

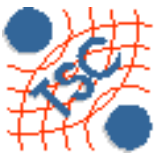


Composite Object Detection in Video Sequences: Application to Controlled Environments

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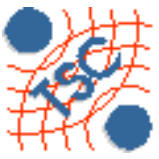




Outline

1. **Introduction** ←
2. Generic algorithm
 - Perceptual Analysis
 - Semantic Analysis
3. Controlled environment
 - Perceptual
 - Semantic
4. Results
5. Conclusions



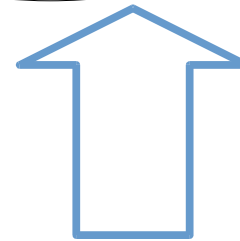


Introduction

- Our goal: Use context to improve object detection performance

There are four laptops in the room located in (X, Y) and with orientation D°

Semantic description



ANALYSIS



Perceptual data (visual)



Introduction

- Our approach: Adapt a **generic indexing** technique to a **specific real-time detection** problem in a controlled environment.

Indexing

- CBIR from database
- Generic environment
- Universal analysis
- Bottom-up
- Loose time restrictions

Detection

- Real-time applications
- Controlled environment
- Class-specific analysis
- Top-down
- Tight time restrictions



ANALYSIS

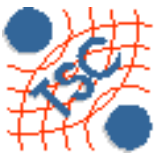
There are four laptops...



Outline

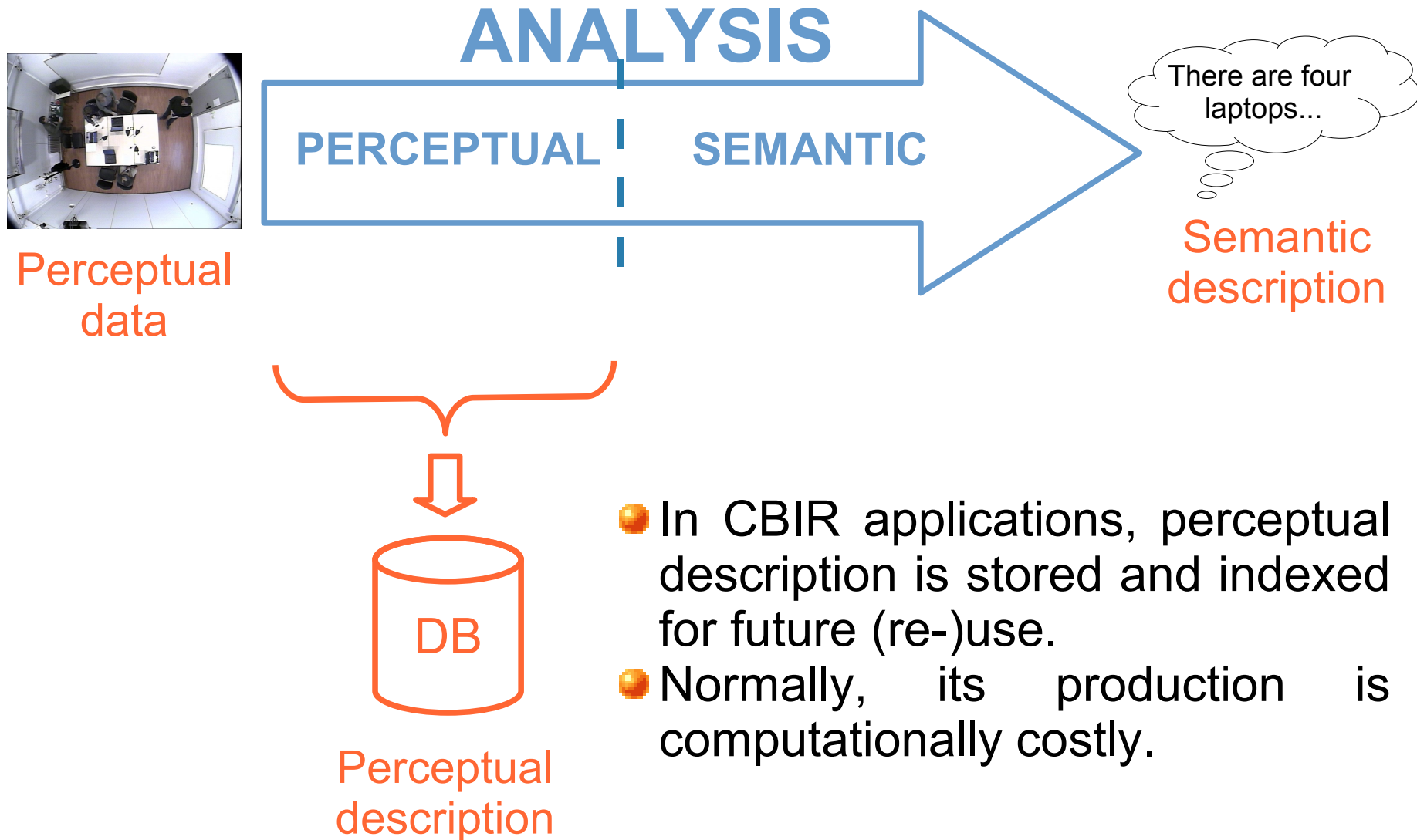
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Generic algorithm: Perceptual

- The proposed generic analysis can be divided in two parts: a perceptual and a semantic one.



- In CBIR applications, perceptual description is stored and indexed for future (re-)use.
- Normally, its production is computationally costly.



Generic algorithm: Perceptual

Region-based [1]

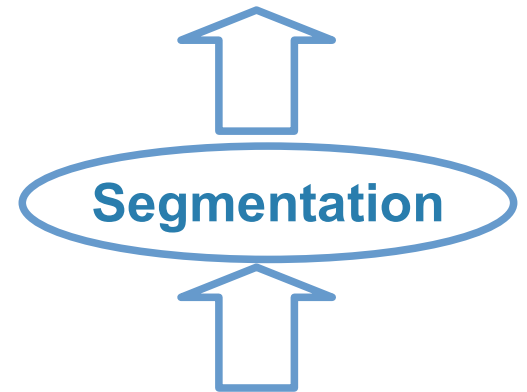
Advantage

- Objects being union of regions
- Solves some over-segmentations
- Simplification: # regions \ll # pixels

Limitation

- Still a large number of possible unions to consider
- Initial partition may present over- and sub-segmentation problems

Regions



Pixels



[1] P.Salembier and F.Marqués, "Region-based representations of image and video: segmentation tools for multimedia services", IEEE Trans. Circuits and Systems for Video Technology (1999).



Generic algorithm: Perceptual

Hierarchy-based [2]

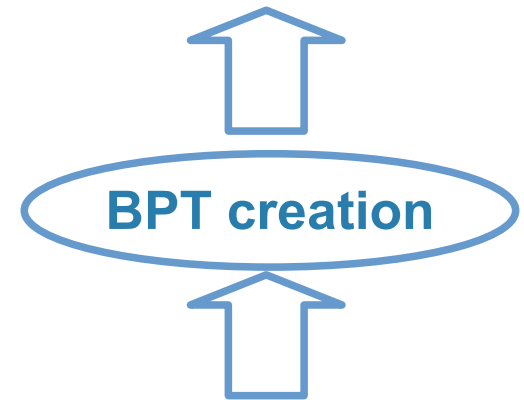
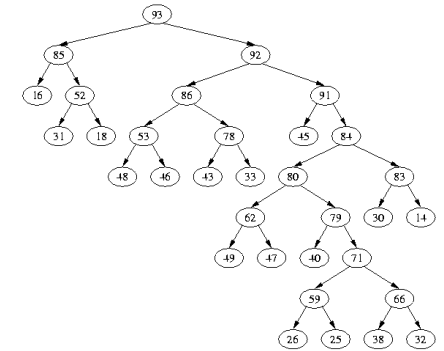
Advantage

- Reduced universe of region unions
- Still multi-resolution

Limitation

- Merging criterion must match the semantic interpretation.

Binary Partition Tree (BPT)



Regions

[2] P.Salembier and L.Garrido, "Binary Partition Tree as an efficient representation for image processing, segmentation and information retrieval", IEEE Trans. on Image Processing (2000)
 [2+] Related approaches: Quad-trees [Pietikainen], component trees [Monasse], min-max trees.



Generic algorithm: Perceptual

Visual descriptors [3]

Advantage

→ Quantization of compact perceptual features relevant for semantic analysis

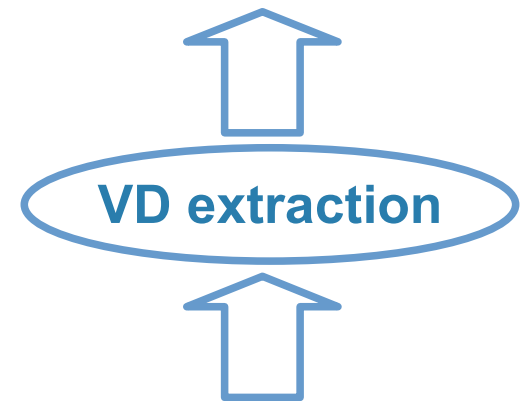
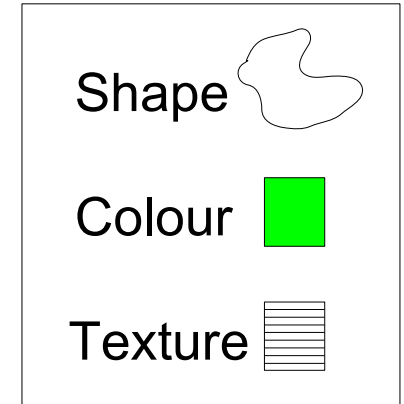
Limitation

→ Extracted descriptors must efficiently characterize the object (or its parts)

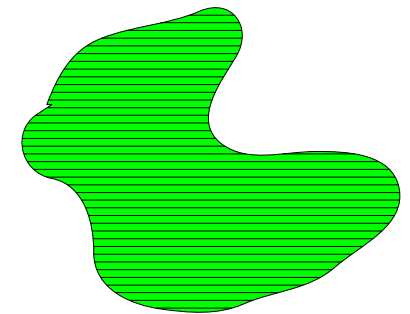
→ Compromise:

↑↑ descriptors ⇒ ↑↑ computation

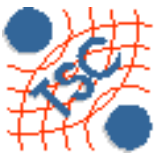
Visual Descriptors (VD)



Regions (BPT nodes)

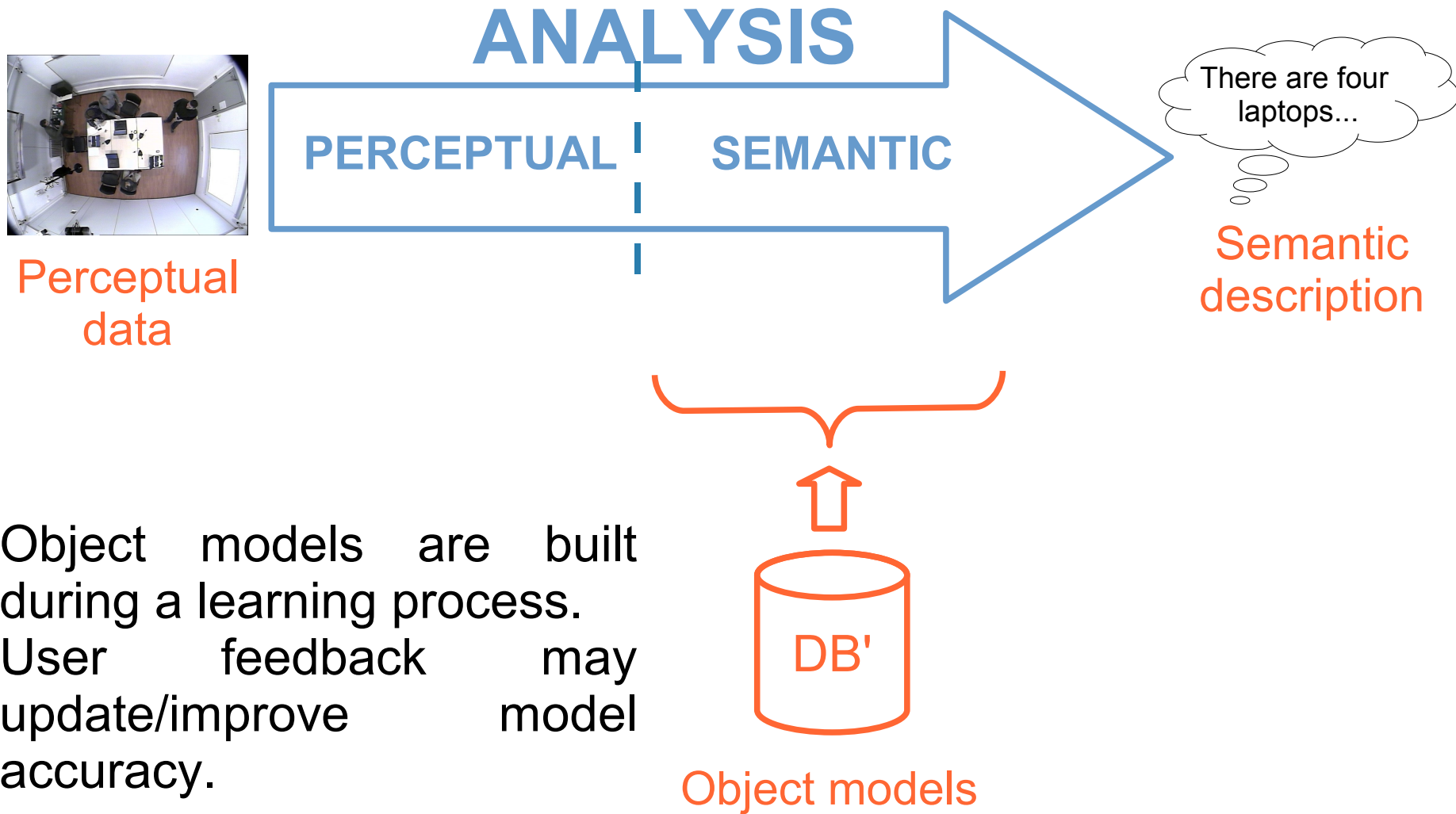


[3] Manjunath, Salembier and Sikora, "Introduction to MPEG-7", Wiley (2002)

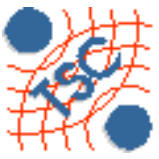


Generic algorithm: Semantic

- The proposed generic analysis can be divided in two parts: a perceptual and a semantic one.

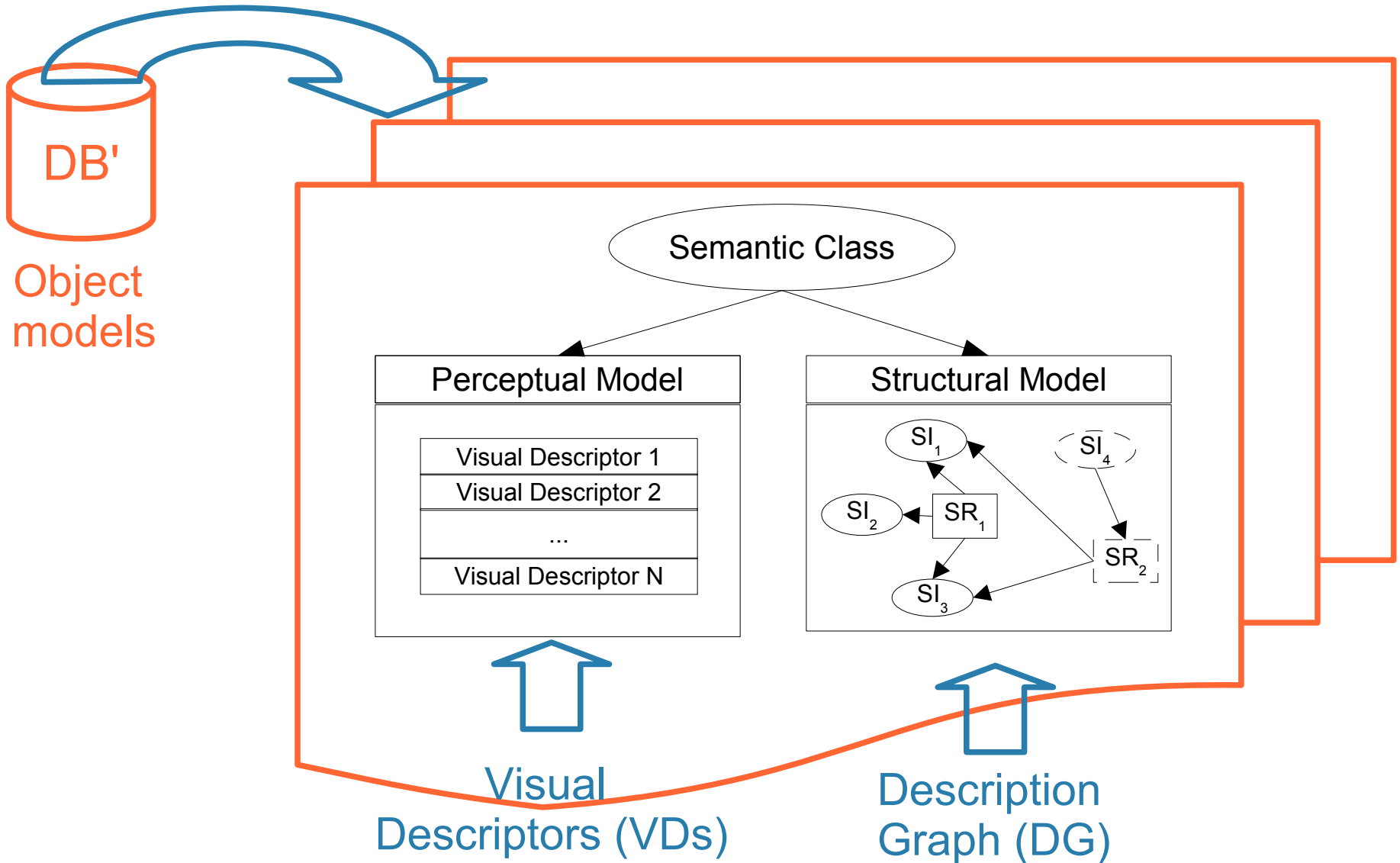


- Object models are built during a learning process.
- User feedback may update/improve model accuracy.



Generic algorithm: Semantic

- Dual models: perceptual (simple) and structural (composite).

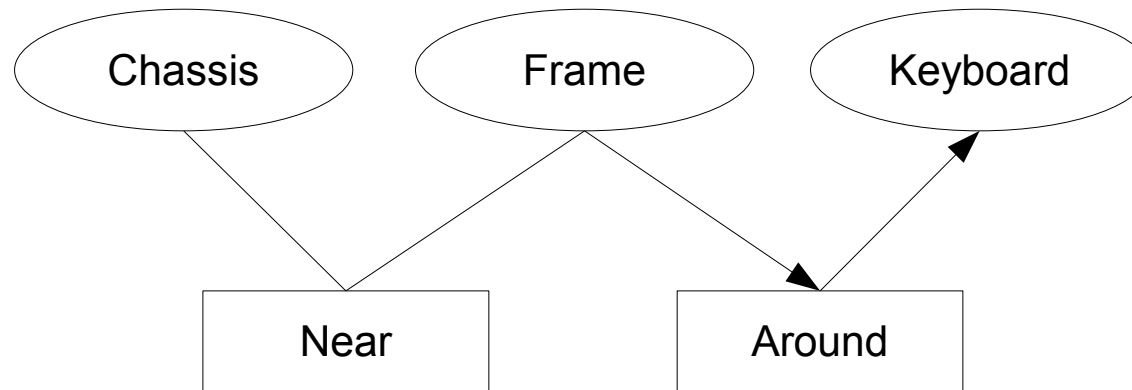




Generic algorithm: Semantic

- Structural models with Description Graph [4].
- A Description Graph (DG) models a semantic class by assigning semantic instances and their Relations to its vertices.

Example: DG of the semantic class “Laptop”



[4] X.Giró and F.Marqués, “Detection of semantic objects using Description Graphs”, ICIP, Genoa, Italy (2005).

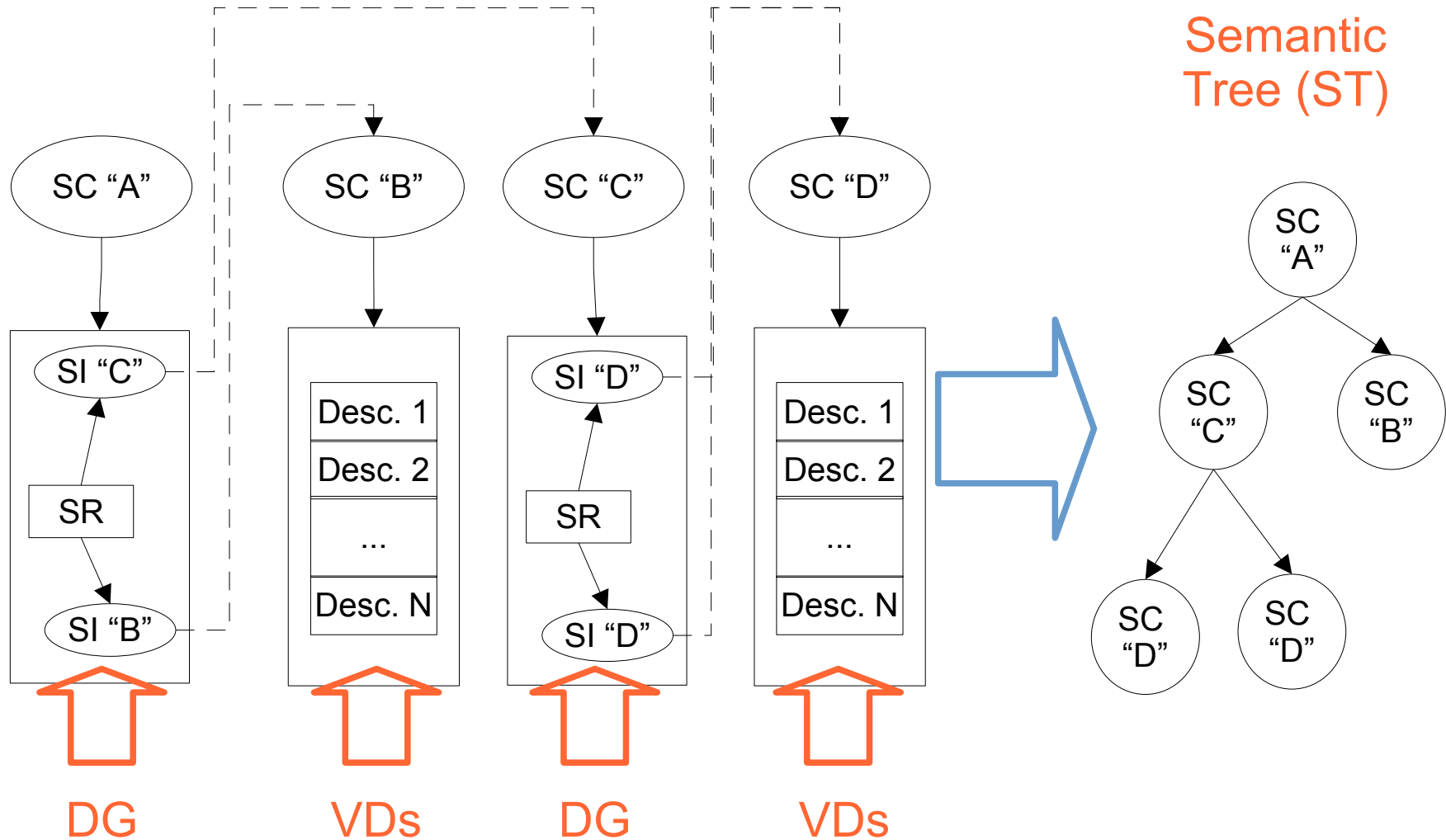
[4+] Related approaches: Constellations [Burl et al.], Factor Graphs [Kozintsev et al], Attributed Relational Graphs [Jaimes et al, Lee et al].

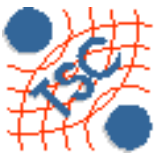




Generic algorithm: Semantic

- Hierarchical model decomposition in Semantic Trees (STs)





Generic algorithm: Semantic

Semantic Trees creation [5]

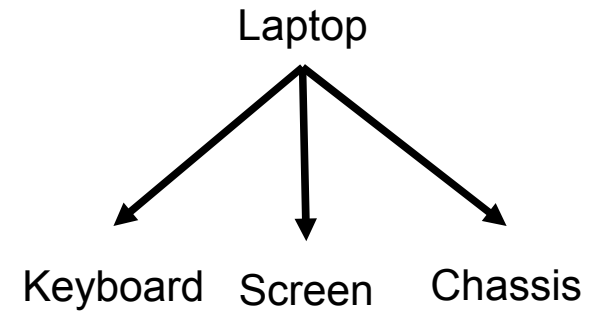
Advantage

- Unique detection algorithm for simple and composite objects.
- Supports non-connected objects.

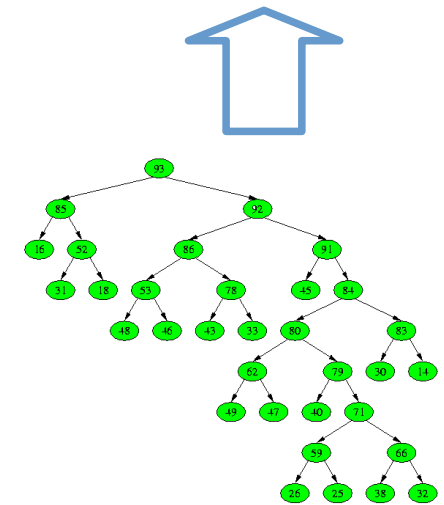
Limitation

- Semantic models and classifier must effectively deal with perceptual variability.

Semantic Tree (ST)



BPT + VDs



[5] X.Giró and F.Marqués, "From Partition Trees to Semantic Trees", MRCS. Istanbul 2006.



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Context-based analysis

- Problem: The generic algorithm developed for indexing is not efficient in object detection in controlled environment.



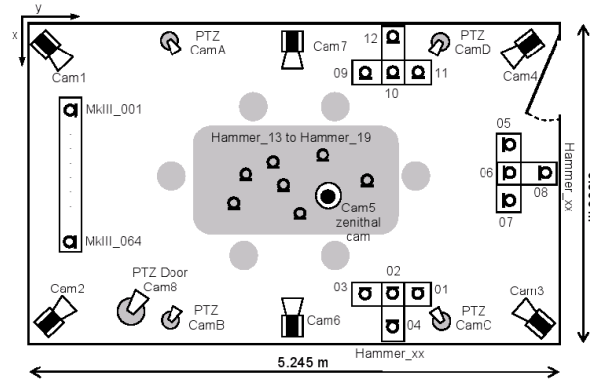
- Improvement: Use the context to reduce the search space



Spatial

- Fixed zenithal camera
- Fixed table
- Static and known background

Smart room



Application

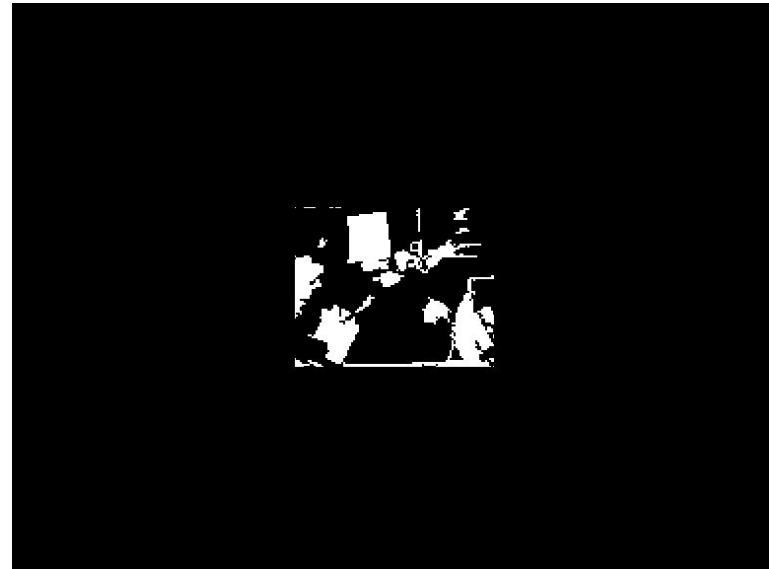
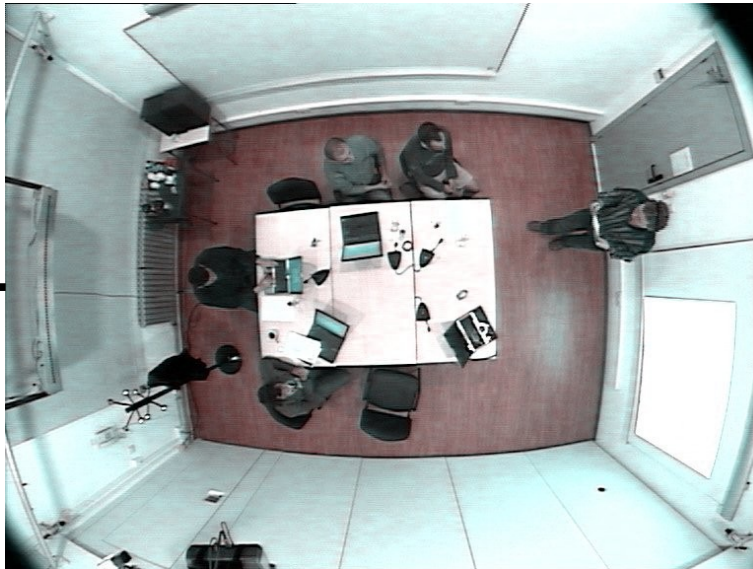
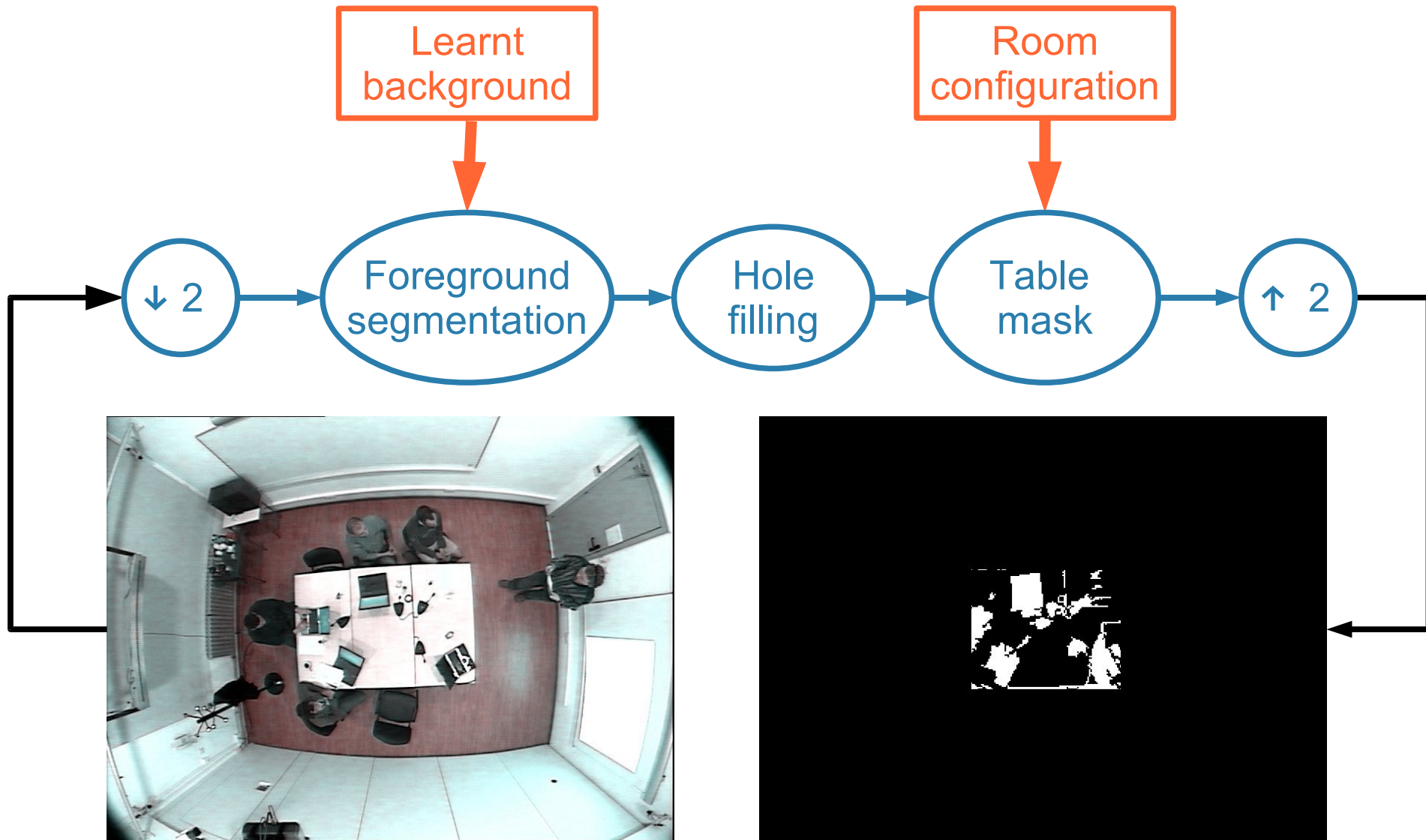
- Class of interest are laptops.
- Limited universe of laptops.
- Only interested in laptops on the table
- Video sequence

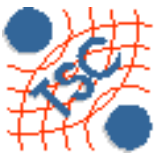




Perceptual analysis: spatial context

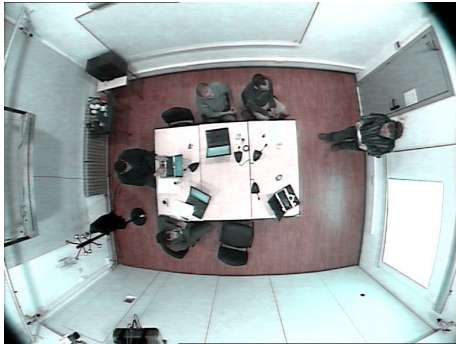
- Dynamic generation of foreground masks on table



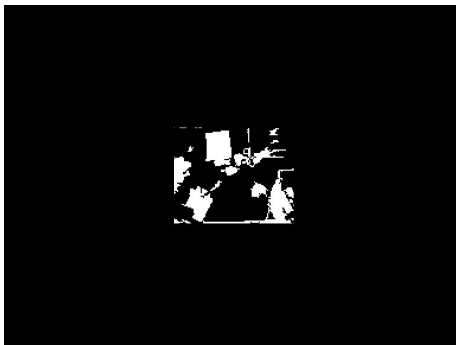


Perceptual analysis: spatial context

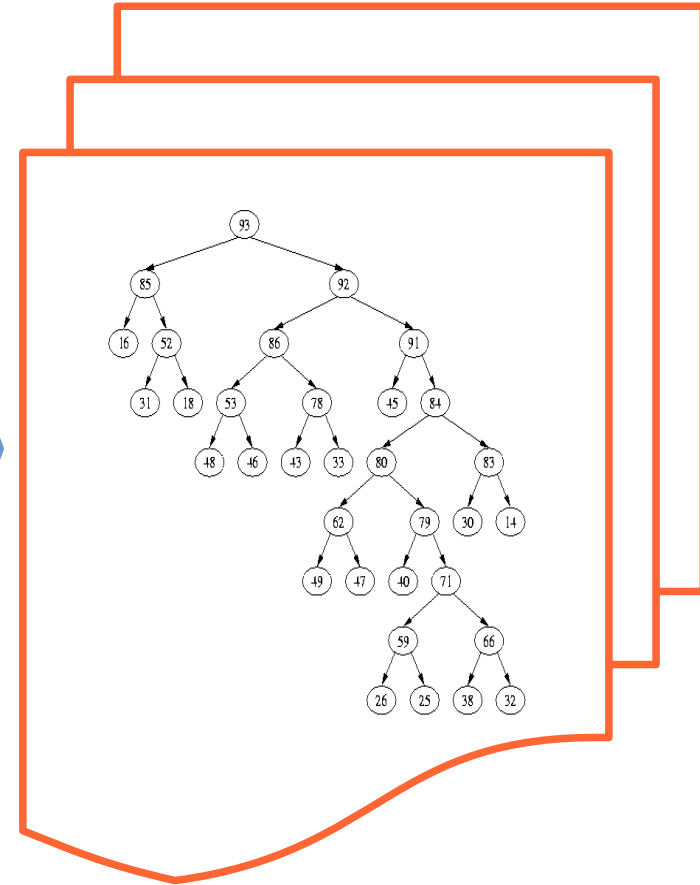
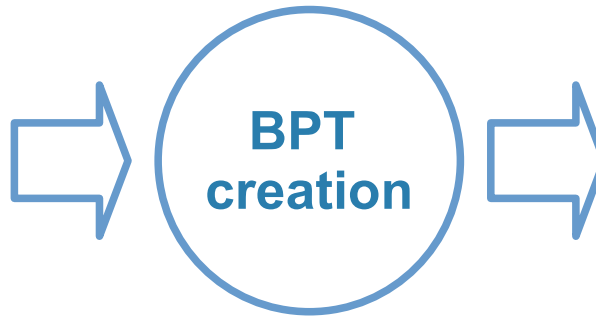
- Disconnected segments in mask generate BPT forests.



+

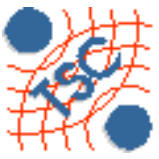


N connected segments



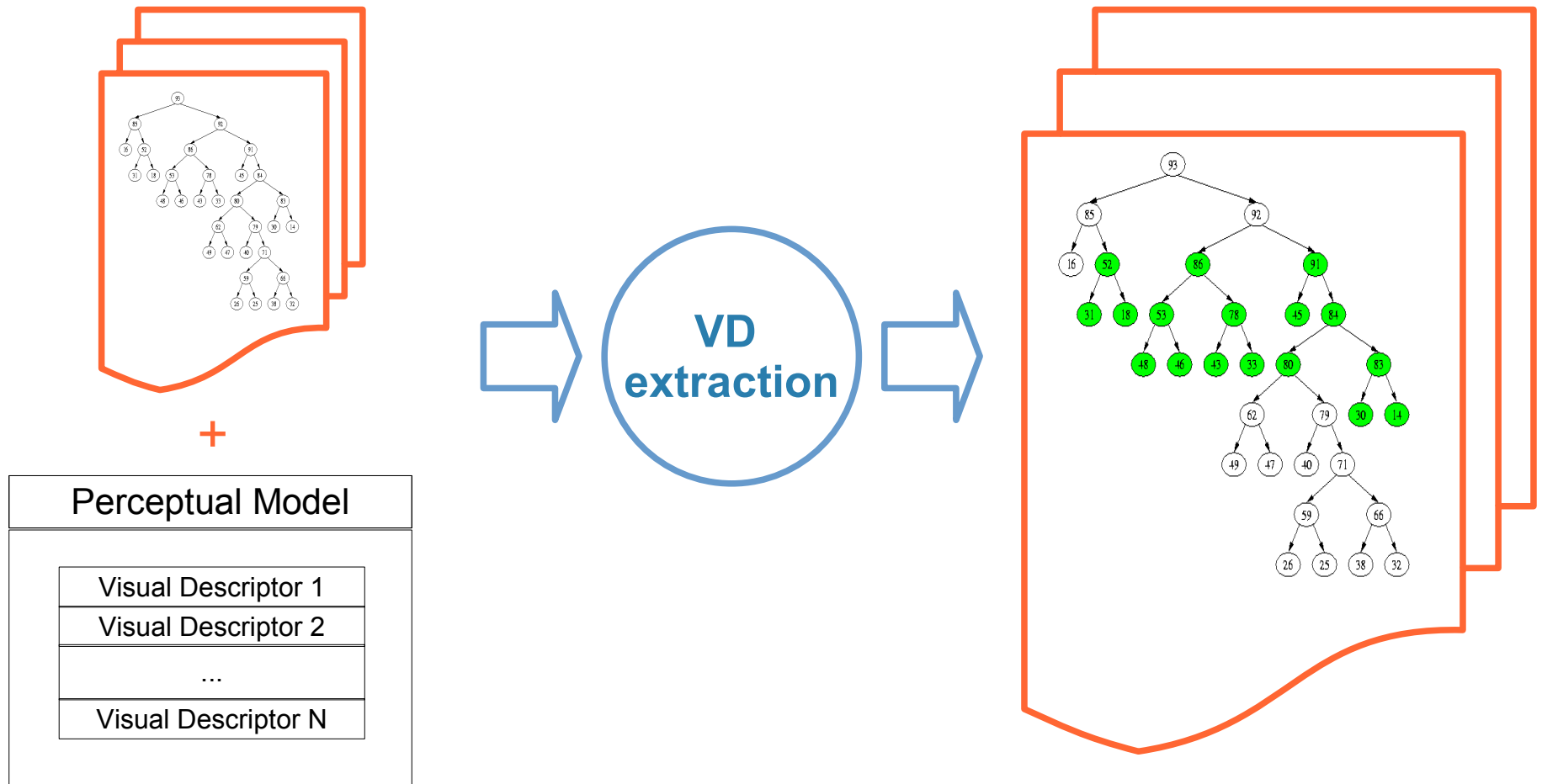
N BPT's
(BPT forest)





Perceptual analysis: application context

- Simple descriptors in models may discard several BPT nodes for VD extraction.



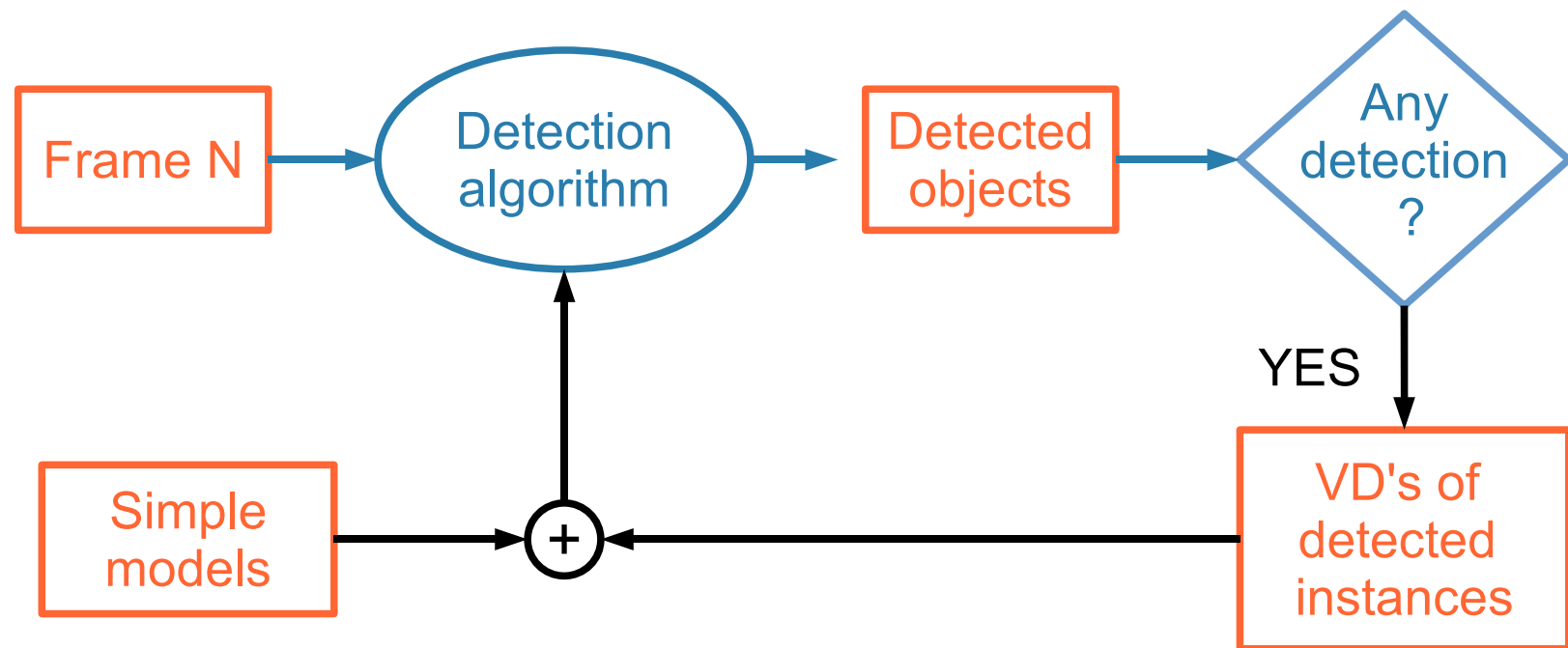


Semantic analysis: application context

- Problem: Partial occlusions gradually in time.



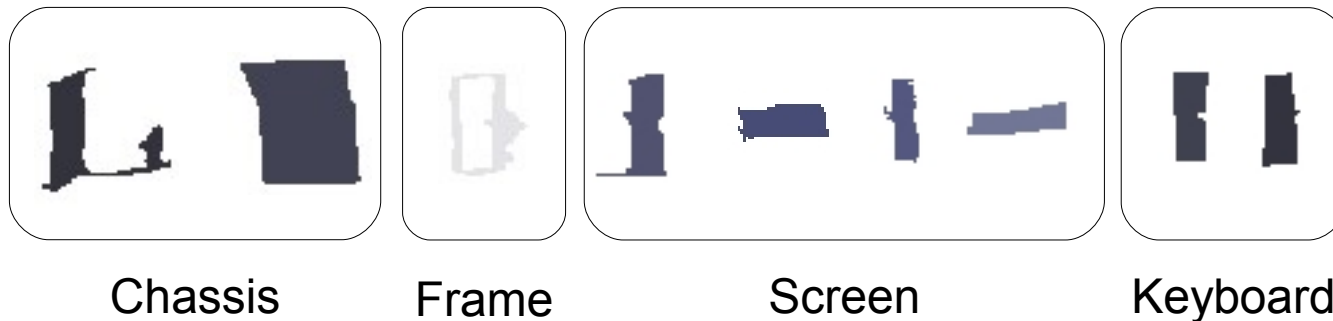
- Improvement: Simple models are updated after each object detection to cope with gradual changes.



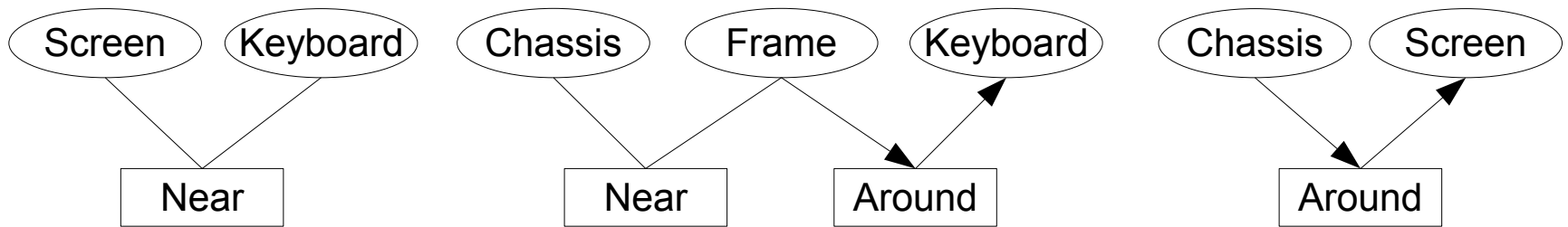


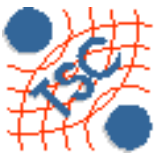
Semantic analysis: application context

- Problem: A single semantic class can be represented by multiple simple and/or composite models.
- Improvement: Support multiple models for single class.
 - simple: multiple sets of descriptors/classifiers



- composite: multiple structural models (eg. DG's)





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Results

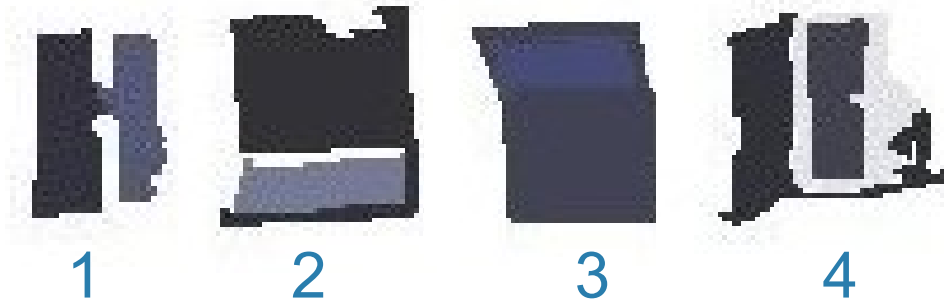
● Perceptual analysis in context

Dynamic Masks → 90-99% pixels discarded

Discriminative descriptor : → 75-80% regions discarded
(area)

● Semantic analysis in context

Detection of four laptops with perceptual variability



Laptop	Precision
1	0,8
2	0,86
3	0,8
4	0,97





Outline

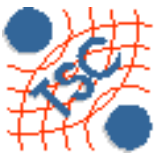
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Conclusions

- Methodology: CBIR → Real time object detection.
- Context provides optimization.
 - **Spatial**: less pixels to analyse.
 - **Semantic**: less descriptors to extract.
 - **Temporal**: models update
- Future goals:
 - Study of descriptors according to their BPT expansion, discrimination and computation cost.



Outline

Further details: http://gps-tsc.upc.es/imatge/_Xgiro/start.html

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Thank your for your attention

