

Vectorization in Graphics Recognition: To Thin or not to Thin

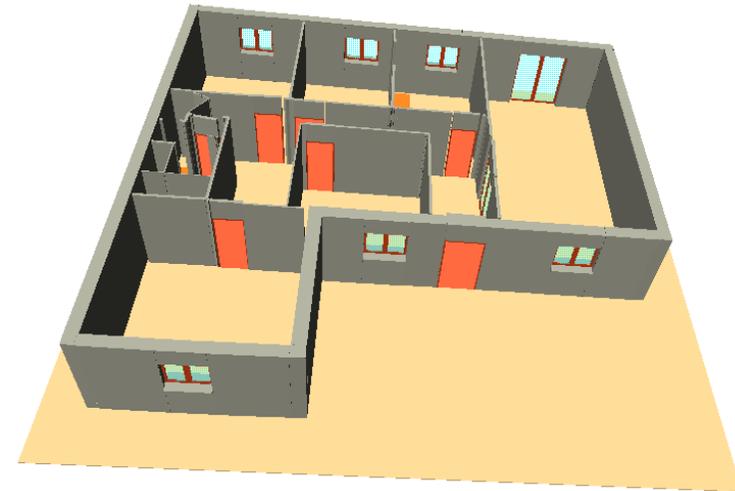
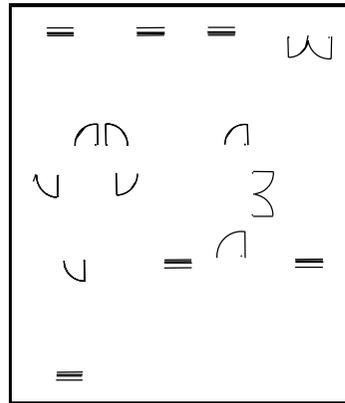
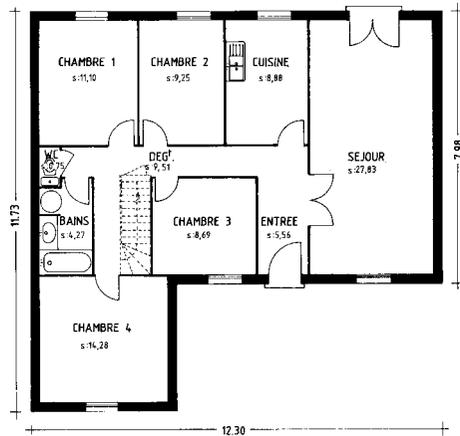
Karl Tombre and Salvatore Tabbone



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Graphics recognition

Analysis of graphics-intensive documents



Main feature = the *vector* \Rightarrow raster-to-vector conversion plays a central rôle

Choice of vectorization method

Compute a medial axis or not?

- ➔ medial axis computations are classical for image processing people
- ➔ skeletons often used
- ➔ faithful representation of binary shape \neq retrieve actual thoughts of draftsman
- ➔ Special problems with junctions and line extremities

The steps of vectorization

- ① Find the lines ←
- ② Approximate the lines into vectors
 - ↳ Rosin & West split and merge
 - ↳ Wall & Danielsson iterative
- ③ Add contextual information
- ④ Dashed-line detection
- ⑤ Arc detection → *poster presented yesterday*

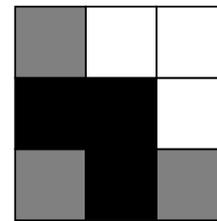
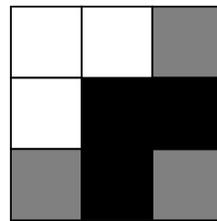
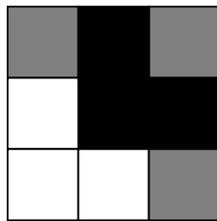
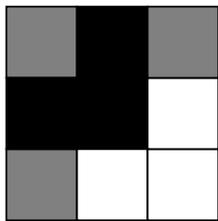
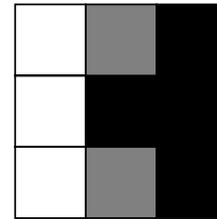
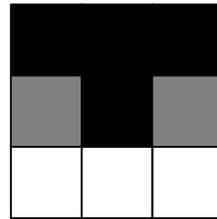
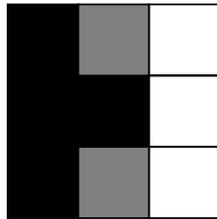
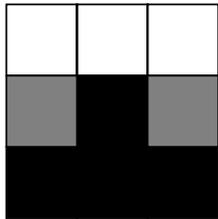
Finding the lines

From raster image to lines (chains of pixels) → significant medial axes

- ① Compute the skeleton
 - iterative thinning
 - distance transform
- ② Match the opposite sides of the line
- ③ Sparse-pixel approaches

Skeletonization

Iterative thinning : “peeling an onion”



- ⤵ small image buffer size
- ⤵ multiple passes, sensitive to noise

Distance transform : 3–4 DT [Sanniti di Baja]

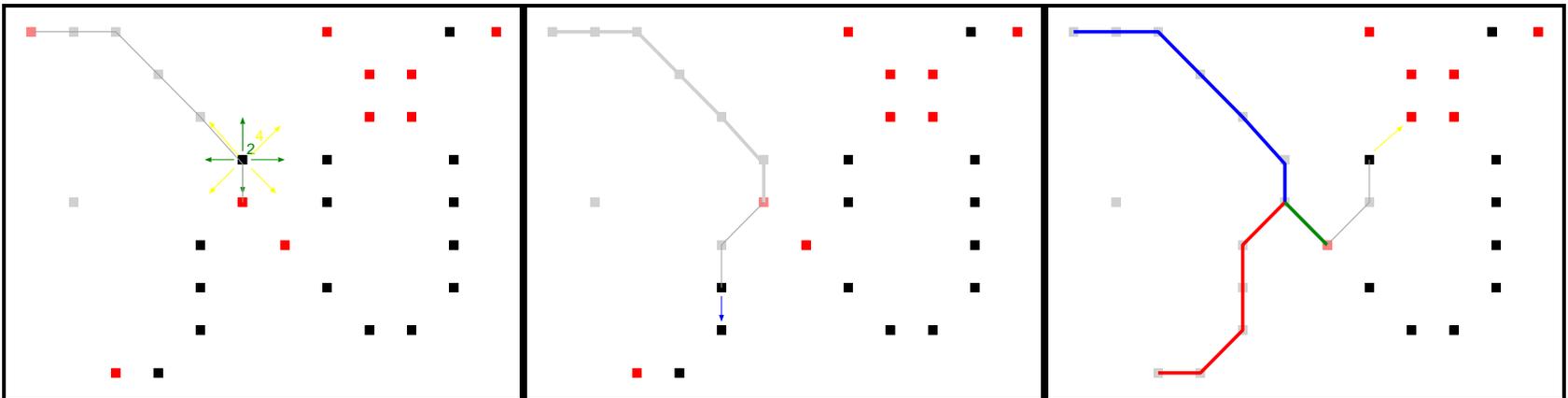
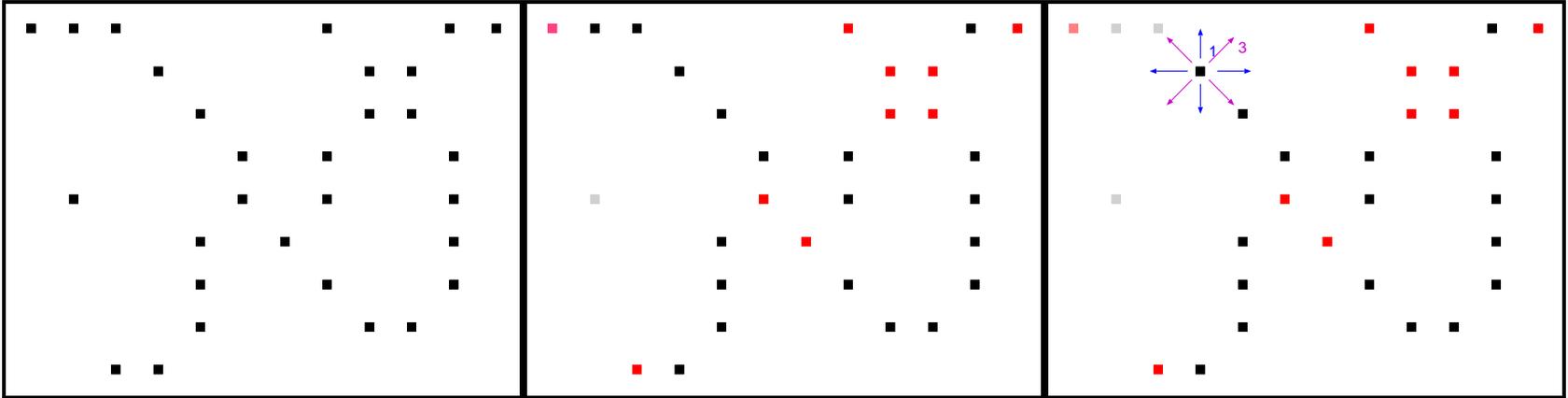
V_2	V_3	V_4
V_1	X	V_5
V_8	V_7	V_6

Pass 1 : $Y = \min(V_1 + 3, V_2 + 4, V_3 + 3, V_4 + 4)$

Pass 2 : $Z = \min(Y, V_5 + 3, V_6 + 4, V_7 + 3, V_8 + 4)$

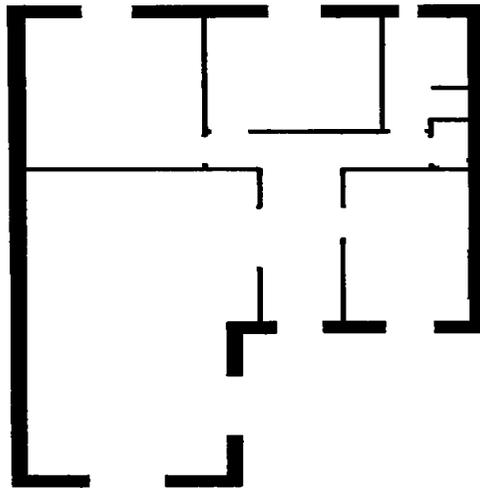
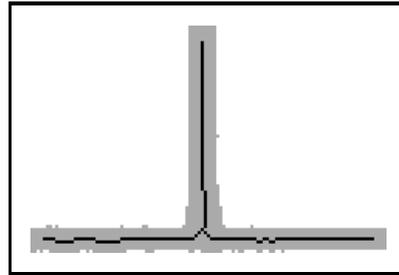
- Detection of local maxima
 - Tracking the skeleton
 - Reduce to thickness 1
 - Barb removal
- good compromise w.r.t. precision, simplicity and robustness

Linking the skeleton

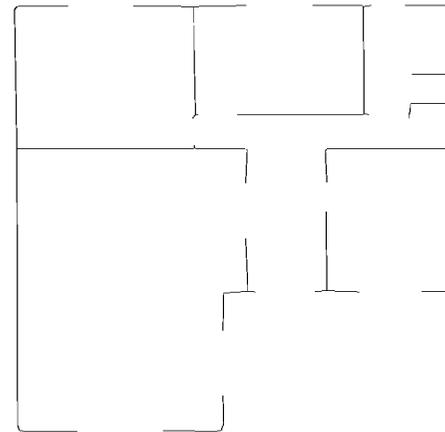


Main weakness

Skeletonization displaces junction points



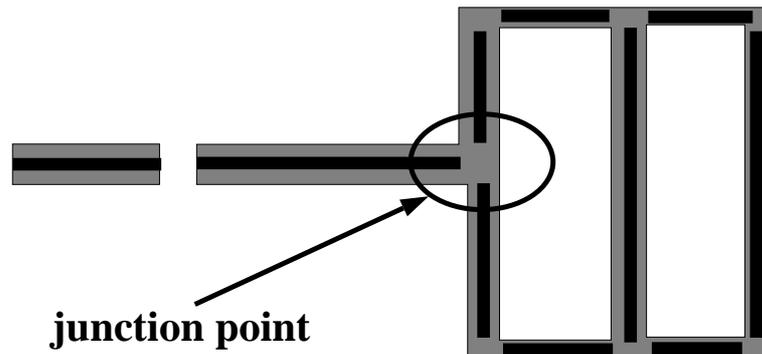
Original image



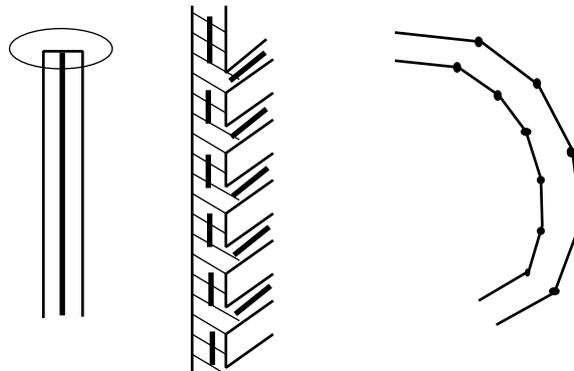
Polygonal approximation of 2–3 skeleton

Matching opposite contours

REDRAW: polygonal approximation of connected components contours, and match segments

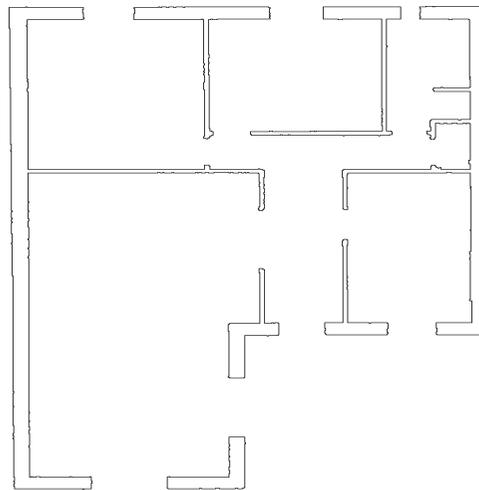


Problems: thresholds, one-to-many and many-to-many matches

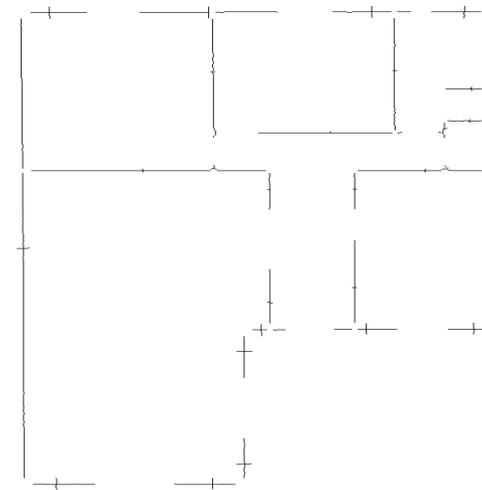


Improvement: use the gradient orientation

Gradient vector (Canny operator) perpendicular to contour \Rightarrow dense and robuste matching of each contour point in the direction of the gradient



Contour detection



"Skeleton" obtained by dense matching

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Contextual information

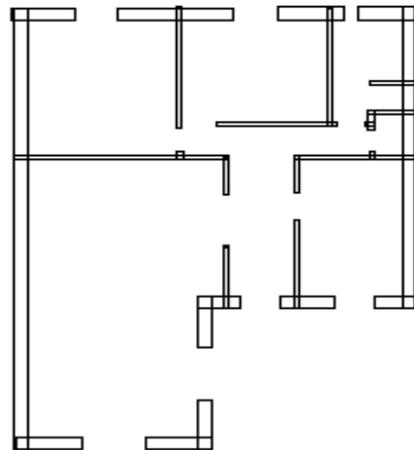
Add domain knowledge

- ✓ specific documents → ad hoc vectorization (e.g. [Chhabra])
- ✓ heuristic correction → adds new thresholds and parameters
- ✓ add constraints [Röösli & Monagan]
- ✓ optimize junction position through iterative fitting [Janssen]
- ✓ introduce models of ideal junctions (T, L, Y . . .) and correct by fitting

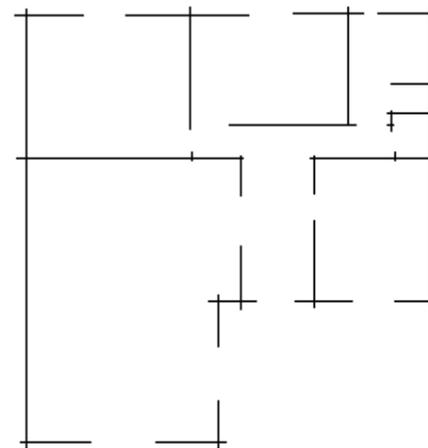
Knowing what you look for

The case of architectural drawings: looking for the walls

Dense contour-matching followed by search for horizontal and vertical rectangles



Reconstructed rectangles



Medial axes of rectangles

Discussion

- ↳ Contour-matching + contextual information yield good results...
- ↳ ... but in general case, contextual information not exhaustive
- ↳ Skeletonization is generic and robust, but yields weak junctions
- ↳ skeleton: emphasizes *distance* criterion
- ↳ dense matching: emphasizes *direction* criterion

A possible solution

Three-step vectorization under study:

- ➔ first crude vectorization to segment a drawing into pieces having same direction
- ➔ directional distance transforms on each piece
- ➔ junction models for better precision at intersections

Conclusion

- ✓ Skeletons still most general and generic
- ✓ Junctions and extremities still not satisfactory
- ✓ Combine directional and distance criteria
- ✓ Keep an eye on emerging methods (Markov Random Fields, ...)