

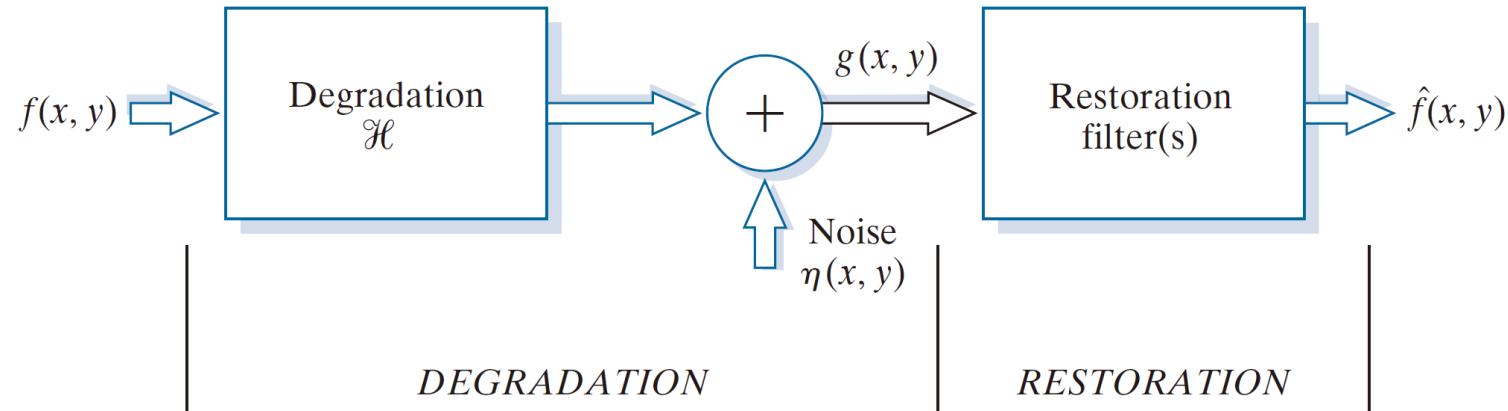
主要内容

- 图像退化/复原过程的模型
- 噪声模型
- 空间域滤波方法
- 频率域滤波方法
- 退化函数的估计
- 逆滤波
- 维纳滤波

5.3 只存在噪声的复原 —— 空间滤波

FIGURE 5.1

A model of the image degradation/restoration process.



$$g(x, y) = f(x, y) \star h(x, y) + \eta(x, y)$$

$$G(u, v) = F(u, v)H(u, v) + N(u, v)$$

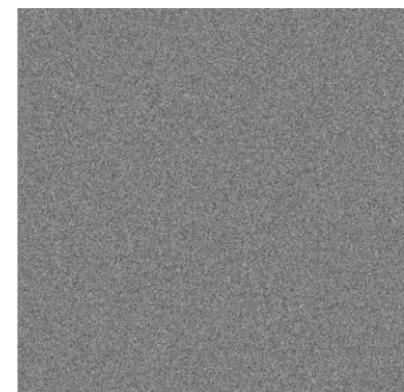
$$g(x, y) = f(x, y) + \eta(x, y)$$

$$G(u, v) = F(u, v) + N(u, v)$$

加性噪声

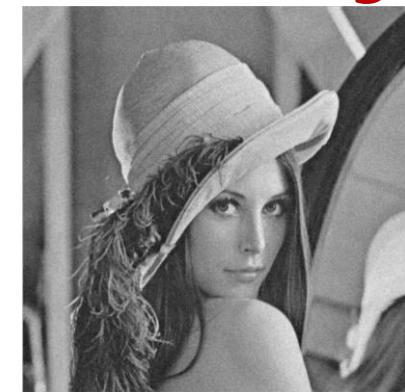


$f(x, y)$



$\eta(x, y)$

• 去噪 denoising

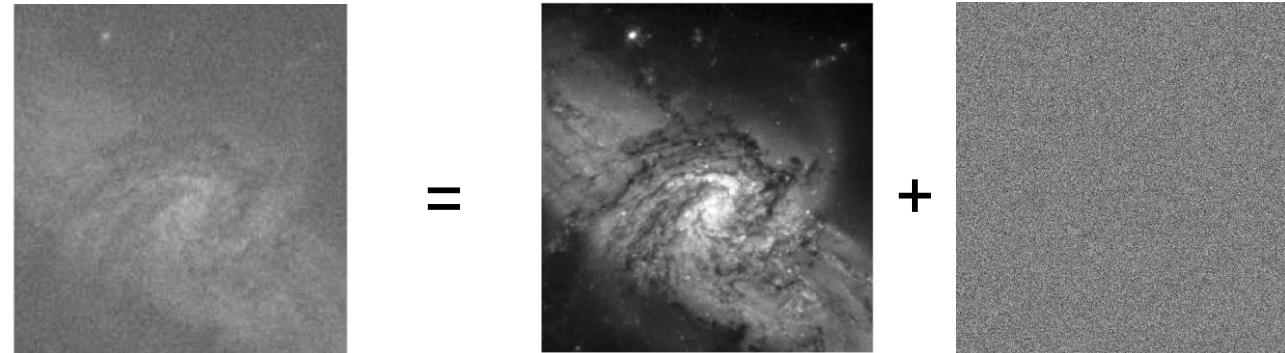


$\hat{f}(x, y)$

图像降噪 —— 噪声图像相加（平均）

$$g(x, y) = f(x, y) + \eta(x, y)$$

退化的图像 理想图像 噪声

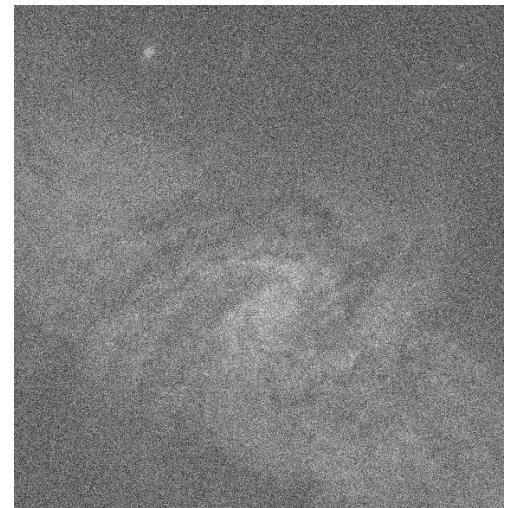


假设噪声服从零均值的高斯分布 $\eta(x, y) \sim N(0, \sigma_{\eta(x, y)}^2)$

输入同一物体的 K 副图像 $g_i(x, y), i = 1, 2, \dots, K$

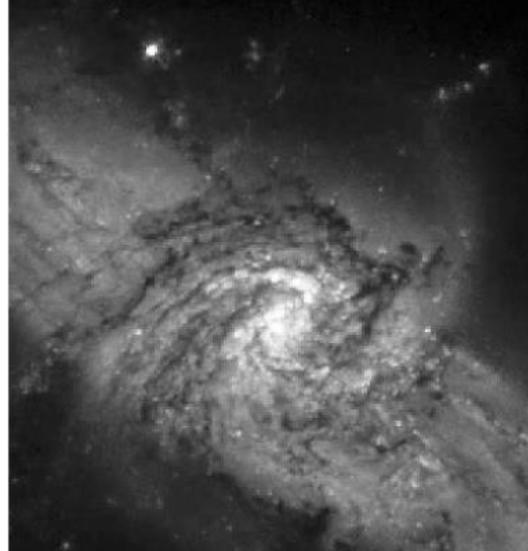
输出图像为 $\bar{g}(x, y) = \frac{1}{K} \sum_{i=1}^K g_i(x, y)$

期望 $E\{\bar{g}(x, y)\} = f(x, y)$ 方差 $\sigma_{\bar{g}(x, y)}^2 = \frac{1}{K} \sigma_{\eta(x, y)}^2$



图像降噪

理想图像 (8 bit)



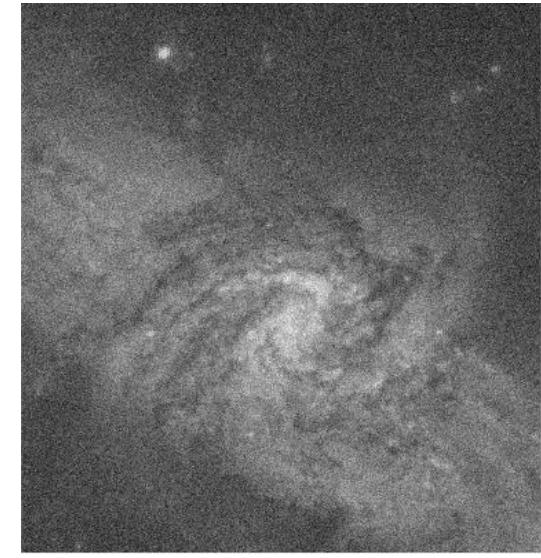
噪声图像 (8 bit)



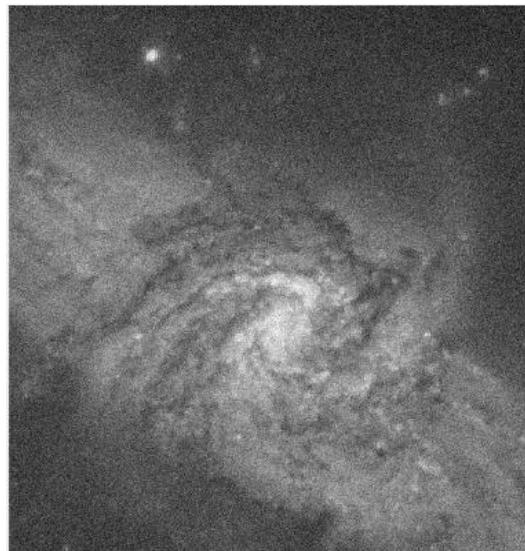
NEX=5



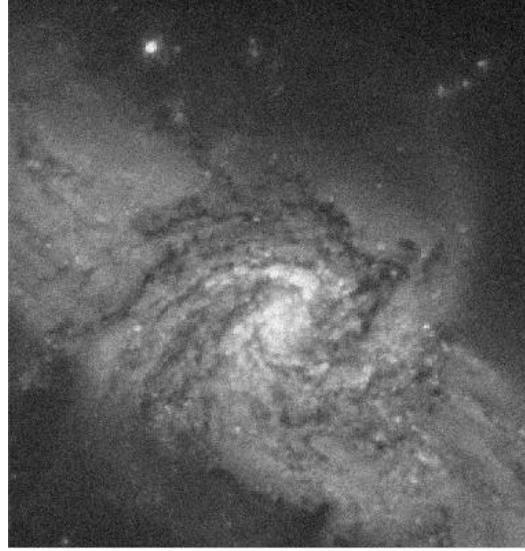
NEX=10



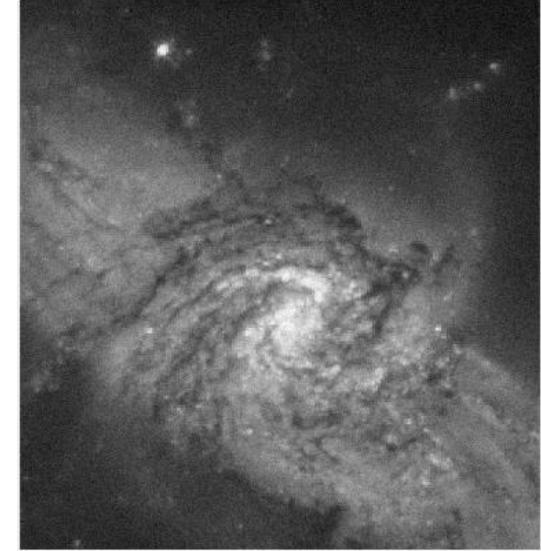
NEX=20



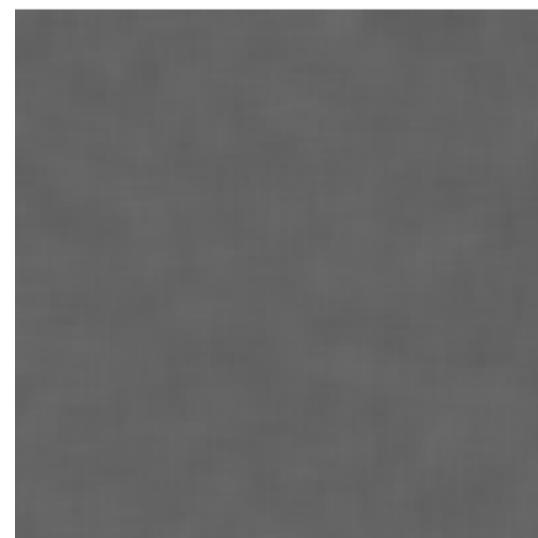
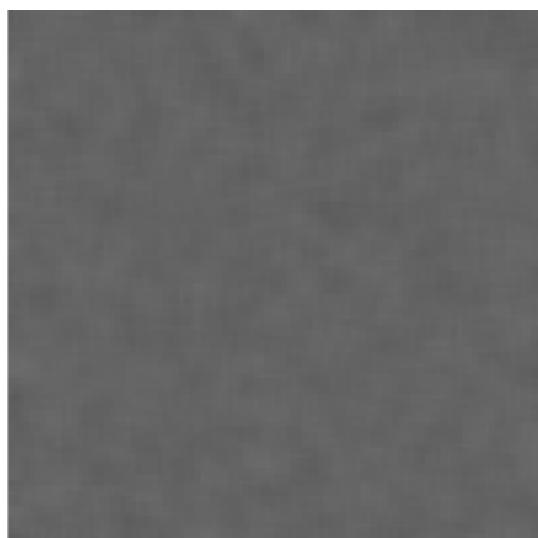
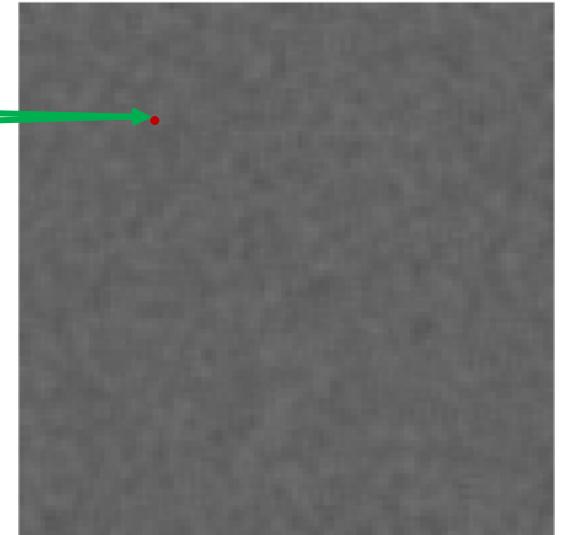
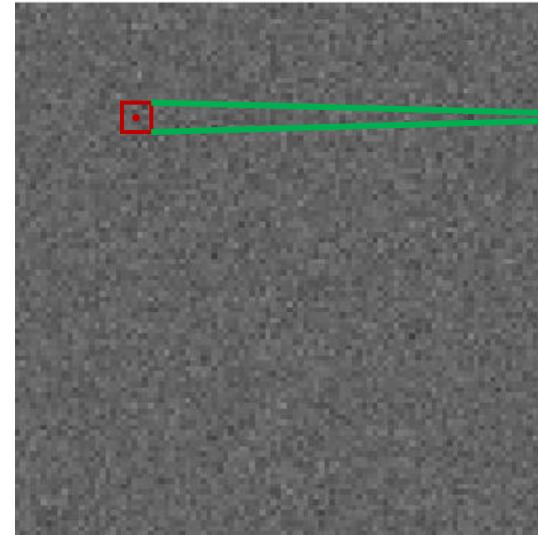
NEX=50

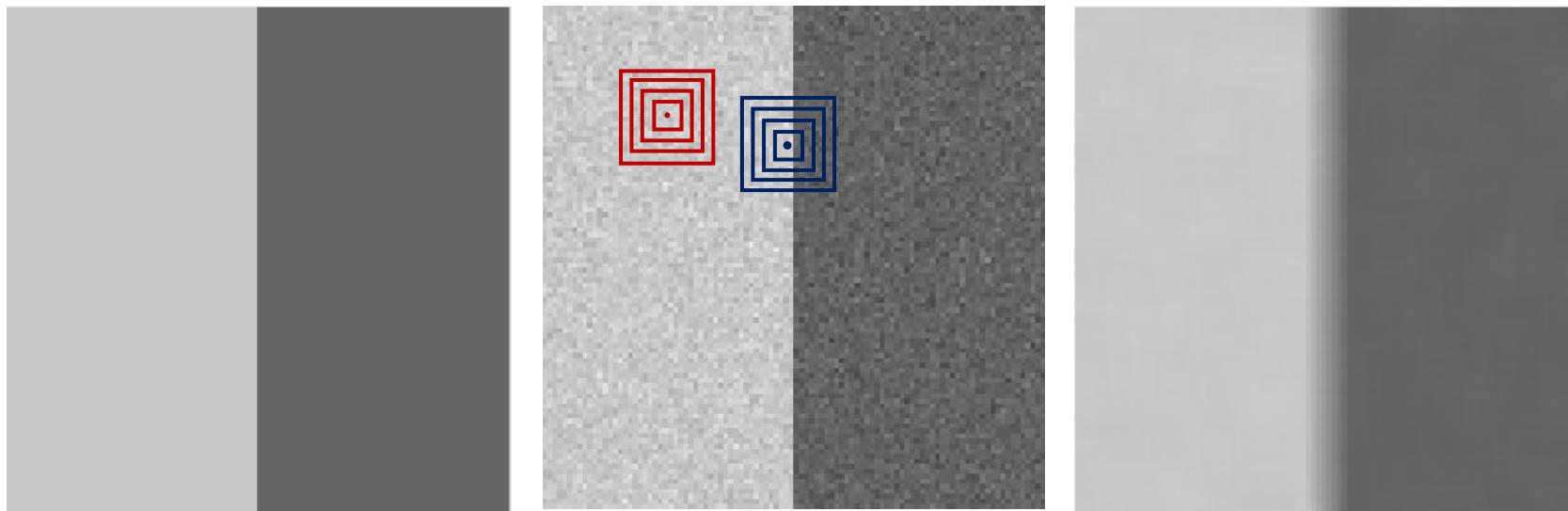


NEX=100



$I=100, \sigma = 10$



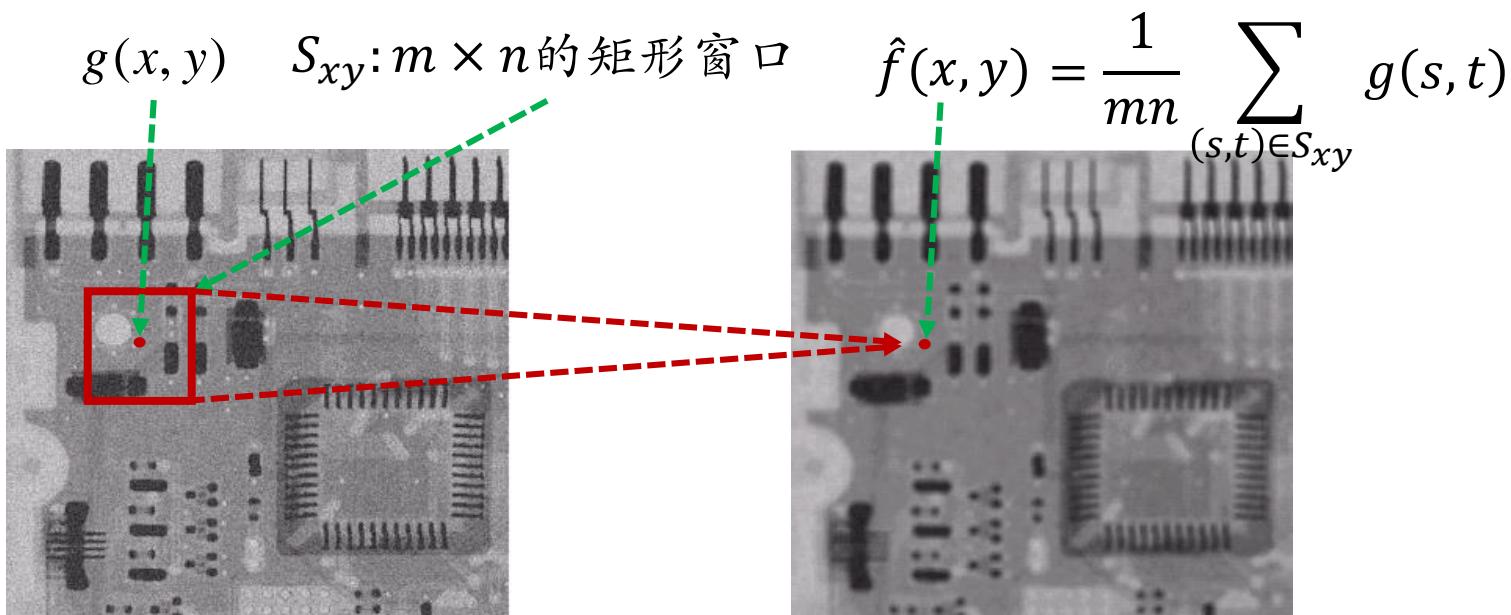


$\sigma = 10$

5.3 只存在噪声的复原 —— 空间滤波

• 均值滤波器

- 算术均值 $\hat{f}(x, y) = \frac{1}{mn} \sum_{(s,t) \in S_{xy}} g(s, t)$



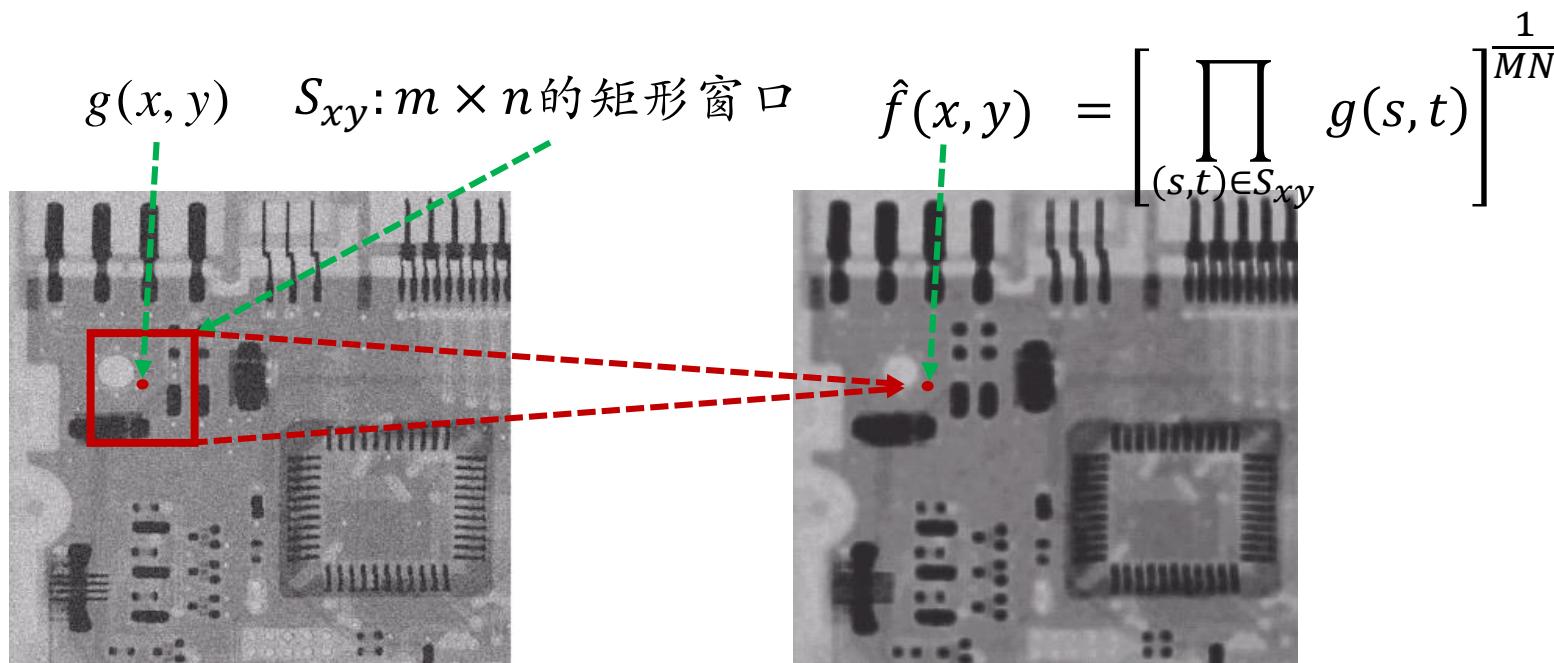
The filter size is 7×7

图像细节模糊
图像噪声降低

5.3 只存在噪声的复原 —— 空间滤波

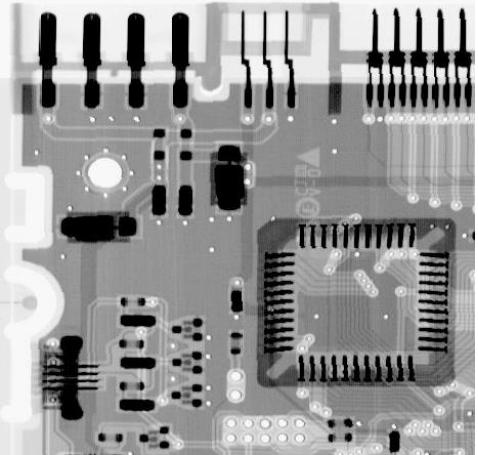
• 均值滤波器

• 几何均值 $\hat{f}(x, y) = \left[\prod_{(s,t) \in S_{xy}} g(s, t) \right]^{\frac{1}{MN}}$

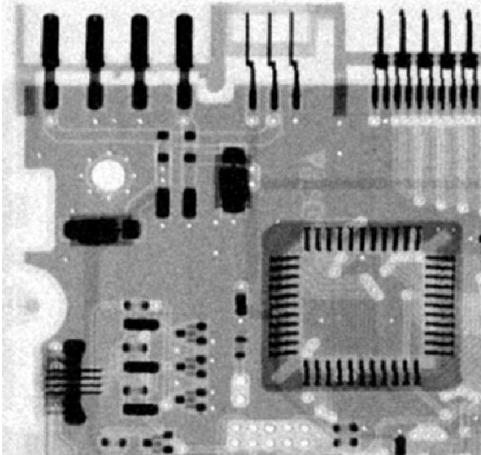


• 算数均值vs几何均值

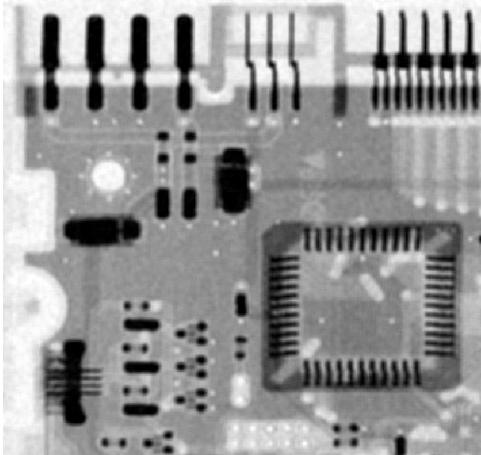
无噪图像



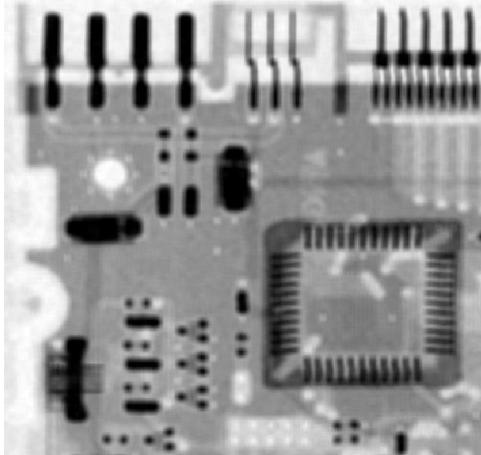
3×3



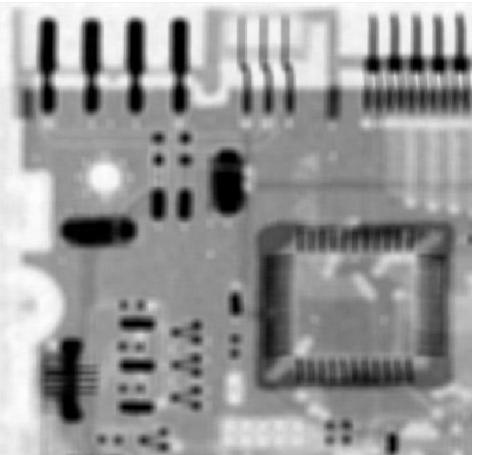
5×5



7×7

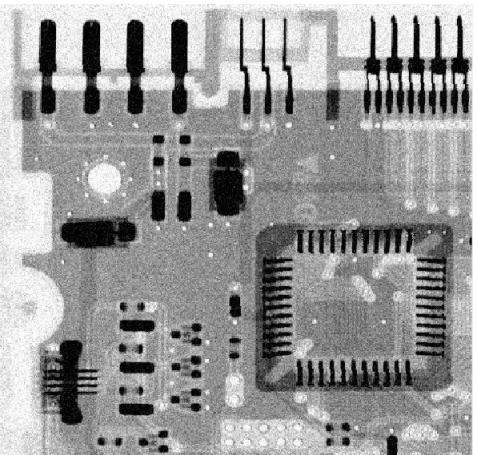


9×9

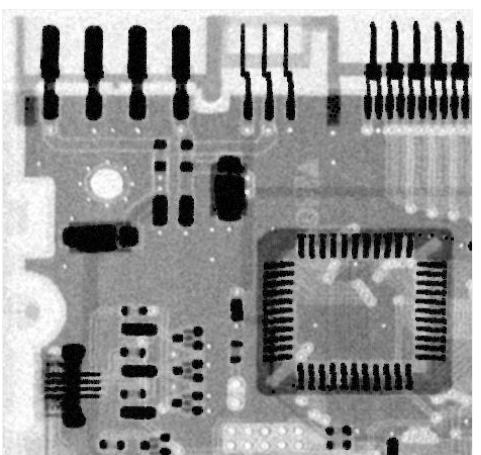


算数均值

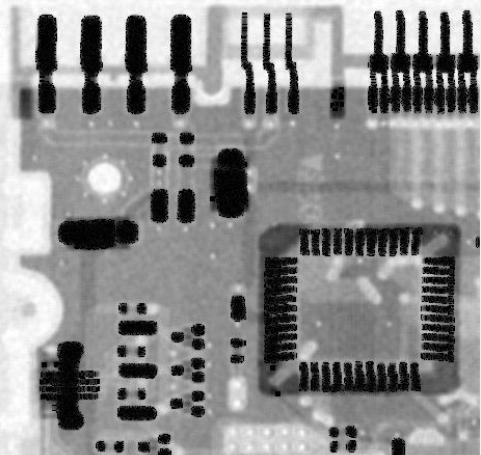
噪声图像



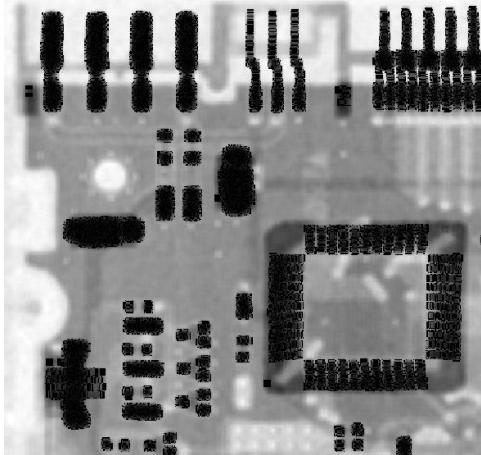
3×3



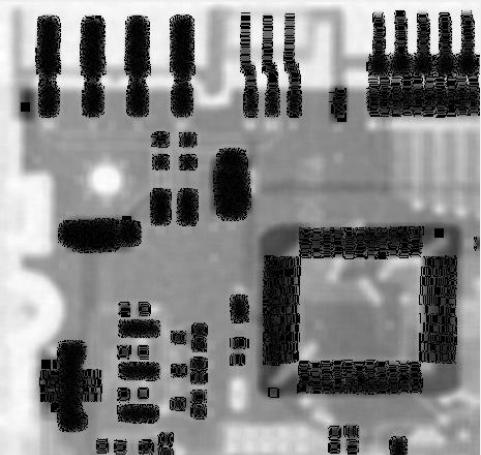
5×5



7×7



9×9



几何均值

5.3 只存在噪声的复原 —— 空间滤波

• 自适应均值滤波

$$\hat{f}(x, y) = w m_L + (1-w) g(x, y)$$

$$\hat{f}(x, y) = \frac{1}{mn} \sum_{(s,t) \in S_{xy}} g(s, t)$$

算术均值

$$w = \frac{\sigma_\eta^2}{\sigma_L^2}$$

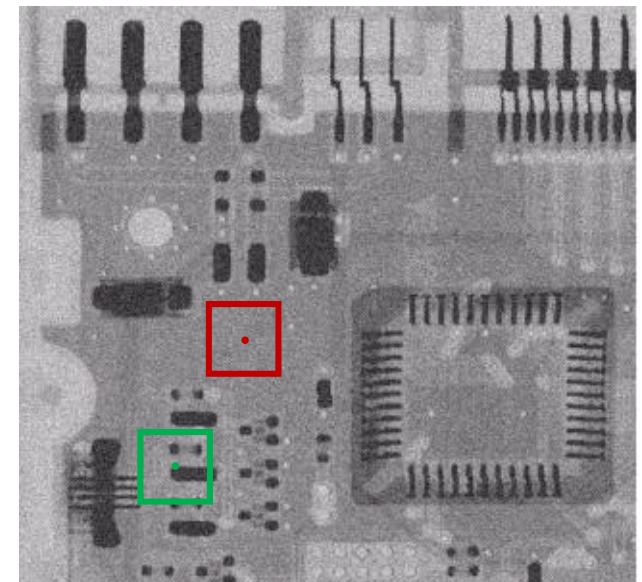
整幅图像像素值的方差
 σ_η^2 S_{xy} 内所有像素值的方差

$$\sigma_\eta^2 = \sigma_L^2 \quad \text{均匀区域} \quad \hat{f}(x, y) = m_L$$

分析

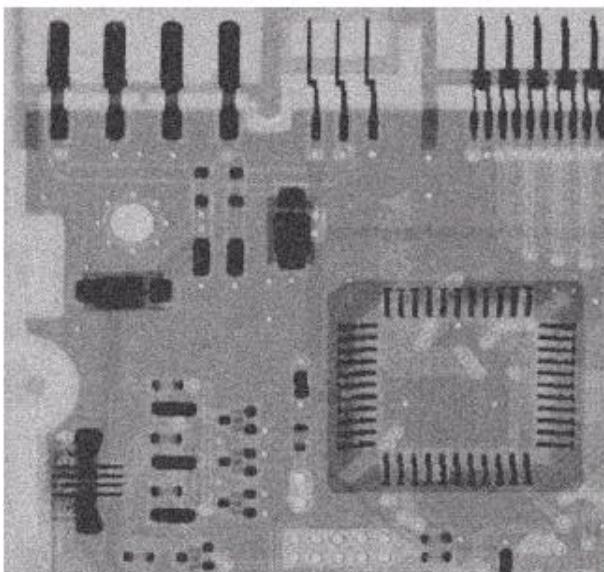
$$\sigma_\eta^2 < \sigma_L^2 \quad \text{细节或是边缘}$$

$$\sigma_\eta^2 > \sigma_L^2 \rightarrow \frac{\sigma_\eta^2}{\sigma_L^2} = 1 \quad \hat{f}(x, y) = m_L$$

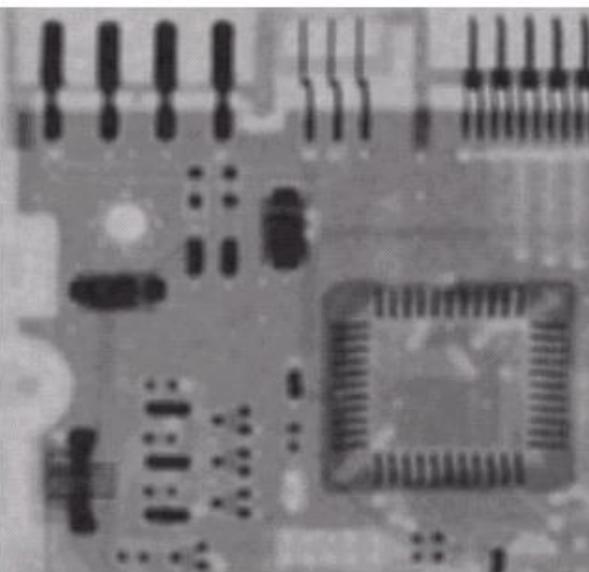


5.3 只存在噪声的复原 —— 空间滤波

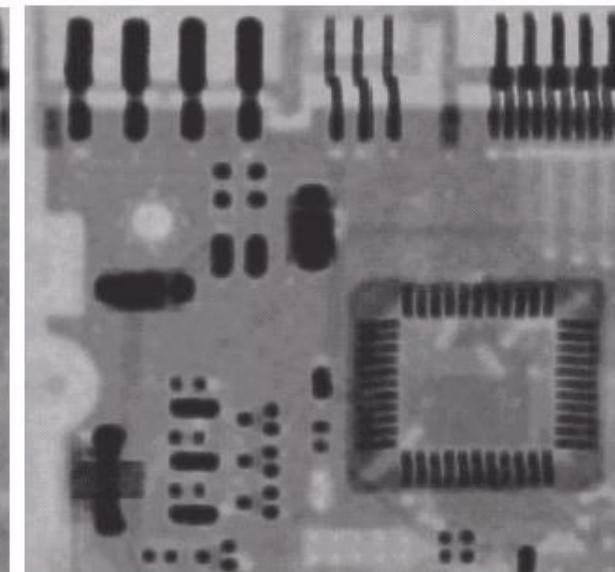
高斯噪声图像



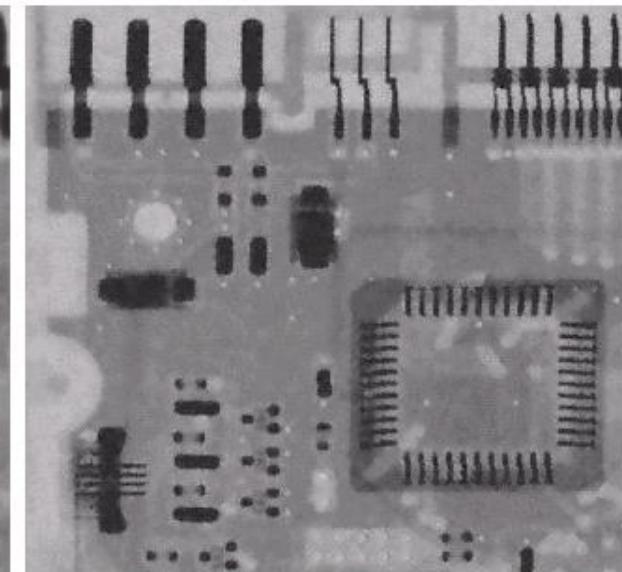
算术均值



几何均值



自适应均值



7×7

7×7

7×7

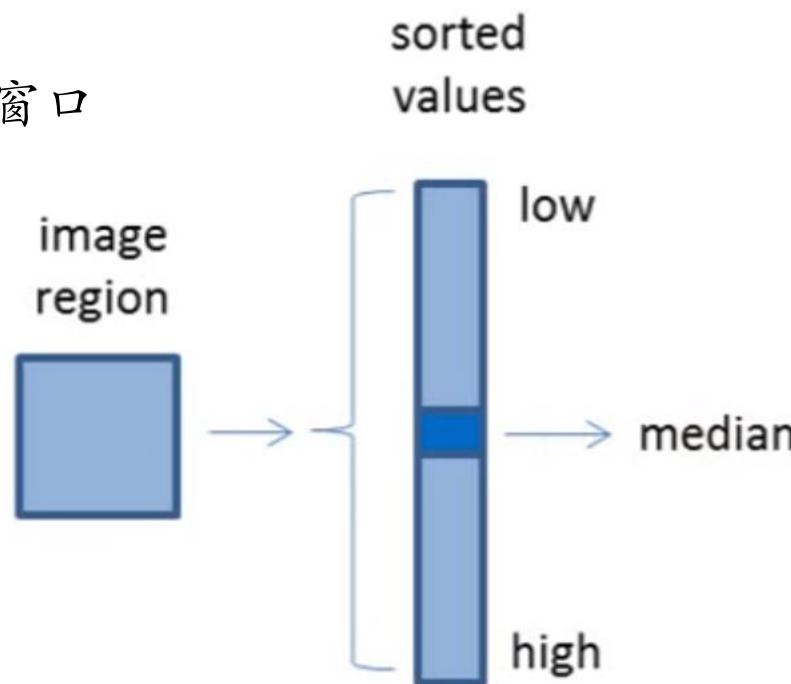
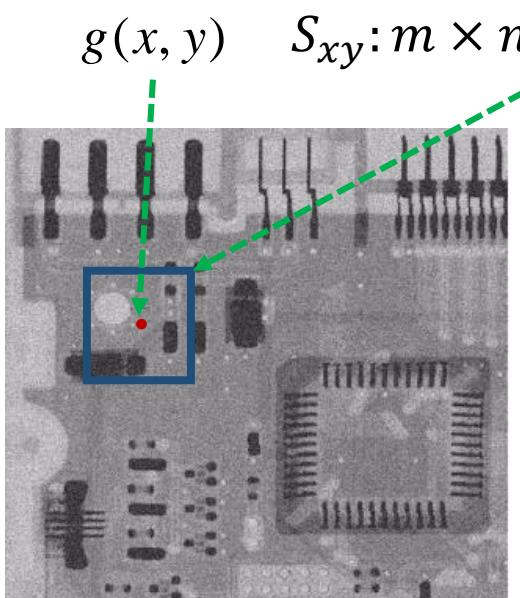
5.3 只存在噪声的复原 —— 空间滤波

- 统计排序滤波器

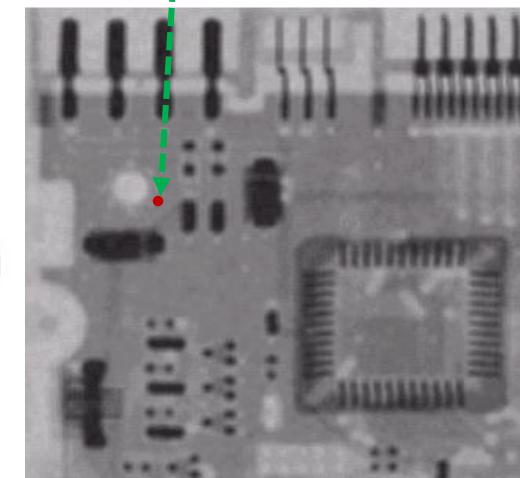
- 中值滤波 $\hat{f}(x, y) = \underset{(s,t) \in S_{xy}}{\text{median}} \{ g(s, t) \}$

对脉冲噪声尤其有效！

比相同尺寸的均值滤波器模糊要轻



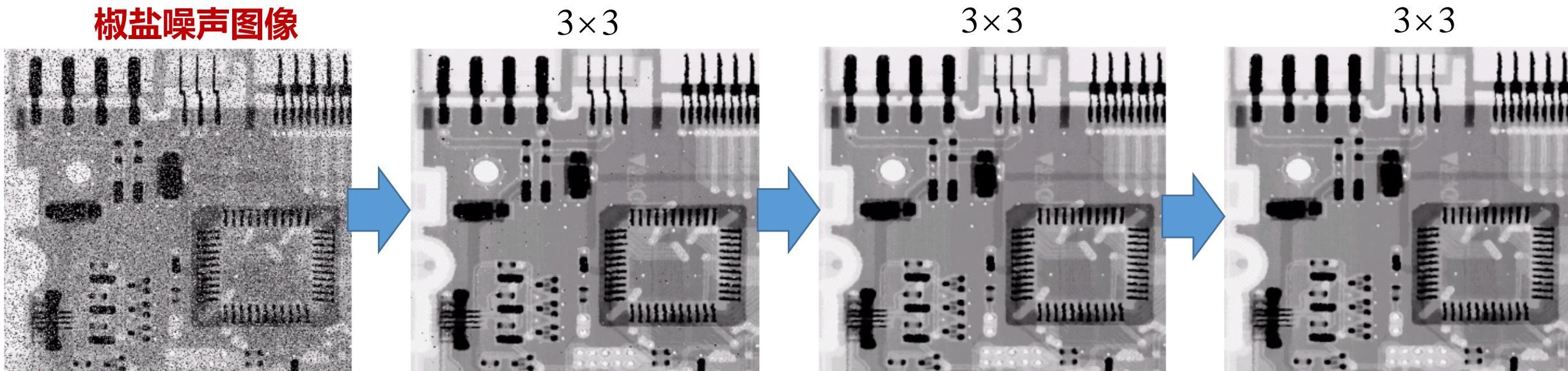
$$\hat{f}(x, y) = \underset{(s,t) \in S_{xy}}{\text{median}} \{ g(s, t) \}$$



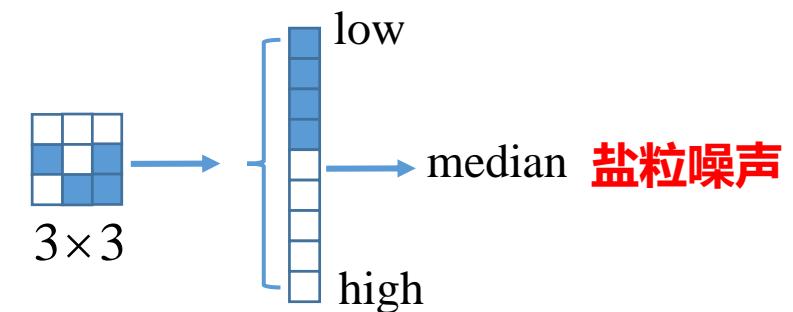
The filter size is 7×7

5.3 只存在噪声的复原 —— 空间滤波

• 中值滤波



$$P_a = P_b = 0.1$$



5.3 只存在噪声的复原 —— 空间滤波

- 统计排序滤波器

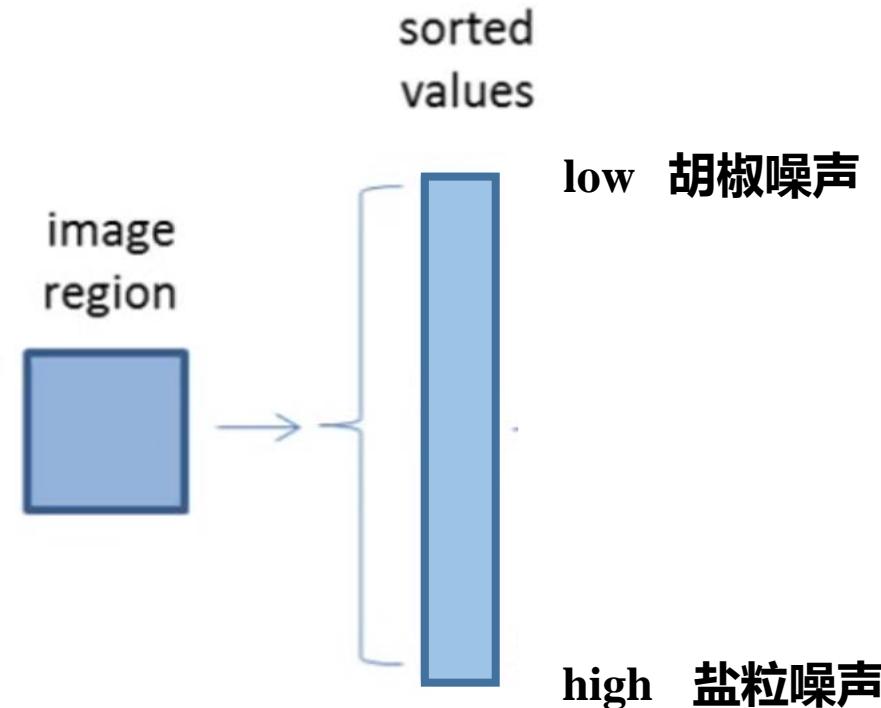
- 最大值和最小值滤波器

$$\hat{f}(x, y) = \max_{(s, t) \in S_{xy}} \{g(s, t)\}$$

■ 适用于胡椒噪声

$$\hat{f}(x, y) = \min_{(s, t) \in S_{xy}} \{g(s, t)\}$$

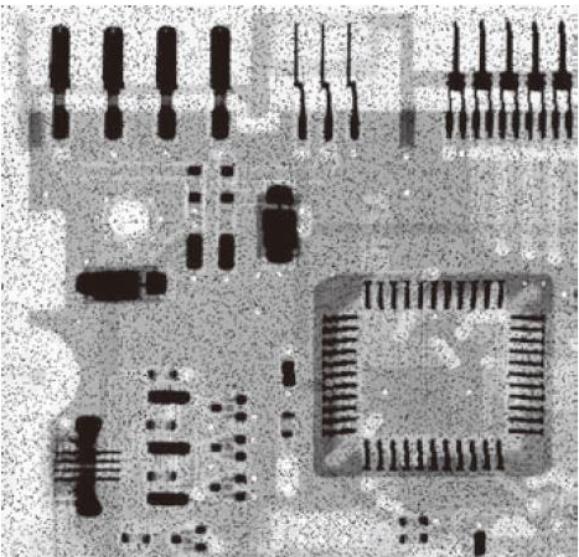
■ 适用于盐粒噪声



5.3 只存在噪声的复原 —— 空间滤波

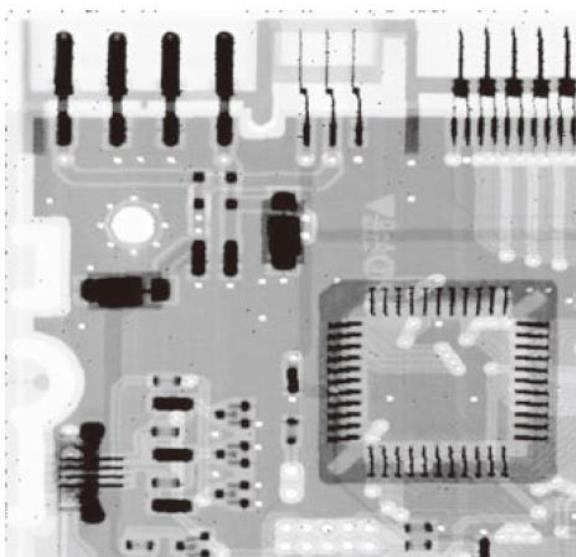
• 最大值和最小值滤波

胡椒噪声图像

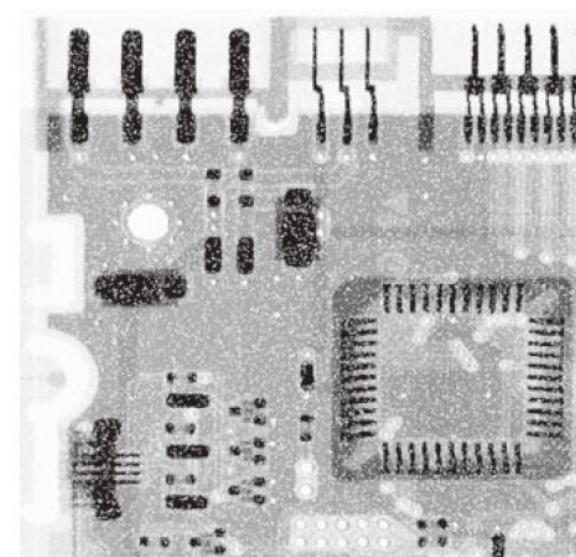


$$P_a = 0.1$$

最大滤波器

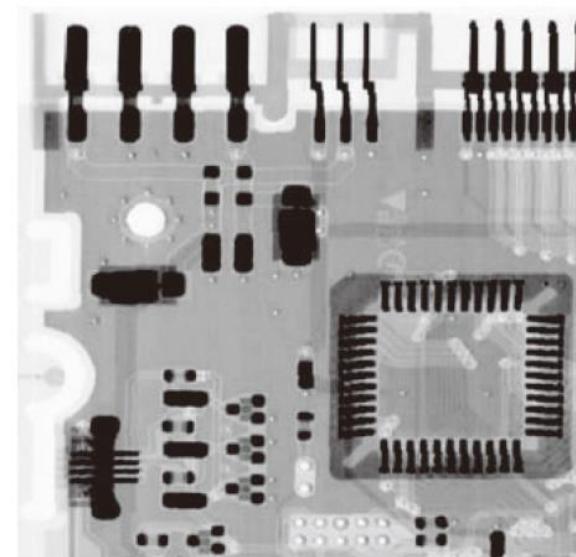


盐粒噪声图像



$$P_b = 0.1$$

最小滤波器



5.3 只存在噪声的复原 —— 空间滤波

- 统计排序滤波器

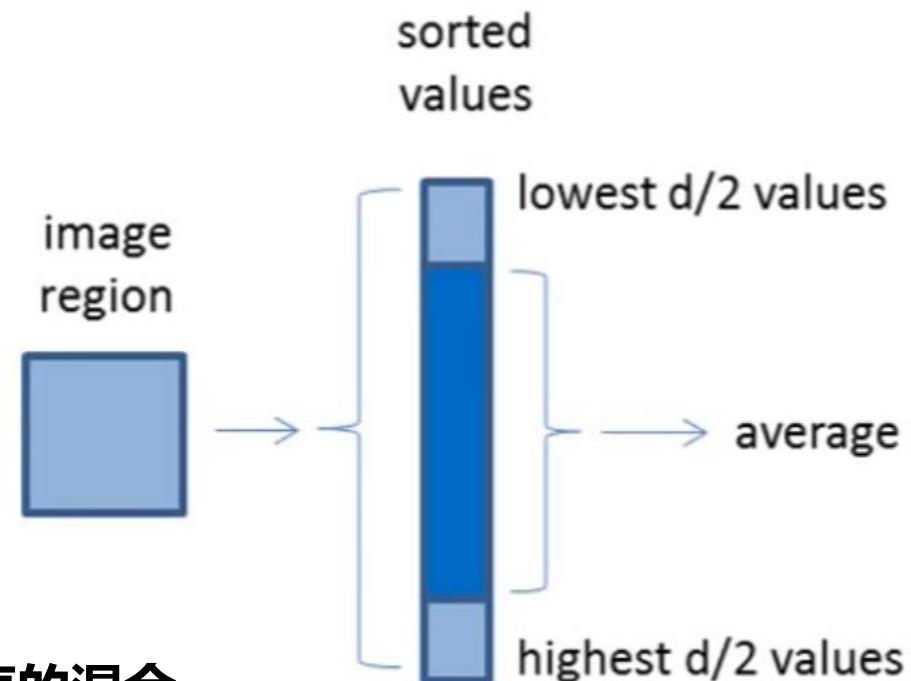
- Alpha-trimmed 均值滤波

$$\hat{f}(x, y) = \frac{1}{mn - d} \sum_{(s,t) \in S_{xy}} g_r(s, t)$$

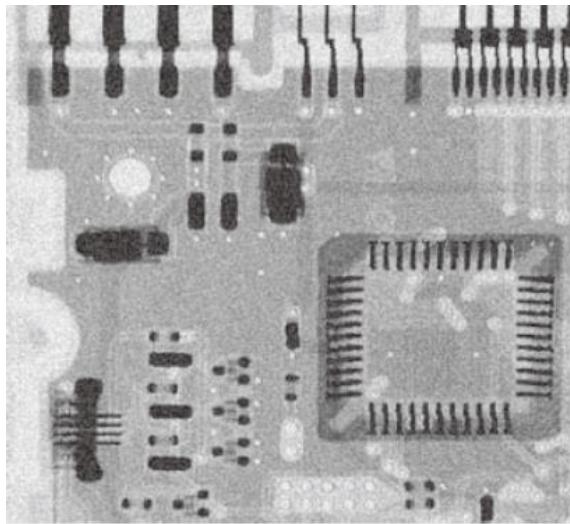
其中 $d \in [0, mn - 1]$ 高斯噪声和椒盐噪声的混合

当 $d=0$ 时 \rightarrow 算数均值滤波器 高斯噪声

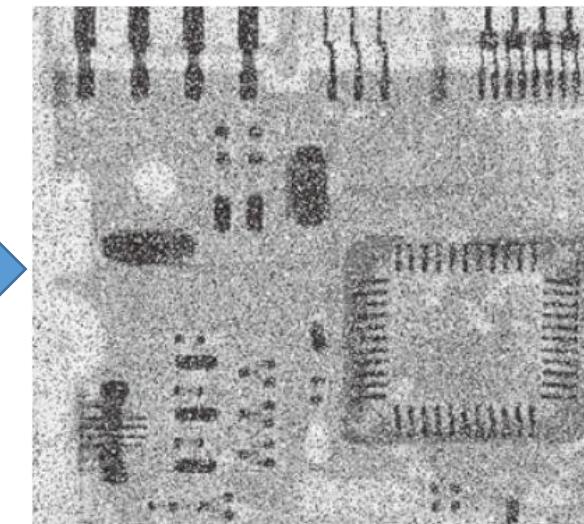
当 $d=mn - 1$ 时 \rightarrow 中值滤波器 椒盐噪声



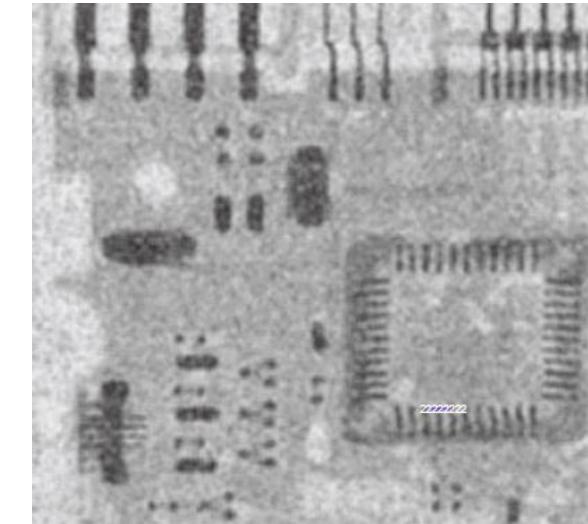
引入均匀噪声的图像



均匀噪声 + 椒盐噪声

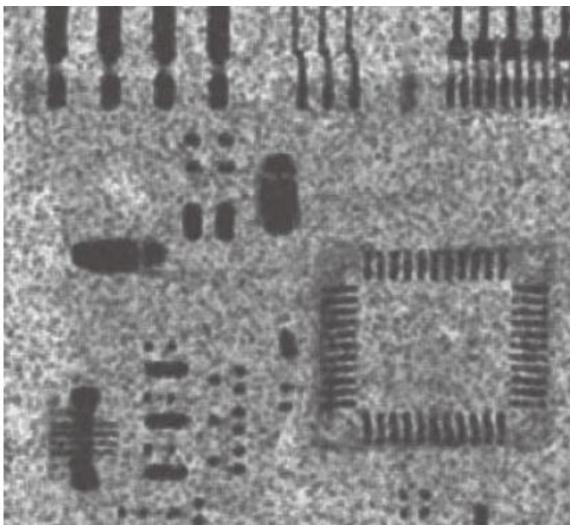


算数均值

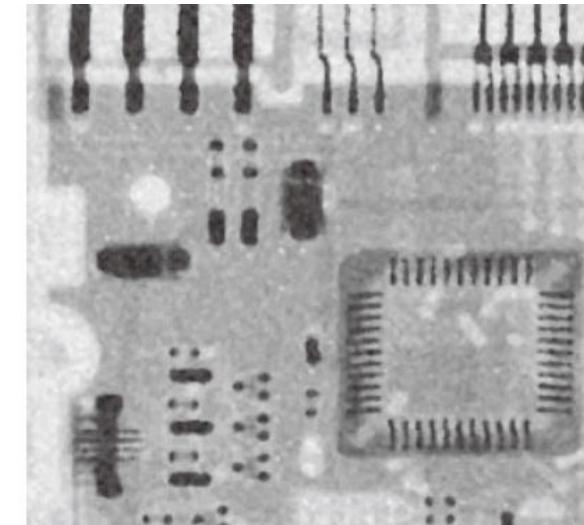


5×5 filter

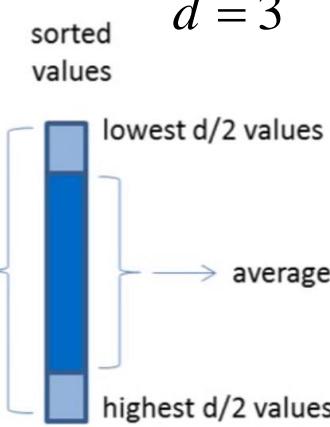
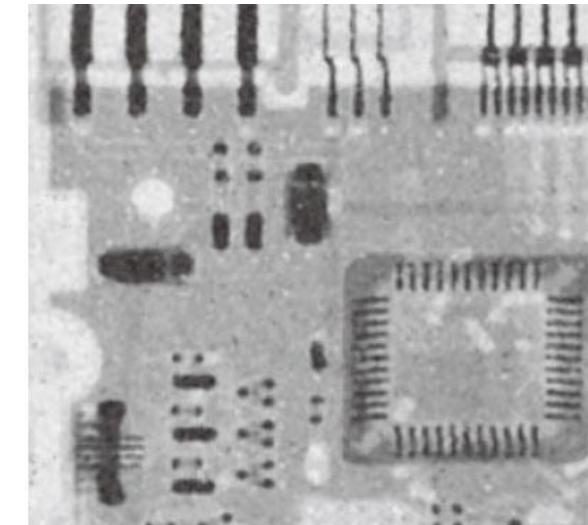
几何均值



中值滤波

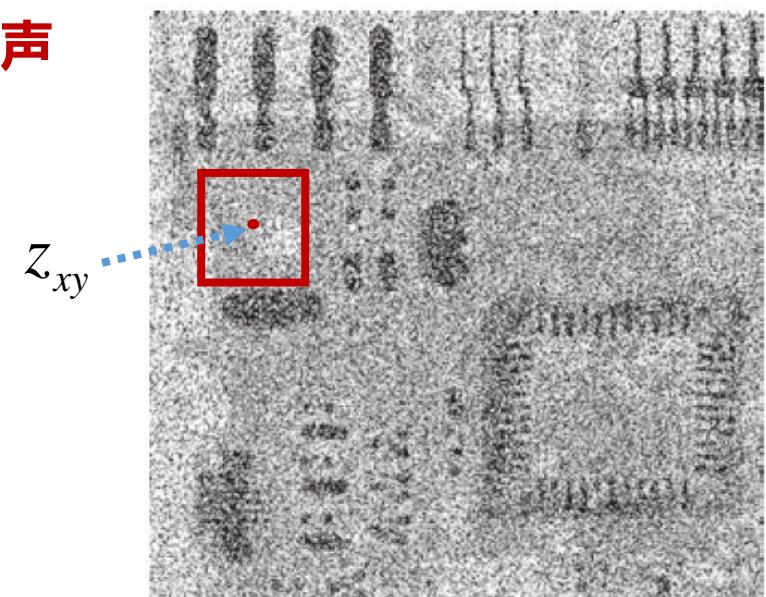
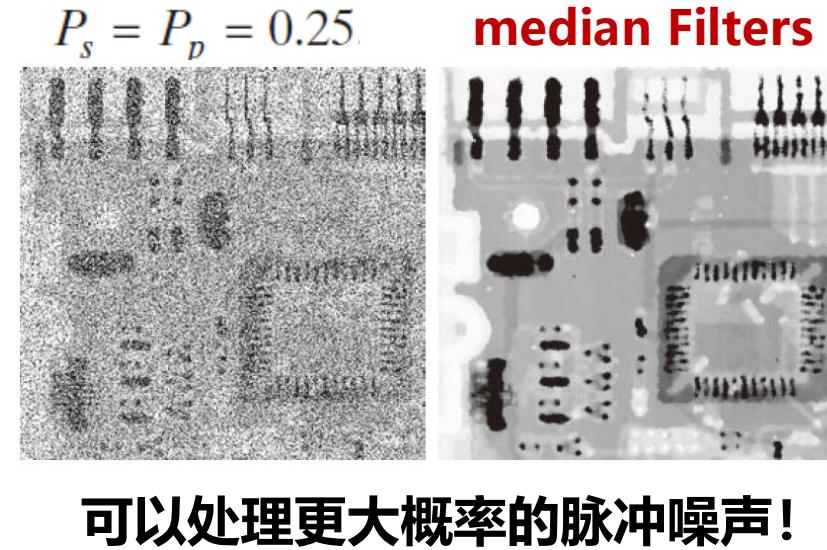
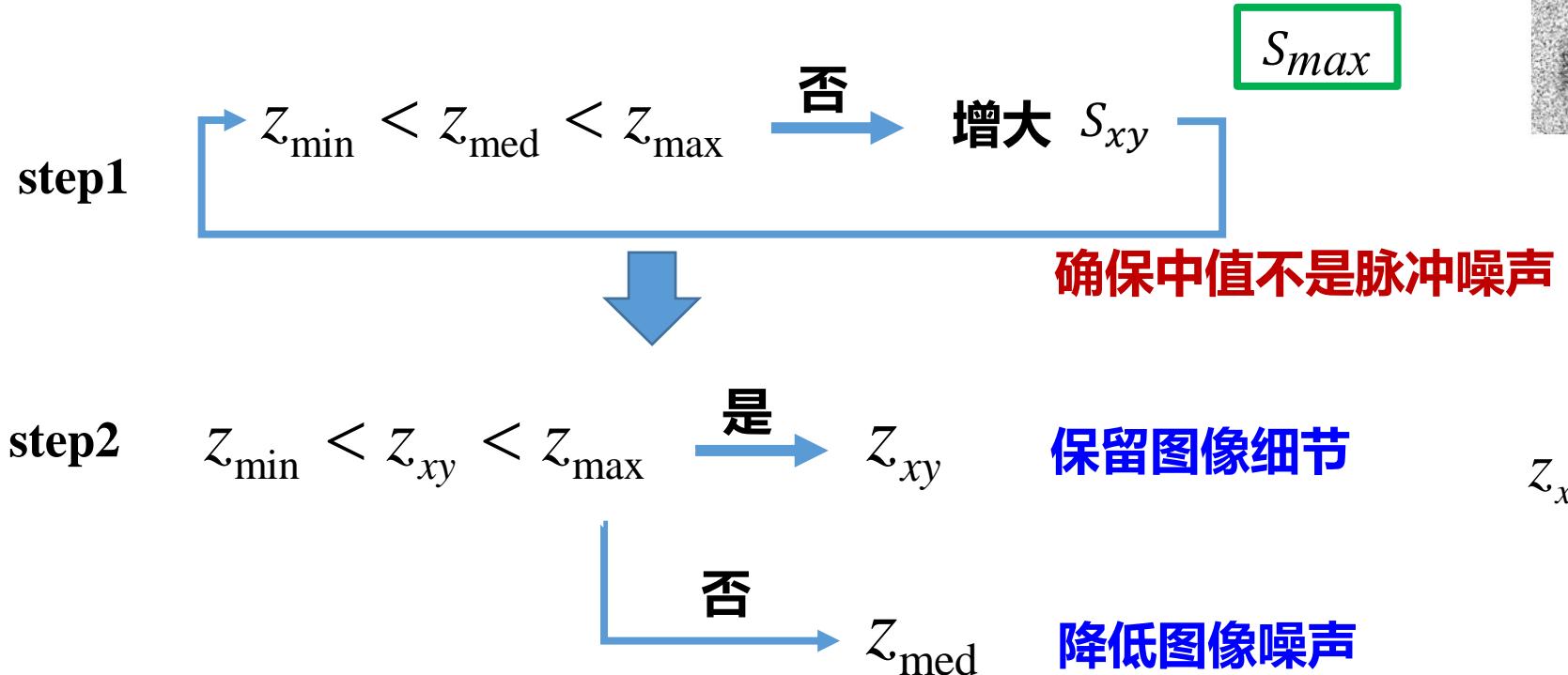


Alpha-trimmed 均值滤波



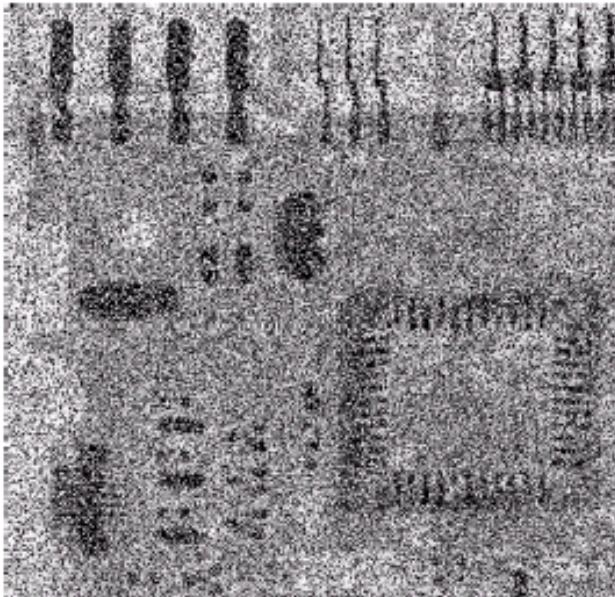
5.3 只存在噪声的复原 —— 空间滤波

• 自适应中值滤波 Adaptive median Filters

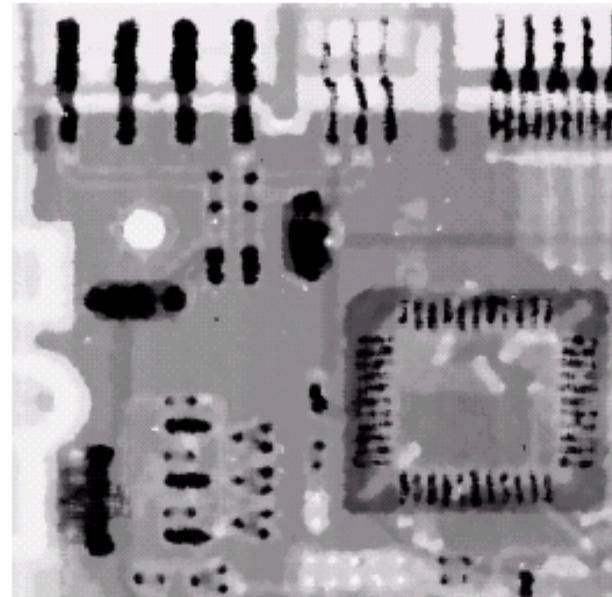


5.3 只存在噪声的复原 —— 空间滤波

椒盐噪声 $P_s = P_p = 0.25$

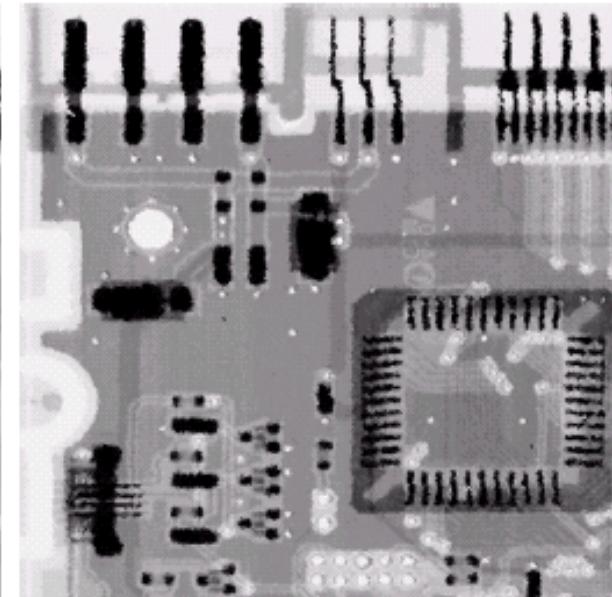


中值滤波



7×7

自适应中值滤波

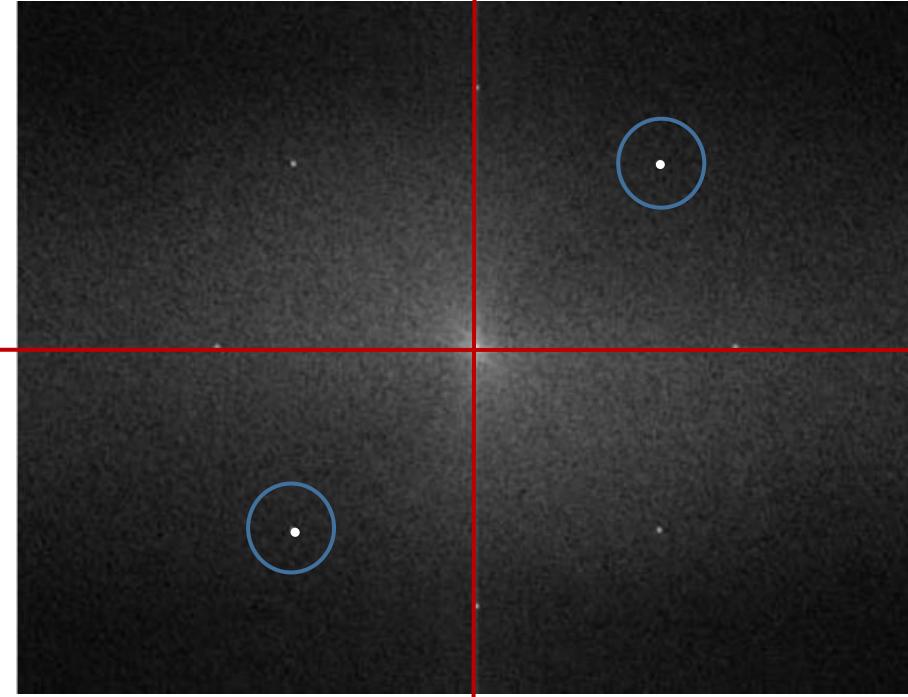
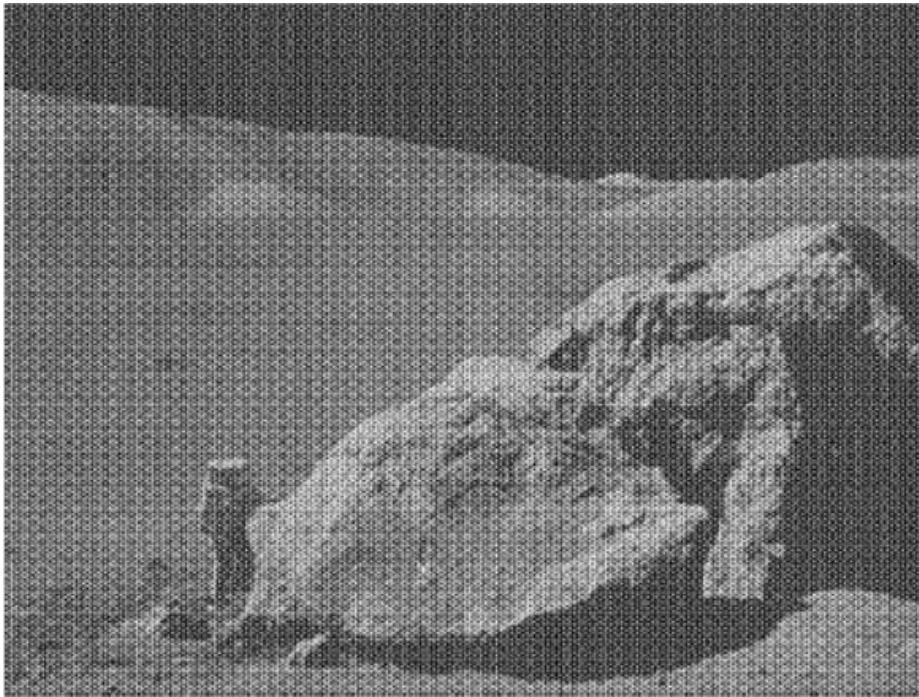


$S_{max} = 7 \times 7$

主要内容

- 图像退化/复原过程的模型
- 噪声模型
- 空间域滤波方法
- 频率域滤波方法
- 退化函数的估计
- 逆滤波
- 维纳滤波

5.4 用频率域滤波消除周期噪声

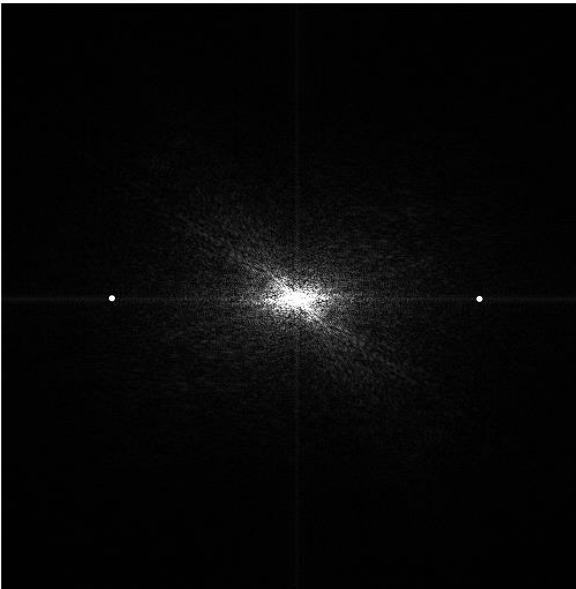
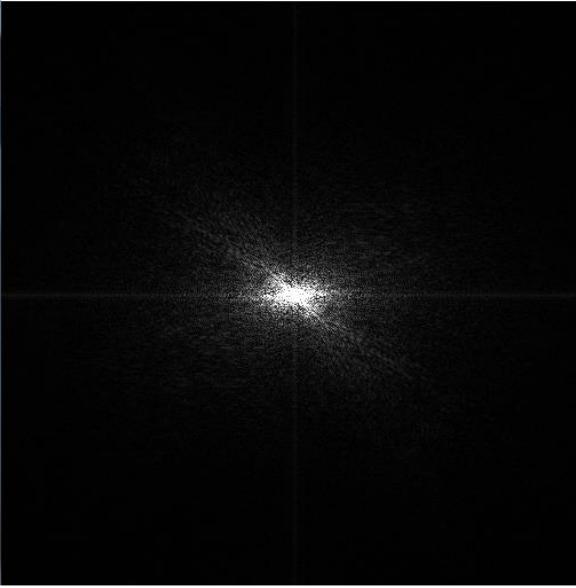


每对共轭脉冲对应于一个正弦波

$$f(x, y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi(\frac{ux}{M} + \frac{vy}{N})}$$

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(\frac{ux}{M} + \frac{vy}{N})}$$

5.4 用频率域滤波消除周期噪声

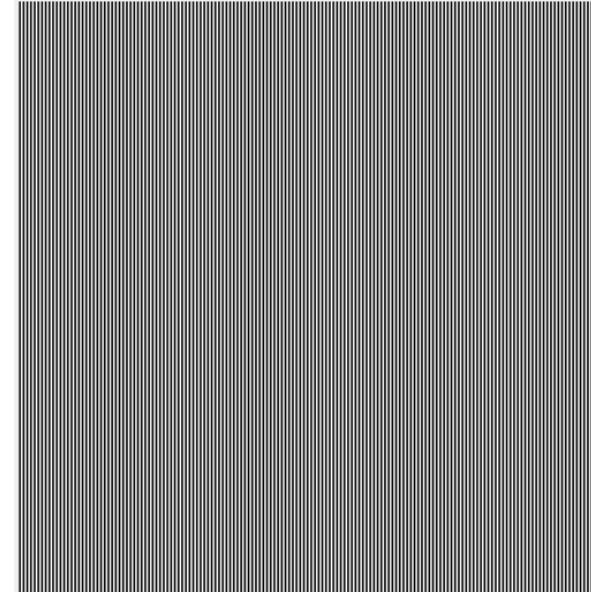


$$F(u_0, v_0) = a + bi$$

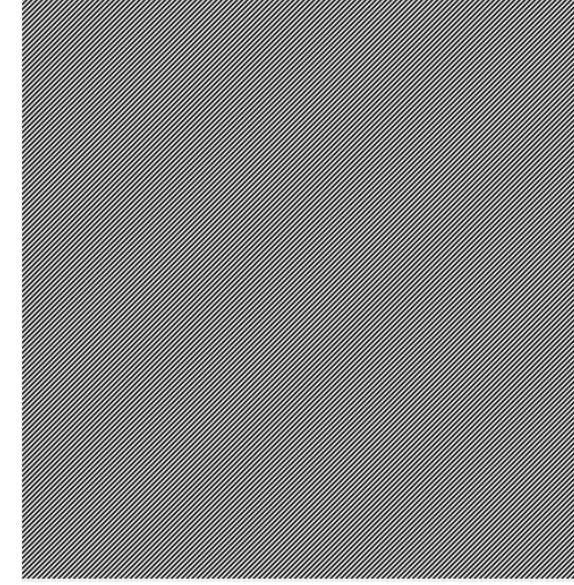
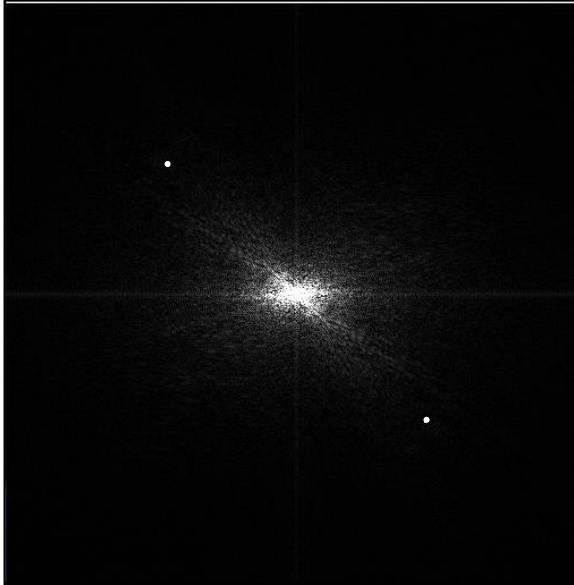
$$F(-u_0, -v_0) = a - bi$$



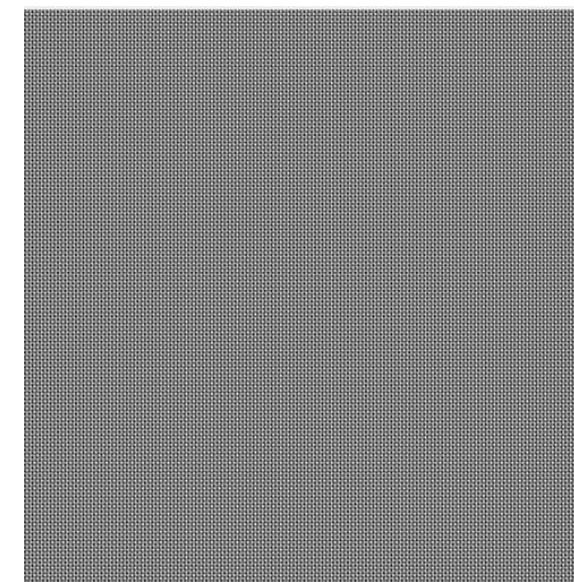
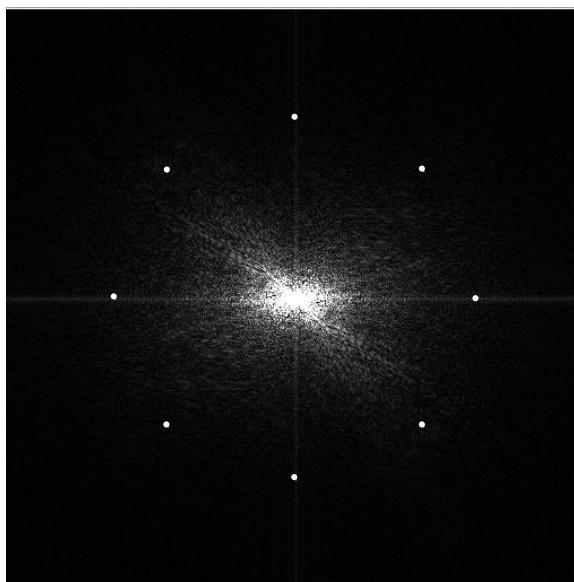
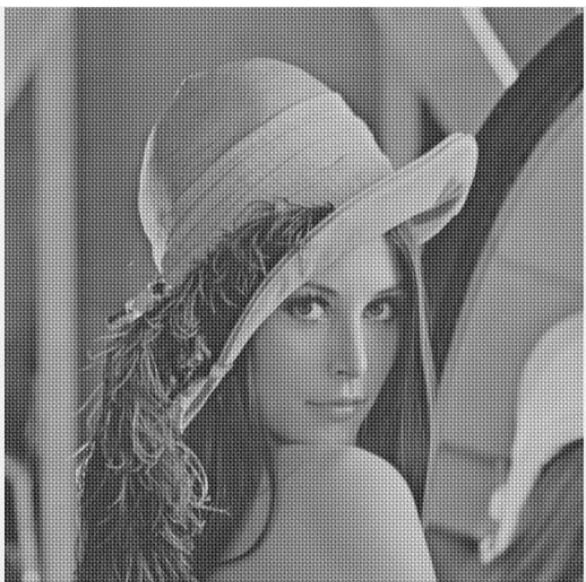
$$f(x, y) = 2a \cos\left(\frac{u_0 x}{M} + \frac{v_0 y}{N}\right) - 2b \sin\left(\frac{u_0 x}{M} + \frac{v_0 y}{N}\right)$$



5.4 用频率域滤波消除周期噪声

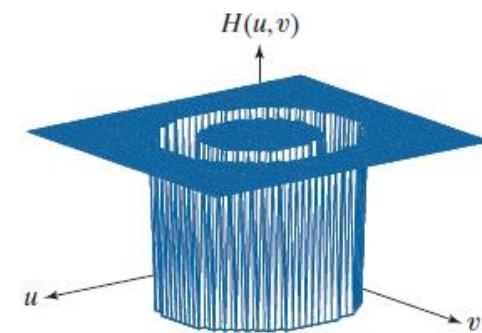
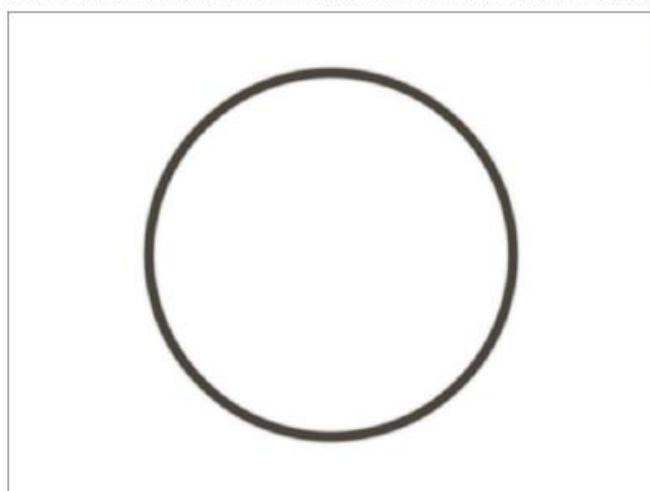
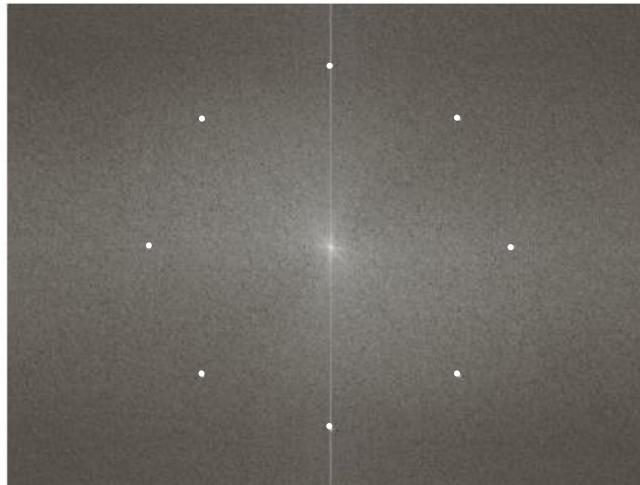
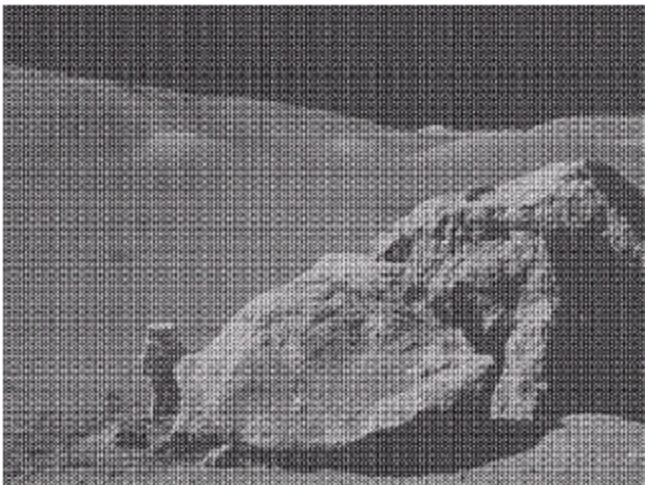


用频率域技术可以有效
分析并滤除周期噪声

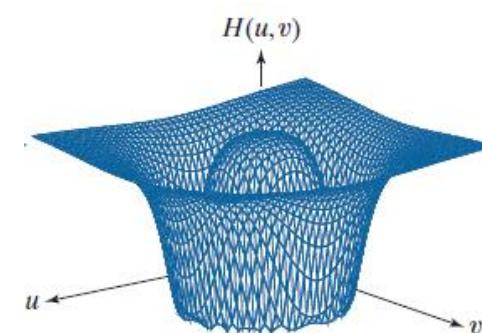


选择性滤波器

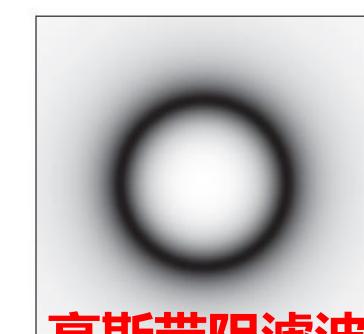
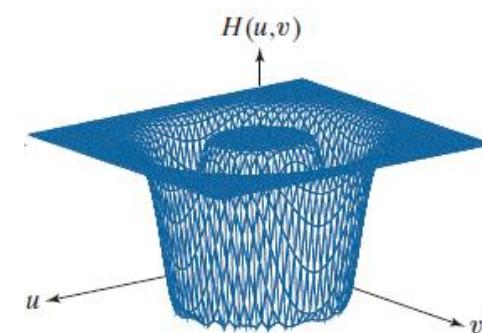
5.4 用频率域滤波消除周期噪声——带阻滤波器



理想滤波器



布特沃斯滤波器



高斯带阻滤波器

5.4 用频率域滤波消除周期噪声——限波滤波器

