

图像认知中的优化问题

陈健生

图像语音与网络信息研究所

清华大学电子工程系青年教师论坛

二零一一年五月

- 优化问题
- 总变差方法 (Total Variation Methods)
- 主动形状模型 (Active Shape Model)
- 图割图像分割 (Graph Cut Image Segmentation)
- 人脸图像光照变换
- 焊缝检测
- 多生物特征融合

- 优化问题
- 总变差方法 (Total Variation Methods)
- 主动形状模型 (Active Shape Model)
- 图割图像分割 (Graph Cut Image Segmentation)
- 人脸图像光照变换
- 焊缝检测
- 多生物特征融合

优化问题

自然过程往往最小化（最大化）某个物理量



势能最小



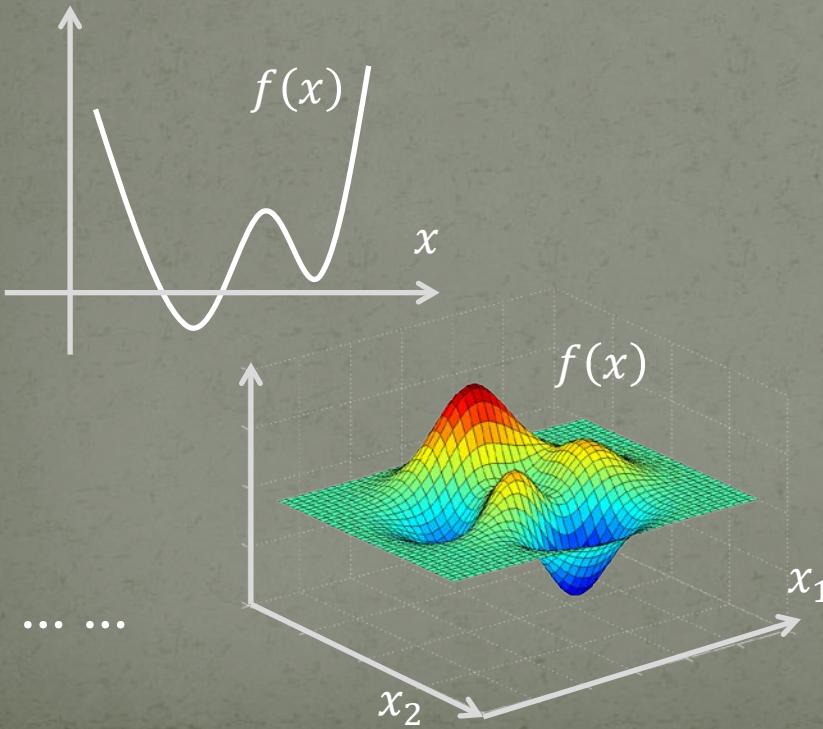
熵最大

优化问题

目标函数

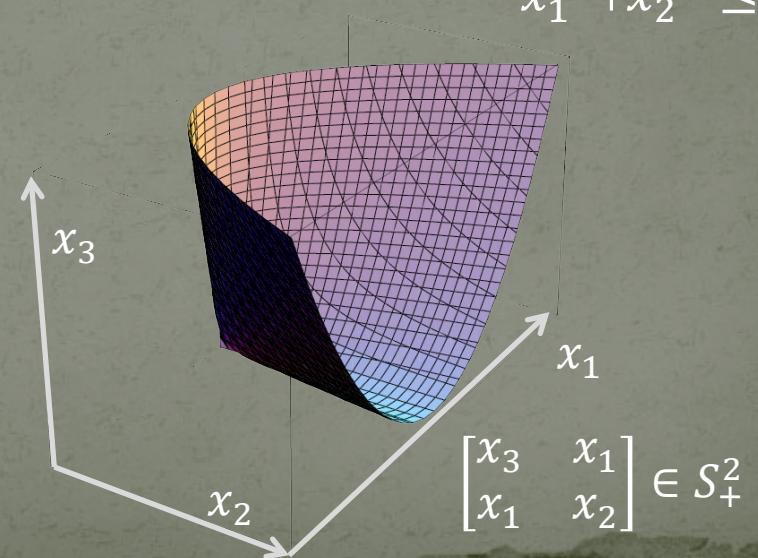
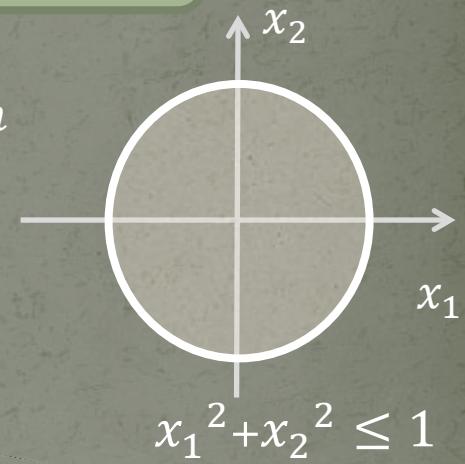
Objective Function

$$\text{minimize } f(x) \quad \text{subject to} \quad x \in C, C \subseteq \mathbb{R}^n$$



约束条件 $g_i(x) = 0$

Constraints $h_i(x) \leq 0$



优化问题

然而，‘一般性’的优化问题是不可解的! [Nesterov ‘04]

minimize $f(x)$ subject to $\{x \in \mathbf{R}^n : 0 \leq x_i \leq 1, i = 1, \dots, n\}$

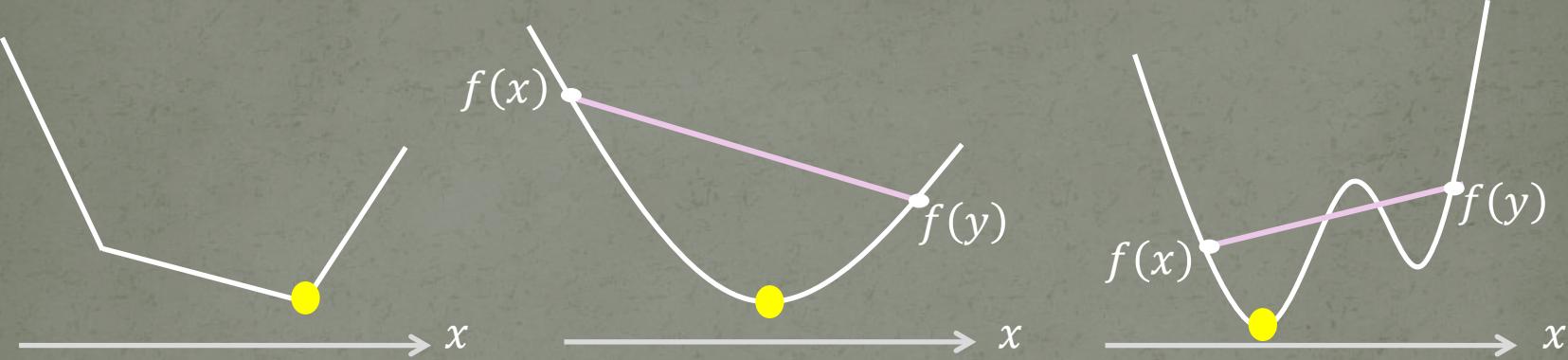
Lipschitz 条件: $|f(x) - f(y)| \leq L|x - y|_\infty, \forall x, y \in C$



设 $L=2$ 、 $n=10$ ，求误差不超过 $\epsilon=1\%$ 的最优解的算法复杂度为

$$\left[\frac{L}{2\epsilon}\right]^n = 10^{20}, \text{ 运行时间} > 10000 \text{ 年! ?}$$

优化问题



线性优化

1984年以前...

凸优化

1984年，Karmarkar 突破性地提出了多项式时间的Interior Point优化算法，‘解决’了凸优化问题。

非凸优化

尚无稳定、有效的优化算法...

[Karmarkar'84, SCI: 1314]

优化问题

“The great watershed in optimization isn't between linearity and non-linearity, but **convexity** and **non-convexity**”

[Rockafellar '93]

凸优化

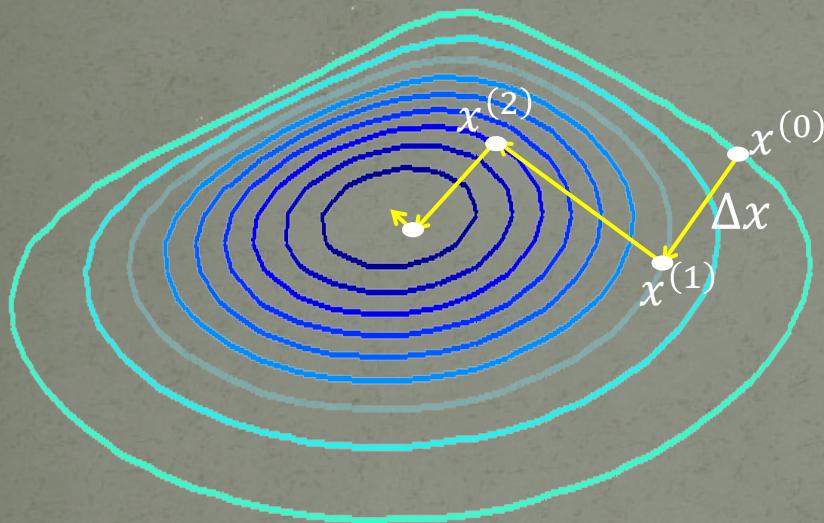
- 保证可以找到全局最优解
- 优化算法**稳定高效**
- 优化结果与初始值无关
- 对问题的建模较不精确

非凸优化

- 基本无法找到全局最优解
- 优化算法因人而异
- 优化结果依赖初始值
- 对问题的**建模精确**

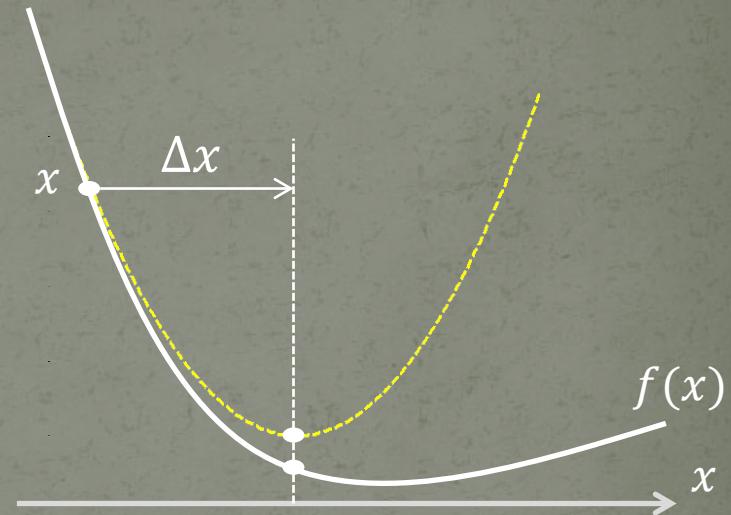
优化问题

凸优化基本算法



梯度下降法

$$\Delta x = -\nabla f(x)$$



牛顿下山法

$$\Delta x = -\nabla^2 f(x)^{-1} \nabla f(x)$$

最陡下降法

$$\Delta x = \operatorname{argmin}\{\nabla f(x)^T v \mid \|v\| \leq 1\}$$

... ...

优化问题

Interior Point Method

$$\text{minimize } f(x) \quad \text{subject to} \quad h_i(x) \leq 0$$



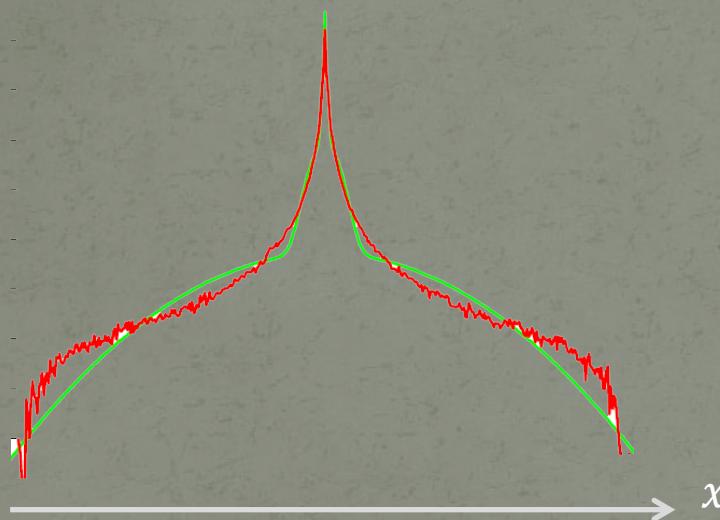
$$\text{minimize } f(x) - \lambda \sum_i \log(-h_i(x))$$

32 iterations is all that you need ...

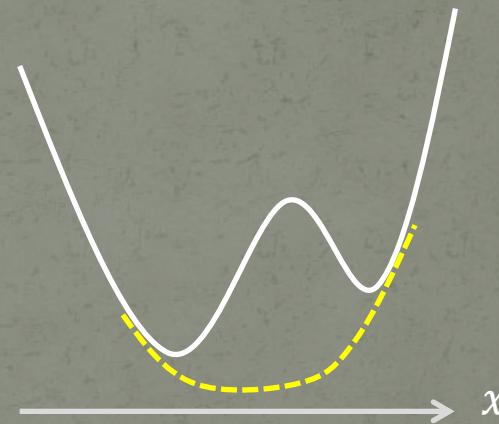
优化问题

非凸优化

凸优化算法 + 启发式搜索 + 运气



分段凸逼近
[Fergus'06]



局部凸松弛
[Kahl'08]

优化问题

大多数图像认知问题都可转化为优化问题 ...



$\text{minimize } ??$
 $\text{subject to } ???$

- 优化问题
- 总变差方法 (Total Variation Methods)
- 主动形状模型 (Active Shape Model)
- 图割图像分割 (Graph Cut Image Segmentation)
- 人脸图像光照变换
- 焊缝检测
- 多生物特征融合

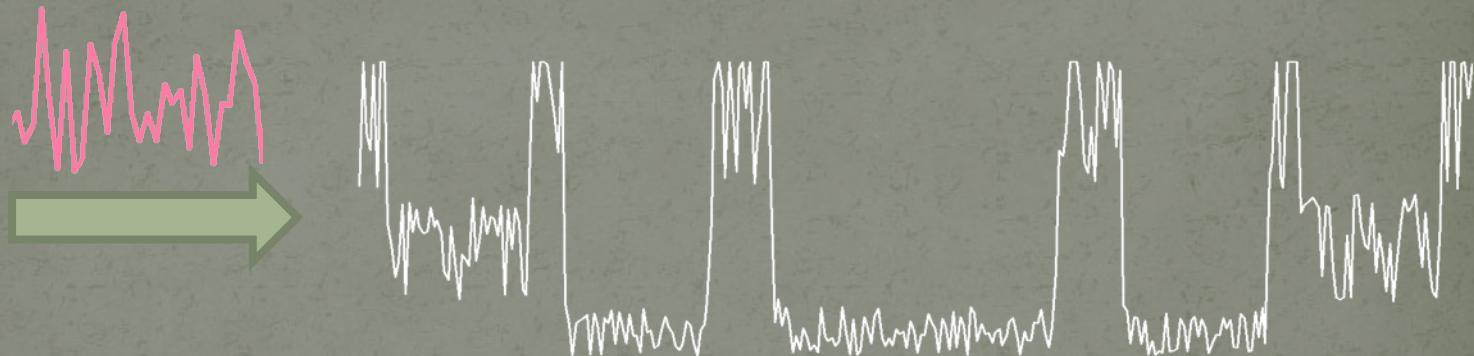
总变差方法

?



总变差方法

?

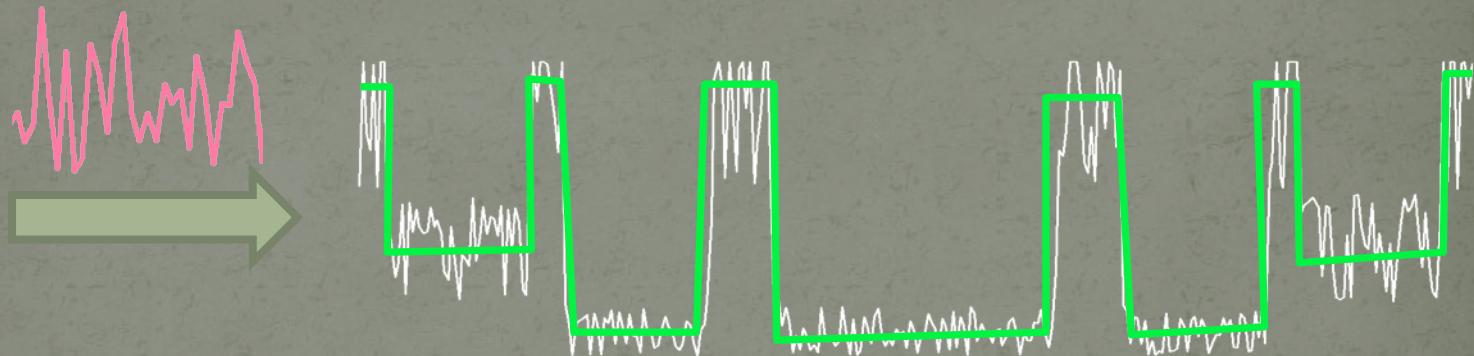


低通滤波器

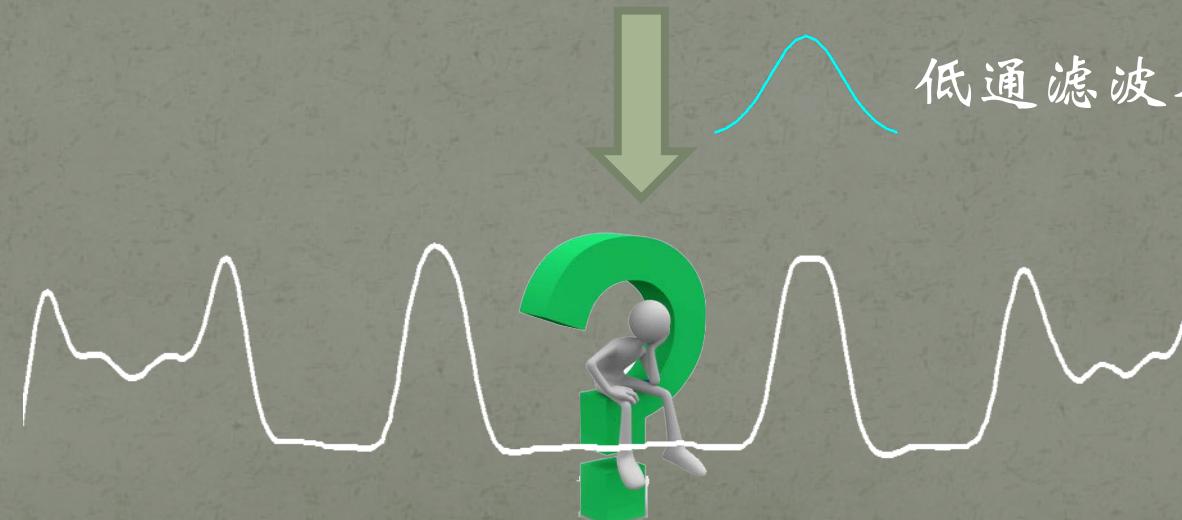


总变差方法

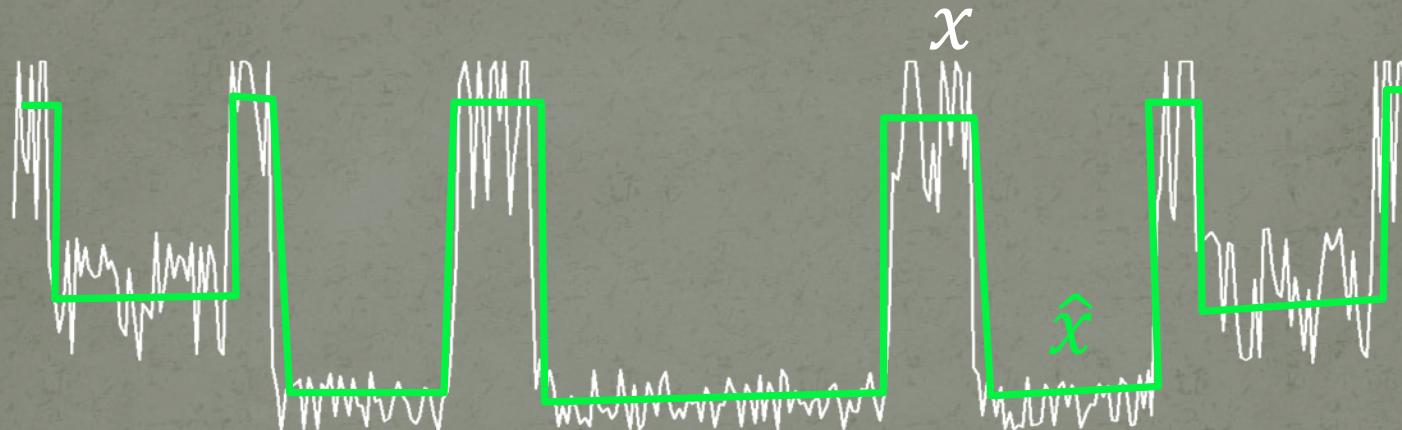
?



低通滤波器



总变差方法



认知一： \hat{x} 和 x 非常接近： $\|\hat{x} - x\|^2 \downarrow$

认知二： \hat{x} 大多数区域平坦： $\int |\nabla \hat{x}| d\hat{x} \downarrow$

Total Variation

总变差方法

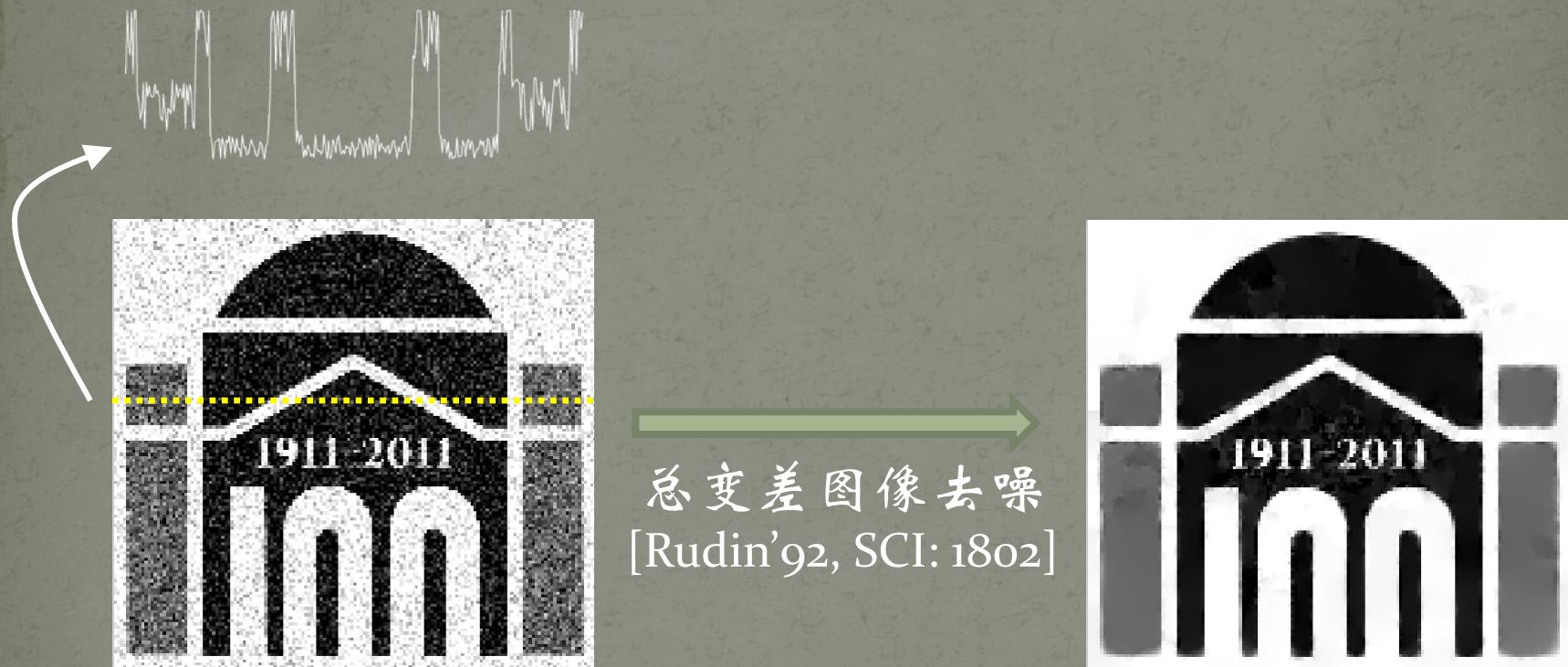
$$\text{minimize} \quad \frac{1}{2} \|\hat{x} - x\|^2 + \int |\nabla \hat{x}| d\hat{x}$$

离散形式: $\text{minimize} \quad \frac{1}{2} \sum_i \|\hat{x}_i - x_i\|^2 + \sum_i |\hat{x}_{i+1} - \hat{x}_i|$

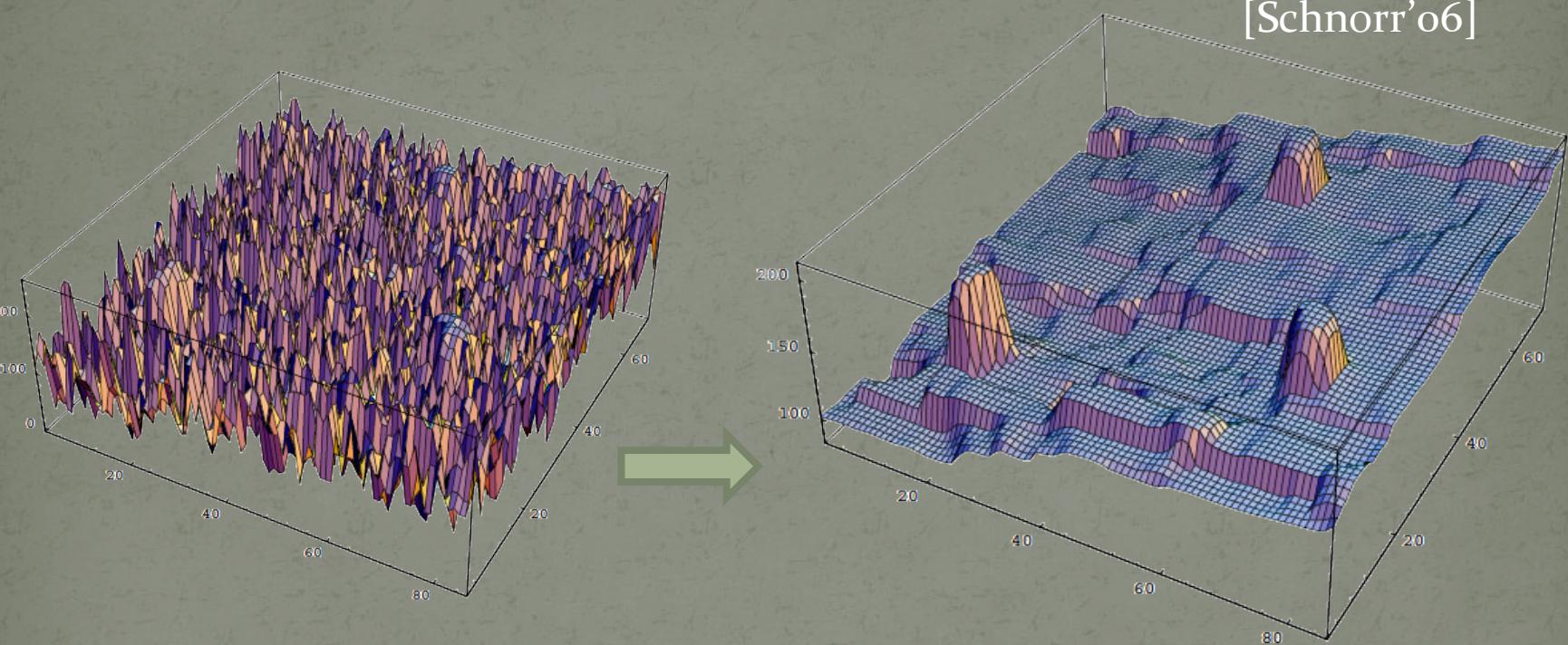
It is convex!



总变差方法

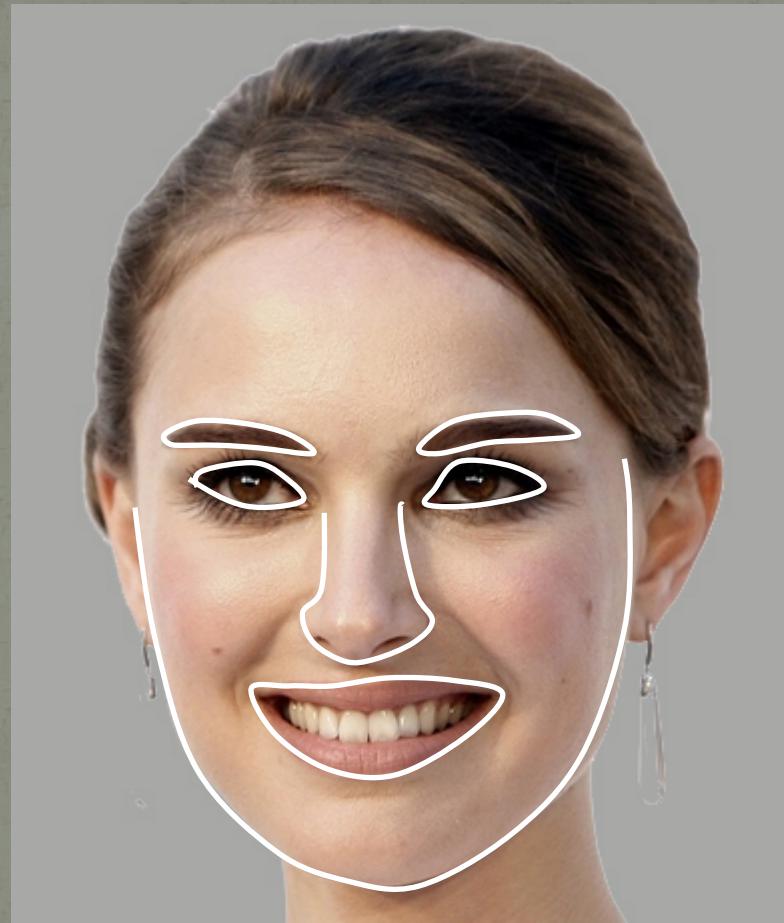


总变差方法



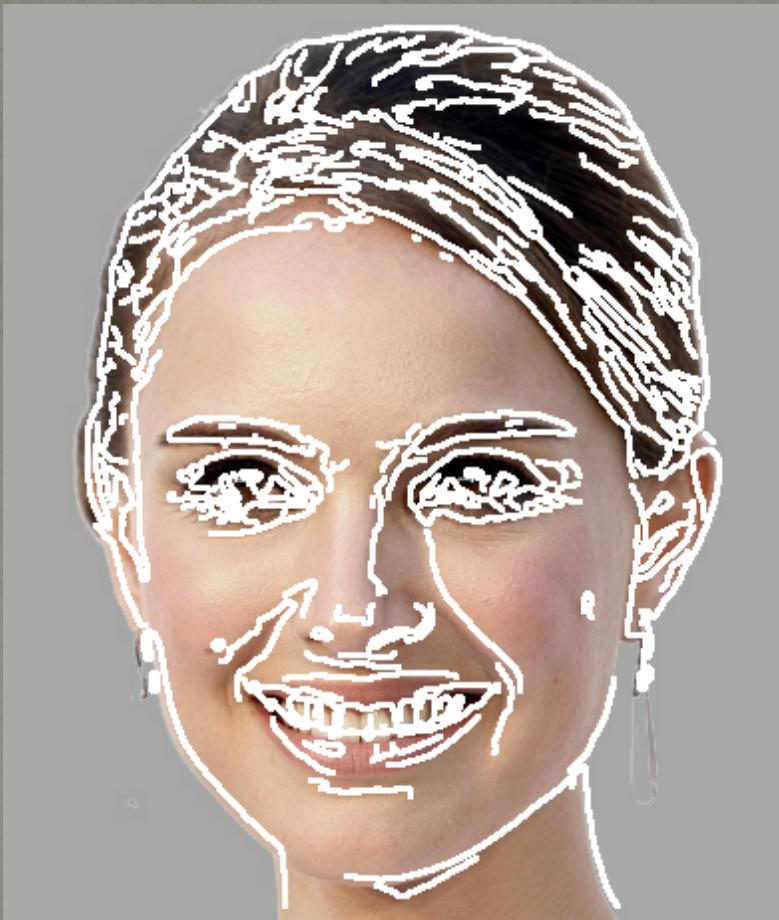
目前，总变差方法被广泛应用于图像去噪、
图像重建、图像分割等问题中。

主动形状模型



勾勒出人脸及五官形状？

主动形状模型

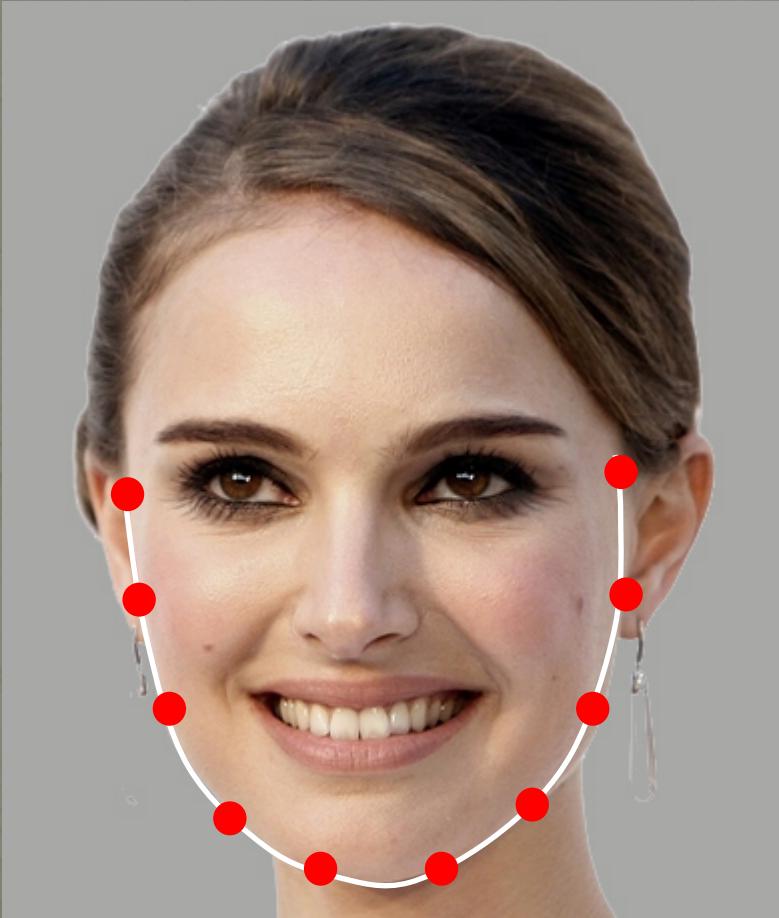


自动勾勒形状

边缘检测?



主动形状模型

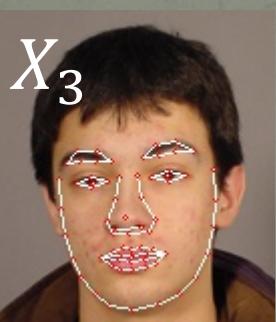
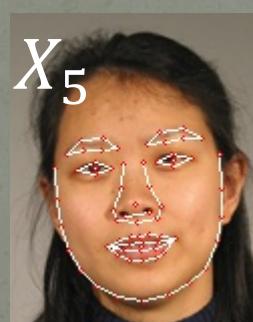
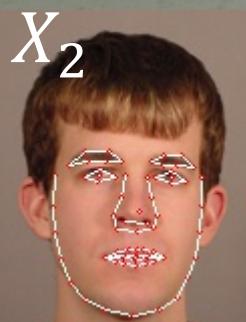
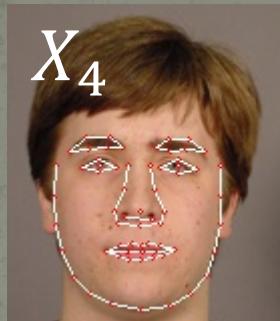
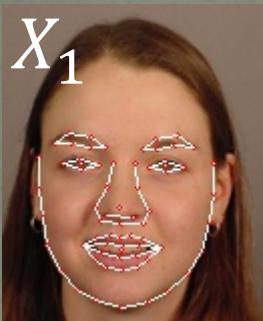


认知一：人脸形状大体上相似（形状模型）。

用一个点序列表示形状。

$$X = [x_1, x_2, \dots, x_n, y_1, y_2, \dots, y_n]^T$$

主动形状模型



⋮

形状模型

在一系列（几百）人脸图像上手工标注形状，构成形状训练集合： X_1, X_2, \dots, X_m

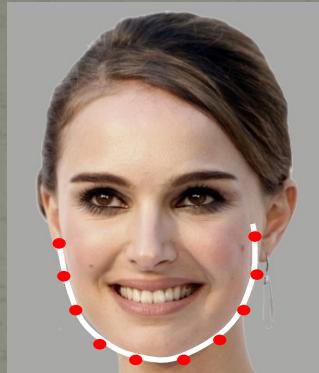
以协方差矩阵
特征向量表示

$$X = \text{均值}(X_i) + b \cdot \text{方差}(X_i)$$

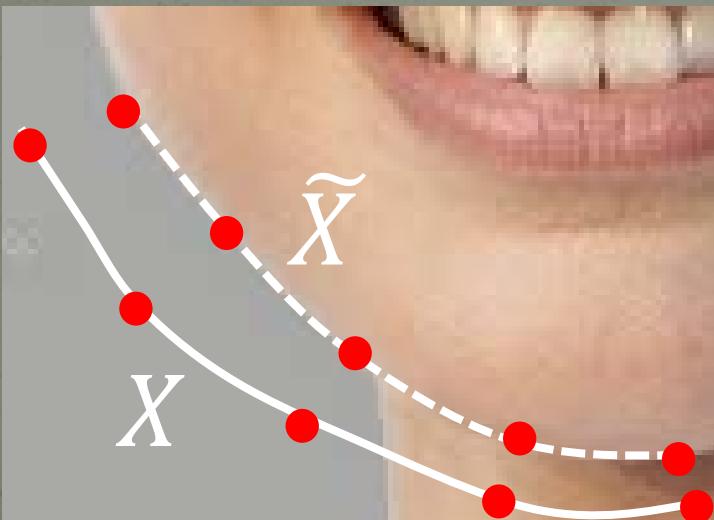
模型参数

$$|b_i| \leq 3\sqrt{\lambda_i}$$

主动形状模型



认知二：形状点应在人脸图像的边缘附近。



$$\begin{aligned} & \text{minimize } \|X(b) - \tilde{X}\|^2 \\ & \text{subject to } |b_i| \leq 3\sqrt{\lambda_i} \end{aligned}$$

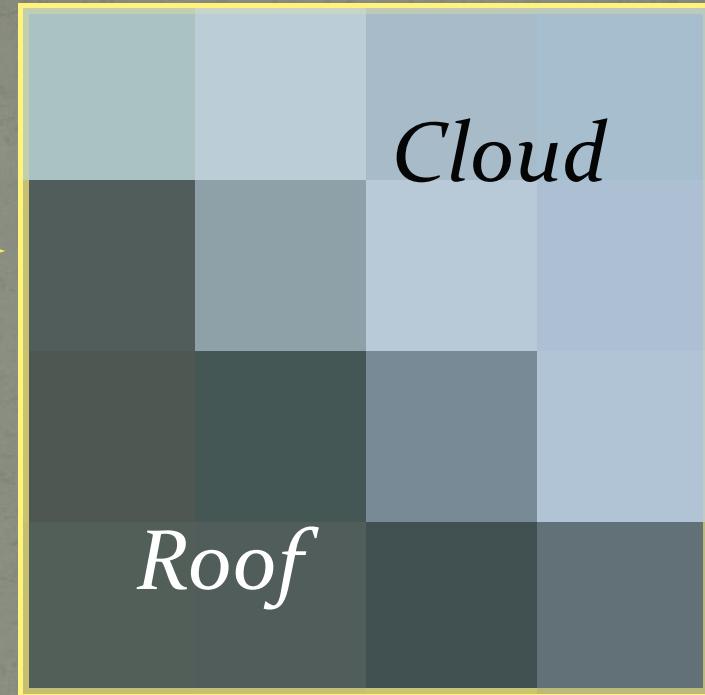
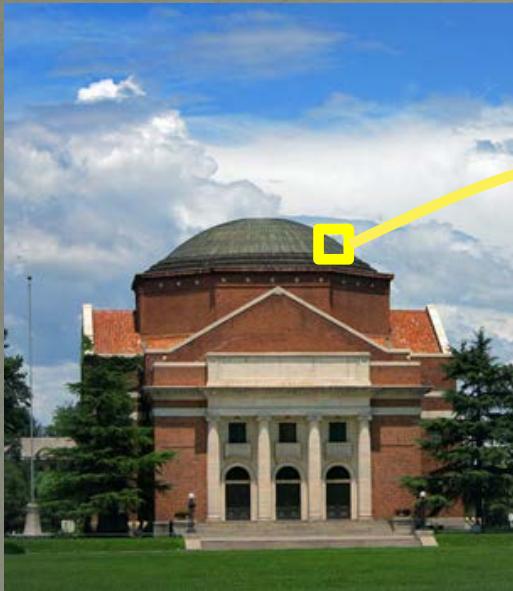
[Cootes'95, SCI: 1831]

图割图像分割



图像分割

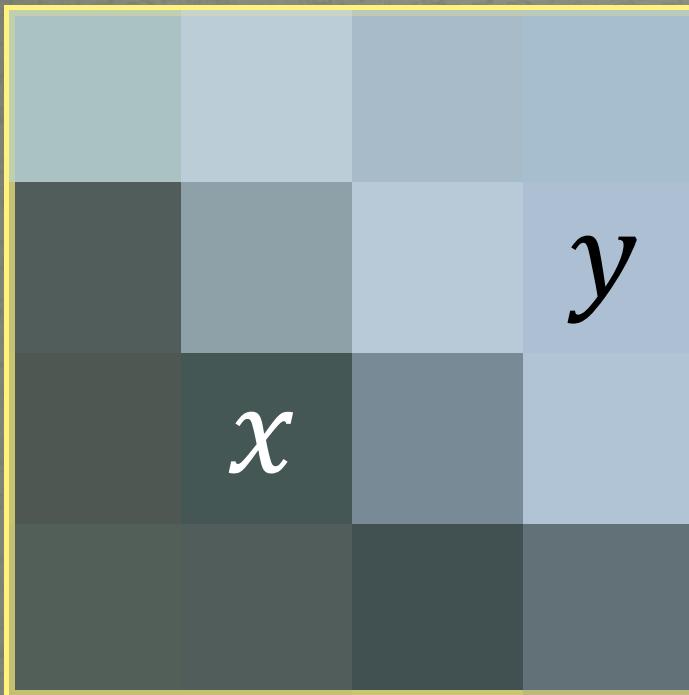
图割图像分割



认知一：颜色(亮度)接近的像素属同一物体

认知二：空间位置相近的像素属同一物体

图割图像分割



定义任意两像素间的关联度

- 颜色关联度 (Affinity)

$$aff_C = \exp(-\|c(x) - c(y)\|/2\sigma_c^2)$$

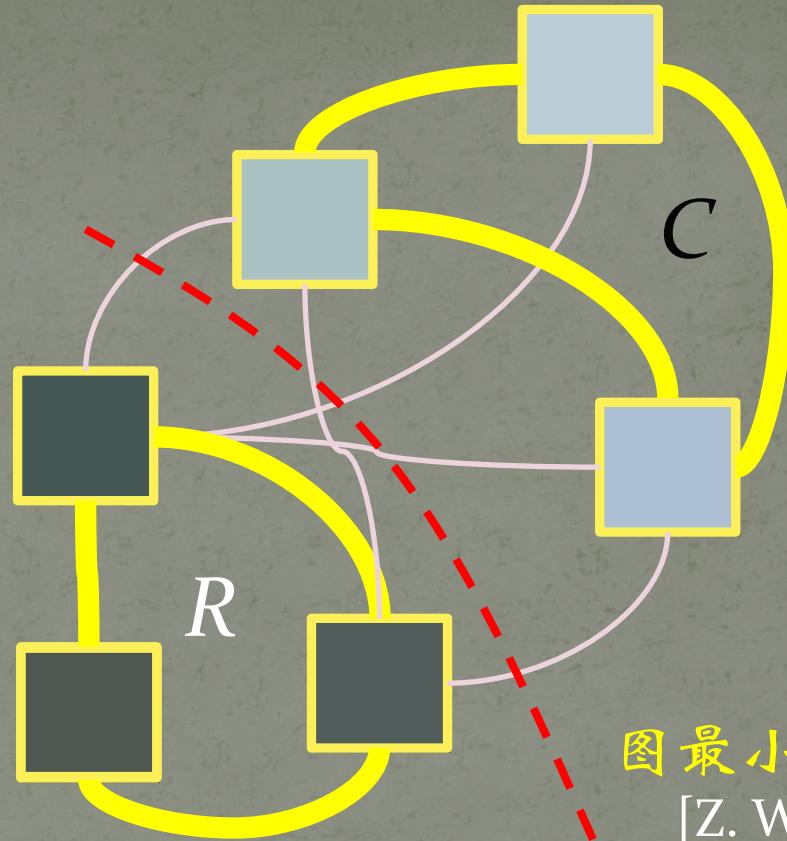
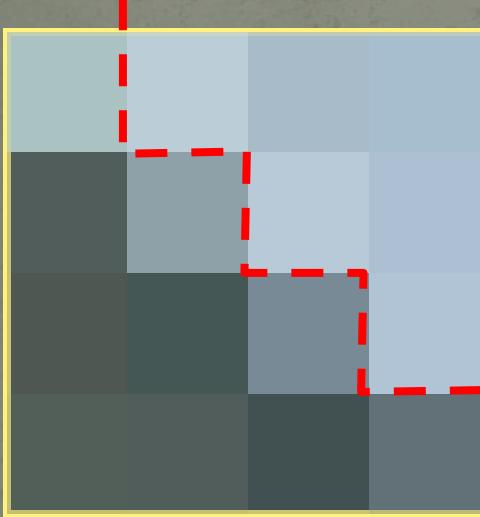
- 距离关联度

$$aff_D = \exp(-\|x - y\|/2\sigma_D^2)$$

- 总关联度

$$aff(x, y) = aff_C + \lambda * aff_D$$

图割图像分割



图最小割问题

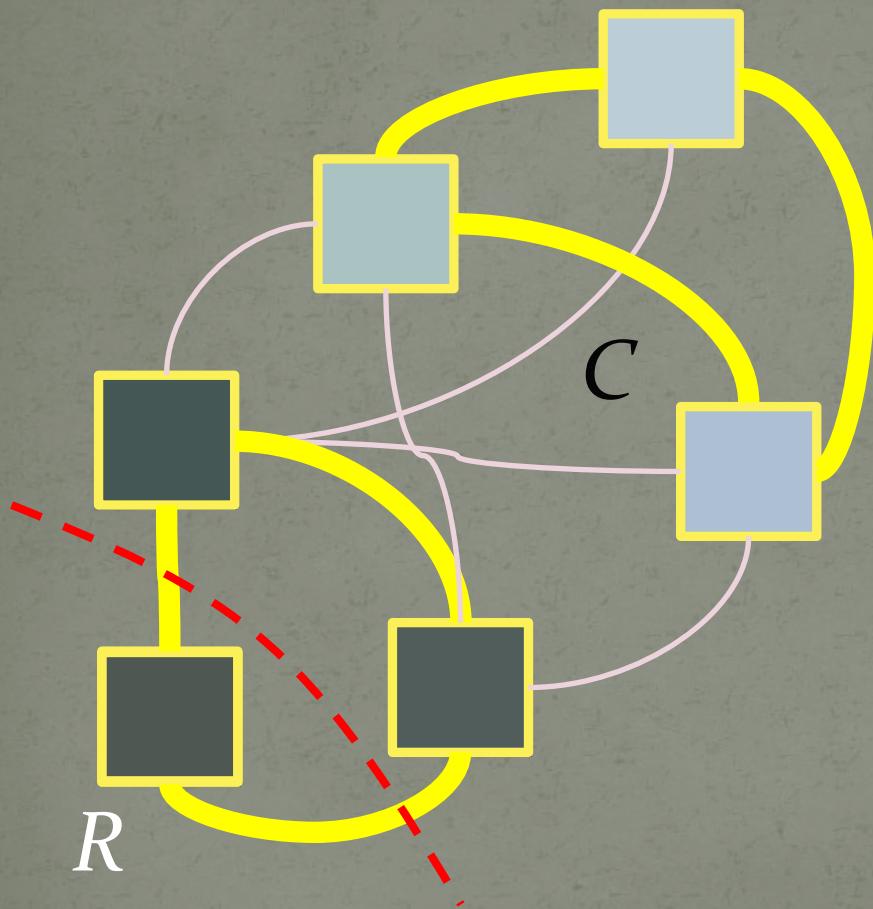
[Z. Wu '93]

$$\text{minimize } \text{cut}(R, C) = \sum_{\substack{x \in R \\ y \in C}} \omega(x, y)$$

多项式时间算法 [Ford & Fulkerson' 62]

构造一个图(Graph) V :
每个像素为顶点，两
像素间路径权重(ω)等
于两像素间的关联度。

图割图像分割



Normalized Cuts:

$$Ncut(R, C) = \frac{cut(R, C)}{aff(R, V)} + \frac{cut(R, C)}{aff(C, V)}$$

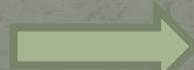
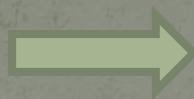
[J. Shi '00, SCI: 1488]

More Examples ...

- Low-rank matrix approximation using L₁ norm
[Eriksson'10, CVPR Best Paper]
- Non-rigid surface detection [Pilet'05, CVPR Best Paper]
- Graph cut based inference with co-occurrence statistics
[Ladicky'10, ECCV Best Paper]
- LMI relaxation in geometric reconstruction
[Kahl'05, ICCV Best Paper]
- Compressive Sensing (L₁ Optimization)
- Local Linear Embedding (Quadratic Optimization)
- Support Vector Machine (Quadratic Optimization)
-

- 优化问题
- 总变差方法 (Total Variation Methods)
- 主动形状模型 (Active Shape Model)
- 图割图像分割 (Graph Cut Image Segmentation)
- 人脸图像光照变换
- 焊缝检测
- 多生物特征融合

人脸图像光照变换



人脸图像光照变换



比值图像 R



人脸图像光照变换

认知一：图像 **总体** 的亮度变化由比值图像决定

认知二：**局部** 的亮度变化平缓

$$\text{minimize} \sum_i \left(\sum_{j \in \omega_i} (\hat{I}(j) - \alpha_i I(j))^2 + \lambda(\alpha_i - R_i)^2 \right) \quad [\text{Chen'10}]$$

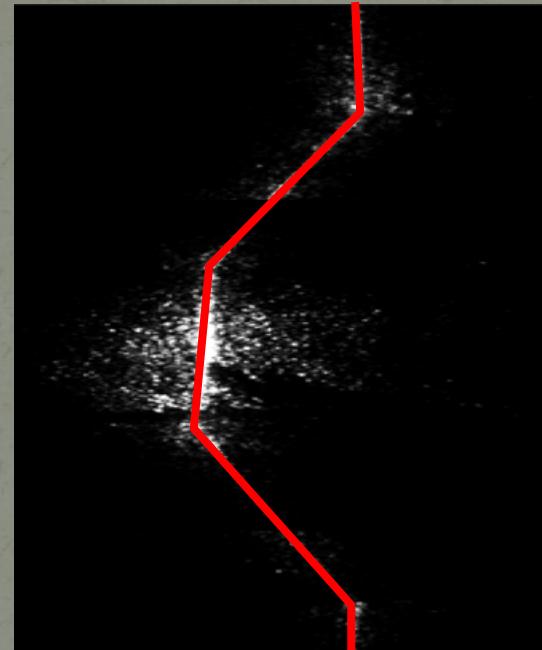


I



\hat{I}

焊缝检测

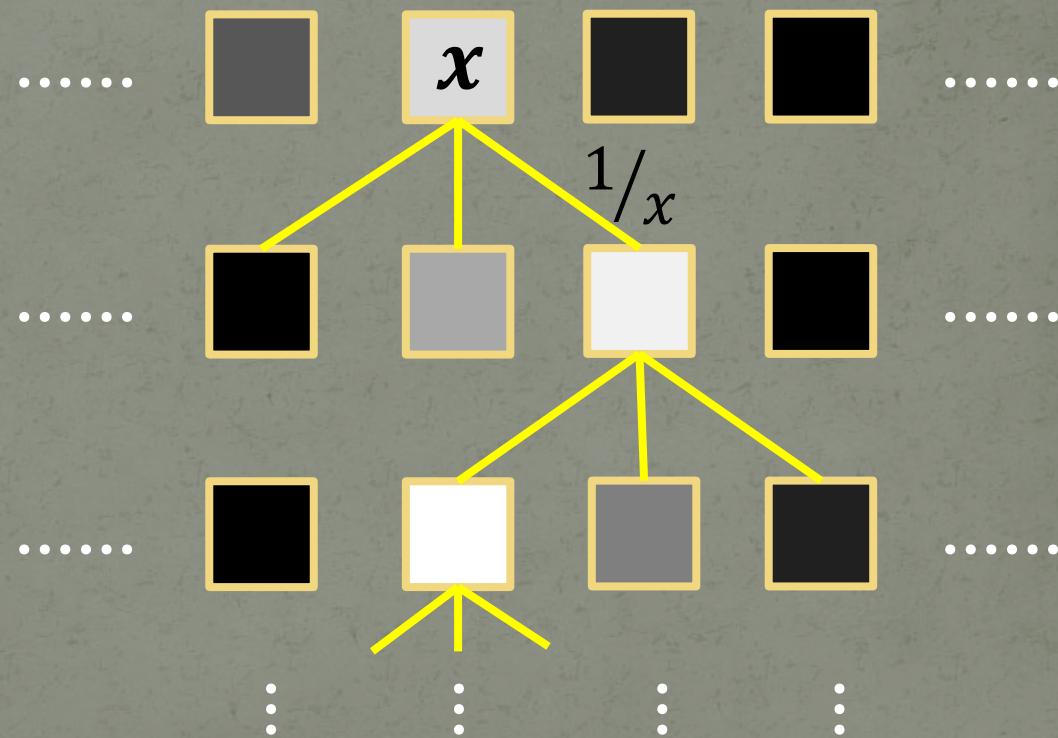
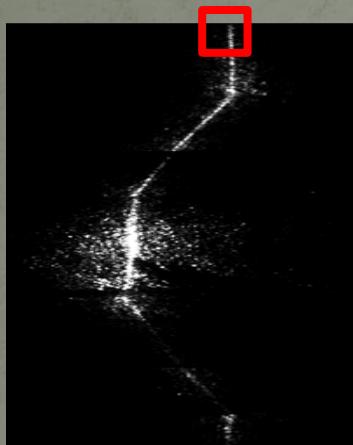


认知一：连续的单像素构成的路径

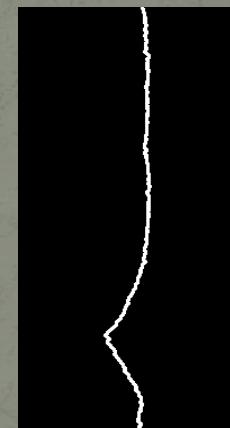
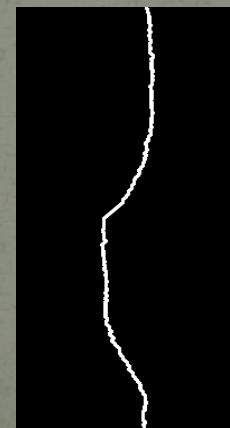
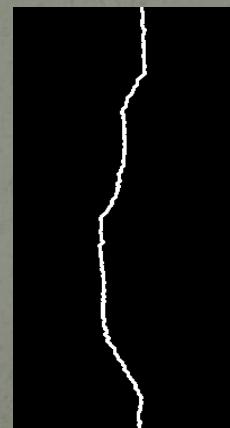
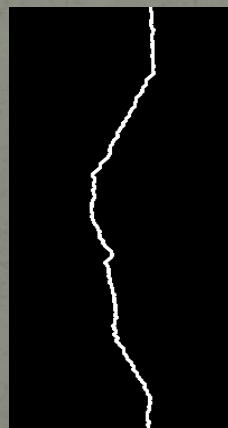
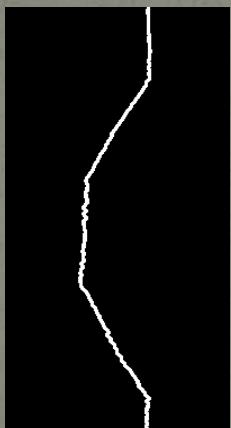
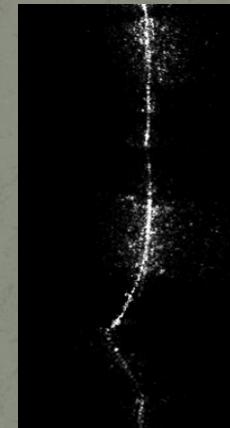
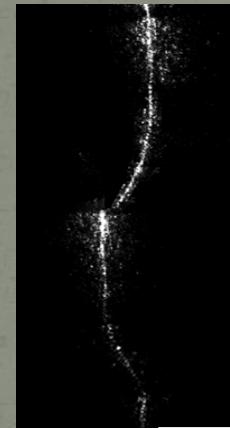
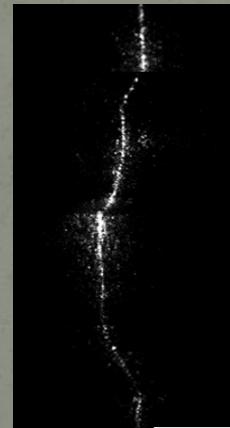
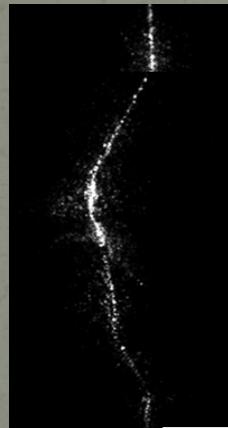
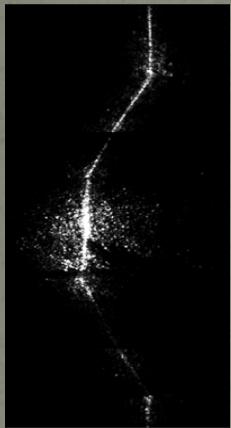
认知二：尽量通过高亮的区域

焊缝检测

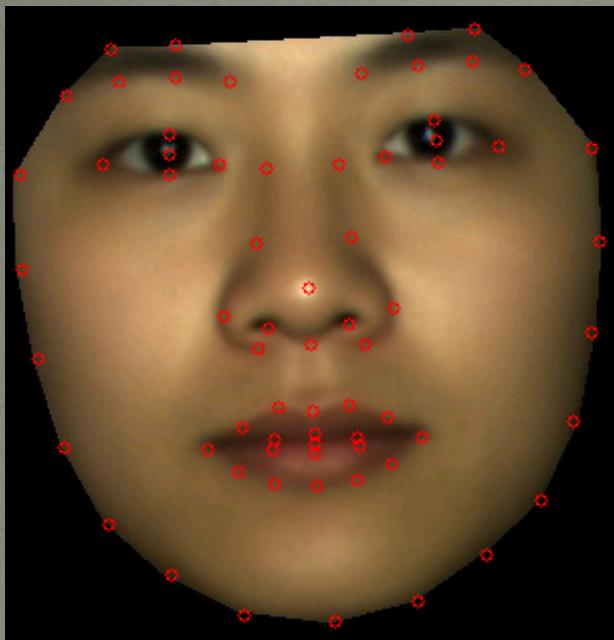
图最短路径 [Dijkstra'59]



焊缝检测



多生物特征融合

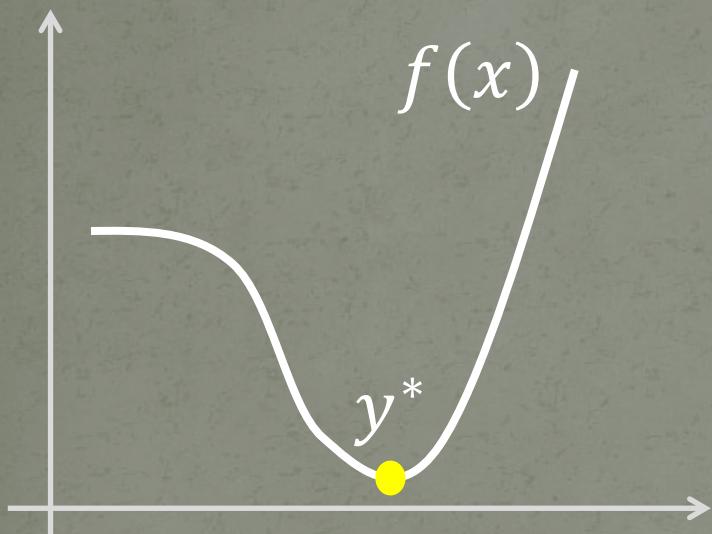


设定局部特征的权重提高识别率?

$$\text{minimize } \sigma_G / (\mu_G - \Psi(I^T * \omega, r) / r)$$

$$\text{subject to } \sum_{i=1}^N \omega_i = 1$$

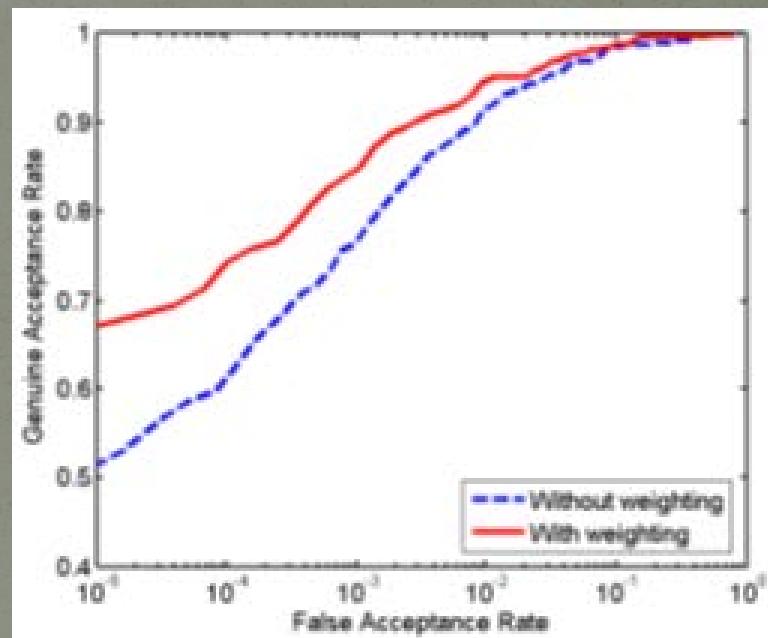
多生物特征融合



拟凸优化二分算法

- given $y_l \leq y^*, y_u \geq y^*$
- do
 - $t = (x_l + x_u)/2$
 - minimize 0 s.t. $f(x) \leq t$
 - if succeed $y_u = t$
 - else $y_l = t$
- while $y_u - y_l > \varepsilon$

多生物特征融合



Everything is not optimization ...

Thanks ↵

参考文献

- [Nesterov '04] *Introductory lectures on convex optimization : A Basic course*
- [Karmarkar'84] “A new polynomial-time algorithm for linear programming”
- [Fergus'06] “Removing camera shake from a single photograph”
- [Kahl'08] “Practical global optimization for multi-view geometry”
- [Schnorr'06] “Convex and Non-Convex Optimization”
- [Rudin'92] “Nonlinear total variation based noise removal algorithms”
- [Cootes'01] “Active shape models-their training and application”
- [Z. Wu'93] “An Optimal Graph Theoretic Approach to Data Clustering: Theory and Its Application to Image Segmentation”
- [Ford & Fulkerson'62] *Flows in Networks*
- [J. Shi'00] “Normalized Cuts and Image Segmentation”

参考文献

- [Erriksson'10] “Efficient Computation of Robust Low-Rank Matrix Approximations in the Presence of Missing Data using the L₁ Norm”
- [Pilet'05] “Real-time non-rigid surface detection”
- [Ladicky'10] “Graph cut based inference with co-occurrence statistics”
- [Kahl'05] “Globally optimal estimates for geometric reconstruction problems”
- [Donoho'06] “Compressive Sensing”
- [Roweis & Saul'00] “Nonlinear Dimensionality Reduction by Locally Linear Embedding”
- [Cortes & Vapnik'95] “Support-Vector Networks”
- [Dijkstra'59] “A note on two problems in connection with graphs”
- [Chen'10] “Face Image Relighting using Locally Constrained Global Optimization”

参考文献

- [Gong'11] “Quasi-convex Optimization of Metrics in Biometric Score Fusion”
- [Xia'11] “Segmenting the Subthalamic Nucleus Using Narrow Band Limited Variational Level Set Method”
- [Chen'12] “Robust Welding Seam Tracking using Image Seam Extraction”