 3. Moving edges tracker

Advantages :

- Gives the equation of the tracked edges
- Fast tracking method
- Useful to initialize visual servoing features implemented in ViSP

Limits :

- The speed depends on the number of points and the size of the search range. If the parameters are not optimal, the algorithm can be “slow”
- vpMeNurbs parameters are difficult to set correctly
- vpMeNurbs is not so robust if the shape of the edge is too complex



## 4. 3D model-based tracker

Model-based tracking :

- Track a 3D model thanks to the moving edges method
- Use a virtual visual servoing
- In ViSP implemented in vpMbEdgeTracker class

Method : Initialization

- Require a 3D model (CAO, WRL, ...)
- Need to compute the initial pose
- The pose is used to project the model on the image
- The moving edges points can be initialized



## 4. 3D model-based tracker

### Method : Tracking

- Assuming that the pose corresponding to the previous image is known.
- The new lines are tracked
- The goal is to “move” the pose to match the object in the new image with the projection of the model
- The pose to compute is defined by :

$${}^c\widehat{\mathbf{M}}_o = \arg \min_{{}^cR_o, {}^c t_o} \sum_i \left( \mathbf{p}_{d_i} - pr({}^c\widehat{\mathbf{M}}_o \circ \mathbf{P}_i) \right)^2$$

- The entire model is taken into account during the minimization

