



Statistical Models of Appearance for Image Interpretation

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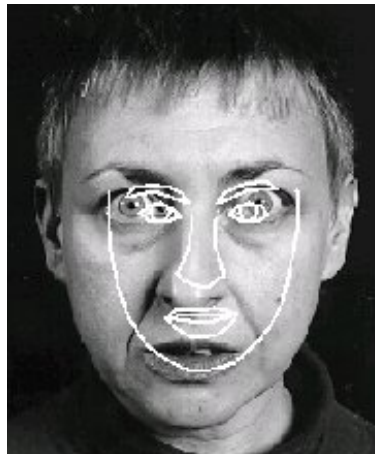
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[Problem Overview]

- Image Interpretation
 - Prior knowledge
 - Structure Recovery
 - Labeling
- Model Based Approach
 - Describe and label expected structure
 - Interpretation \equiv Matching

[Example]



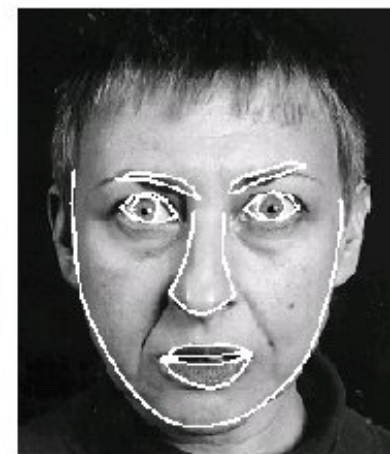
Initial



After 2 iterations



After 6 iterations



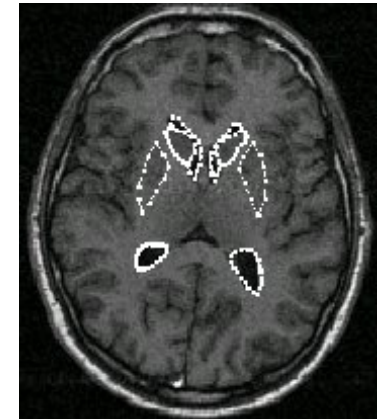
After 18 iterations

[Modeling]

- Model Characteristics
 - What? Shape or Appearance
 - Active Shape Model
 - Active Appearance Model
 - Deformable – General & Specific
- Method
 - Learn variation from annotated image set

[Modeling]

- Training Set
 - Annotated Images:
Key landmarks
- PCA
 - Shape
 - $$\bar{x} = \bar{x} + P_s b_s$$
 - Appearance
 - $$\bar{g} = \bar{g} + P_g b_g$$



[Modeling]

- Possible correlation b/w shape and appearance

- PCA on b_s & b_g

$$\begin{pmatrix} \mathbf{W}_s \mathbf{b}_s \\ \mathbf{b}_g \end{pmatrix} = \mathbf{b} = \begin{pmatrix} \mathbf{Q}_s \\ \mathbf{Q}_g \end{pmatrix} \mathbf{c} = \mathbf{Q} \mathbf{c}$$

$$\mathbf{x} = \bar{\mathbf{x}} + \mathbf{P}_s \mathbf{W}_s^{-1} \mathbf{Q}_s \mathbf{c} \quad , \quad \mathbf{g} = \bar{\mathbf{g}} + \mathbf{P}_g \mathbf{Q}_g \mathbf{c}$$

- Generating shape & texture in image

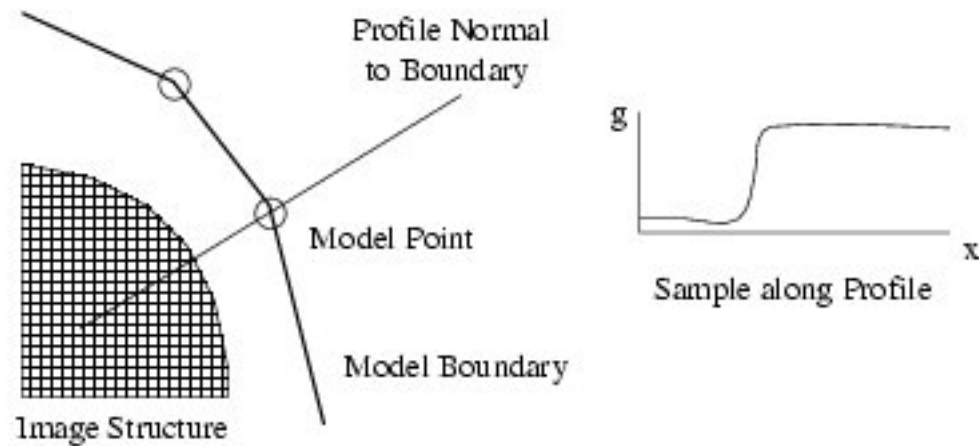
- $X = S_t(x)$

- $g_{im} = T_u(g)$

[Active Shape Model]

- Rough starting approximation of (b, t)
- Iterate
 - Create an instance of X
 - Examine region **around** X_i
 - Update X_i by replacing by best match X_i'
 - Update (b, t) to best fit new X
 - Repeat until convergence

[Active Shape Model]

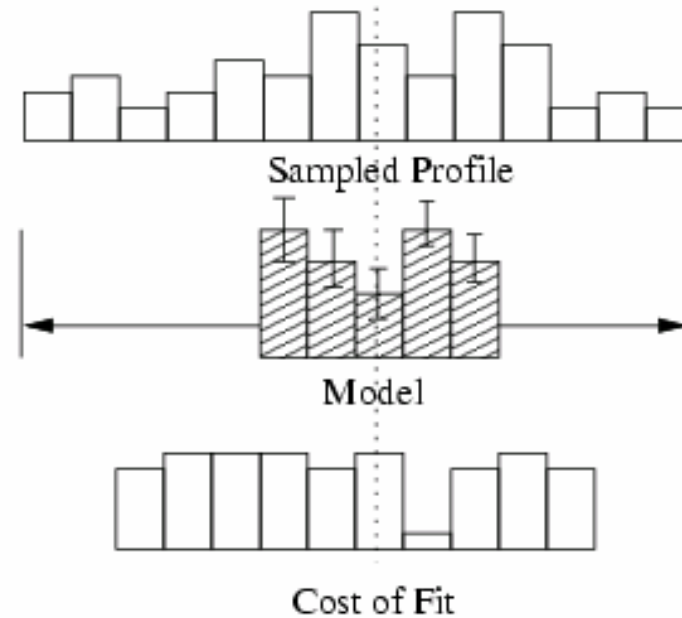


[Modelling Local Structure]

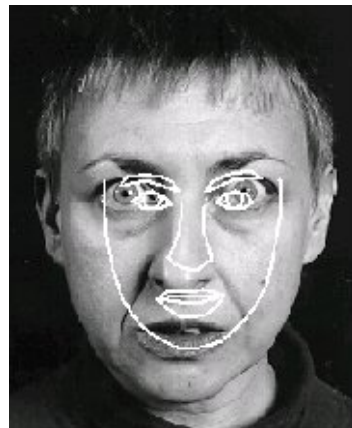
- Sample along a profile k pixels on each side to get $2k+1$ samples, g_i
- Normalize g_i
- Assume multivariate gaussian distribution, estimate \bar{g} & S_g
- Quality of fit of a new sample g_s

$$f(\mathbf{g}_s) = (\mathbf{g}_s - \hat{\mathbf{g}})^T \mathbf{S}_g^{-1} (\mathbf{g}_s - \hat{\mathbf{g}})$$

[Modelling Local Structure]



[ASM: Example]



Initial



After 2 iterations



After 6 iterations



After 18 iterations



Initial



After 2 iterations



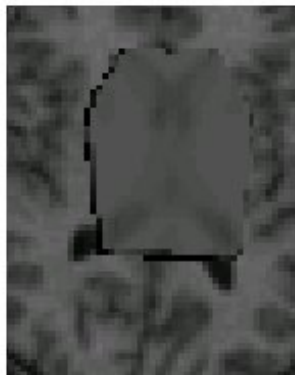
After 20 Iterations

[Active Appearance Models]

- Rough starting approximation of (c, t, u)
- Iterate
 - Create an instance of X
 - Sample pixels in this region, g_{im}
 - $g_s = T_u^{-1}(g_{im})$
 - $g_m = g^- + P_g Q_g c$
 - $r(p) = g_s - g_m$ $(p' = (c'|t'|u'))$
 - $p = p + k^* r(p)$

[AAM: Example]

- Central structure of brain slice
- 10000 pixels & 30 **c** parameters



Initial



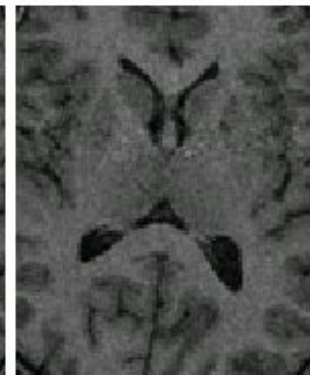
2 its



6 its



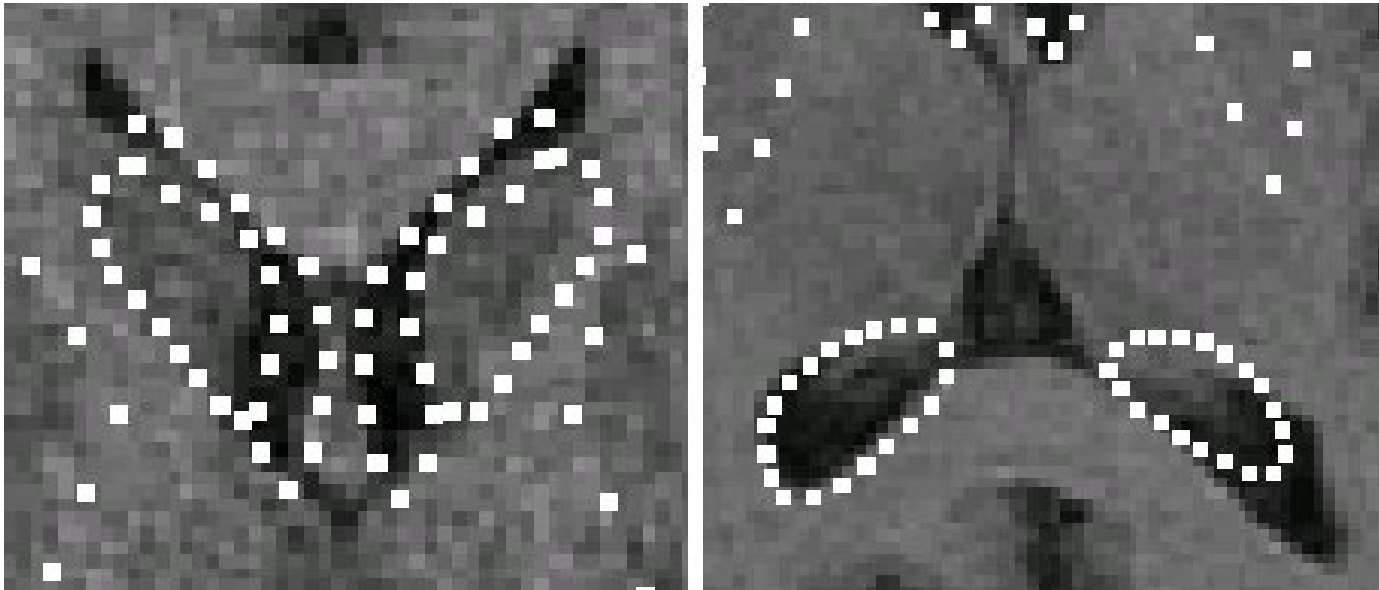
16 its (converged)



original

[AAM: Limitations]

- AAM usually close to optimal but not exact



[AAM: Improvements]

- Model whole visible structure
- Explicit searching outside current patch e.g. along normal as ASM
- Multiple starting points
- Combine AAM & ASM: Weighted average of independent runs

[Multi-Resolution Framework]

- Methodology

- Generate image pyramid by sub-sampling
- Search subject in coarse image
- Refine location at higher resolution

- Advantages

- Efficiency
- Robustness

[Comparison: ASM v/s AAM]

- ASMs look around, AAMs look inside
 - ASM larger capture range, fast
 - AAM more stable
- ASM need more model points, AAM, convincing model with relatively small number of landmarks
- ASM, more accurate feature point location, AAM, better texture match