User Guided Image Abstraction for Vectorization

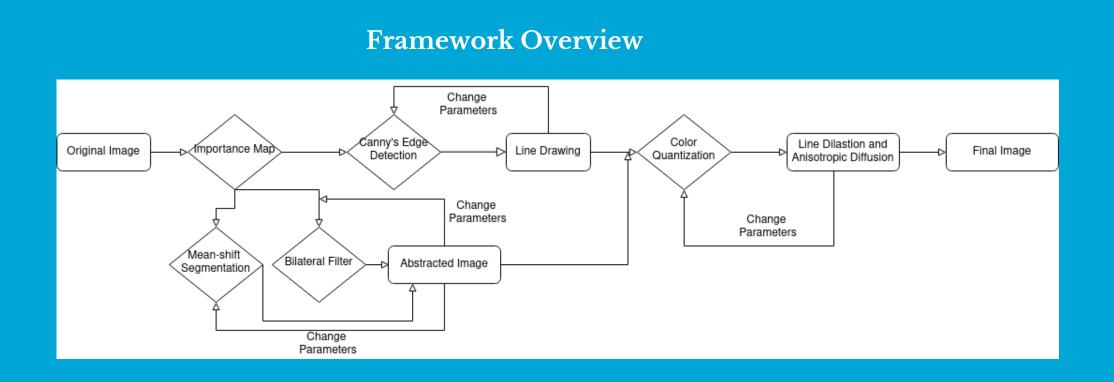
1. Introduction

- Pixelated images do not scale well
- Low-resolution images -> easily scaling vector images
- Essential step: Image Abstraction
- Reducing the complexity of an image while keeping the important information intact
- Important information:
 - > Saliency, foreground & background, colors, light, edges, depth

The Goal of the Framework

2. Research Question

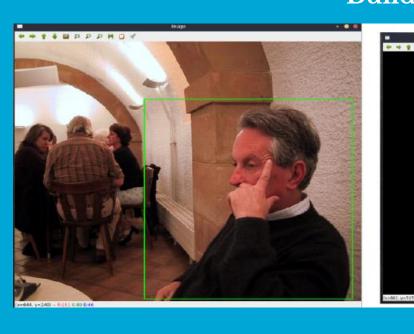
How can a photograph be abstracted to simplify its components with the goal of producing suitable content for vectorization with user interaction?



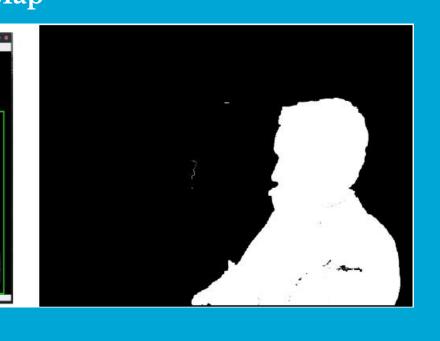
3. Method

- 1. Importance map creation using GrabCut
 - Guided by user annotations
 - Later stages: important components are abstracted differently/less

Building the Importance Map

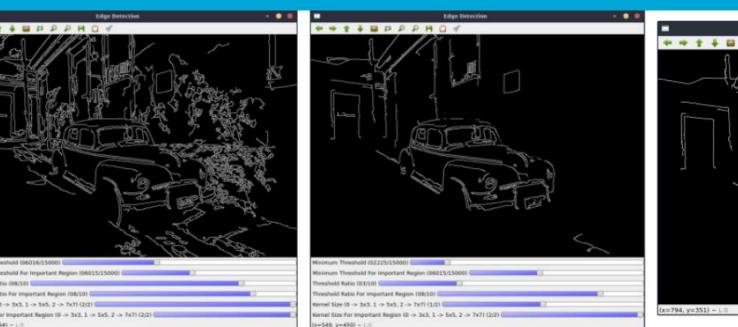


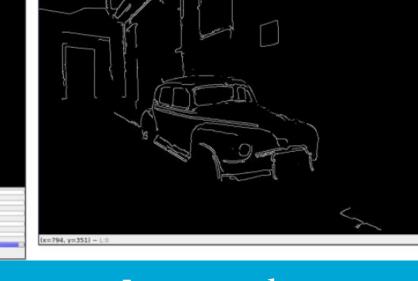
Without importance



- 2. Line drawing using Canny's edge detection
 - > Various parameters, delete incorrect edges, independent control over important region

Line Drawing by Edge Detection





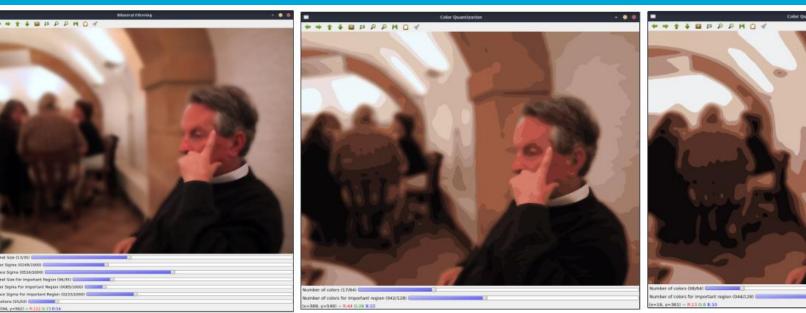
Incorrect edges

3. Abstraction with independent control over important region

With importance

- > Mean shift segmentation
- Hill-climb clustering
- Bilateral filtering
- Smoothing, edge preserving
- 4. Color quantization
- K-means clustering to reduce color palette, important region and background altered independently

Different Bilateral Filtering Results



Only bilateral

bilateral filter & less color quantized

More abstracted bilateral filter & more color quantized

5. Line thickness and anisotropic diffusion

4. Results and Conclusions

- Images are abstracted
 - > Reduced color space,
 - > Clear boundaries and edges
 - > Regions with same color
 - > Important information retained
- Both methods work well with automatic vectorization software (AutoTrace)
- Provide user input at every stage
 - user stays in the loop
- > Results depend on the user's choices and needs
- Abstraction while retaining important features
 - > Without the need of special hardware like eye-trackers
 - > Bilateral filtering and meanshift work well in different types of images
 - > Color quantization adds flexibility
- Bilateral filtering with quantization:
 - > Color space reduced less, smaller uniform-colored regions
- ➢ Better for landscape photos → complicated vectorization that requires more detail
- Mean-shift with quantization:
- > Larger regions with uniform color
- > Abstracted more > less detail. suits the need for simpler vectorization

Portrait Photograph Vectorized

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Portrait Photograph

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Landscape Photograph

Bilateral & Color Quantization



Complex Photograph





5. Limitations and Future Work

- Bilateral: false edges or loose edges
- Mean-shift: number of clusters. high-dimensionality causes artifacts, textures are lost
- Canny: disconnected lines -> Add scale space theory
- Importance map: one level of importance -> Add many levels
- Quantization: sensitive to outliers, different clusters and compactness each time -> Explore different clustering algorithms

6. References

mountains-landscape. Pixnio.10

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