

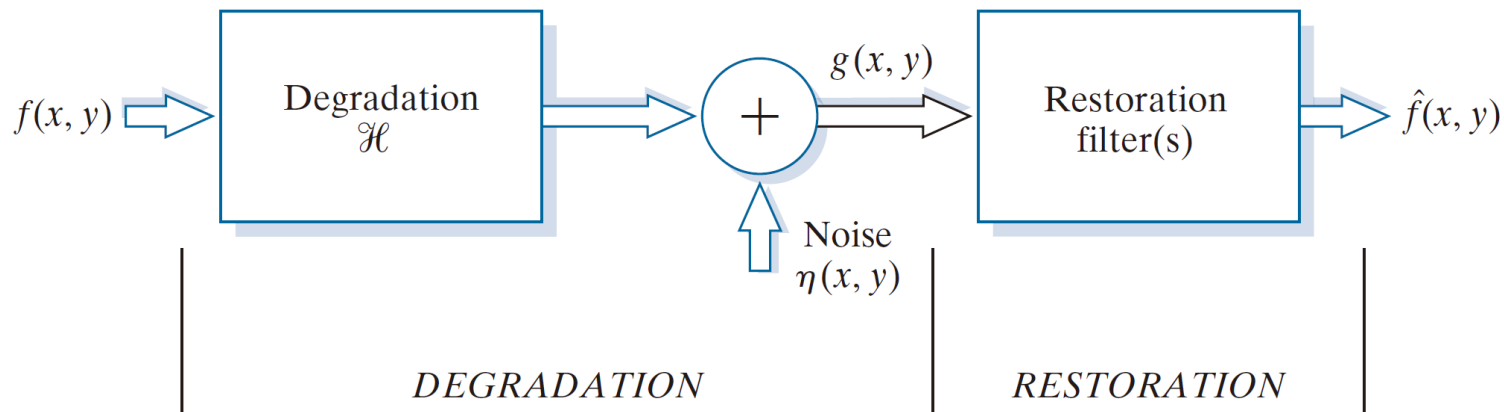
主要内容

- 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.



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

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



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

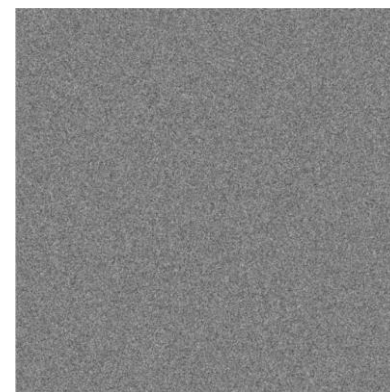
$$G(u, v) = F(u, v) + N(u, v)$$

加性噪声



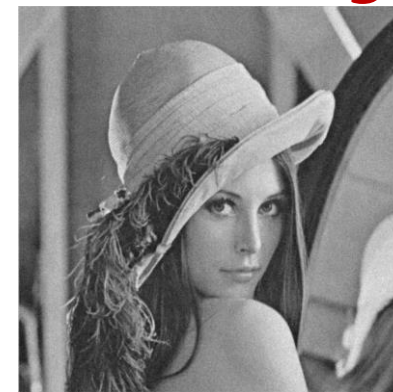
$f(x, y)$

+



$\eta(x, y)$

=



$g(x, y)$

• 去噪 denoising

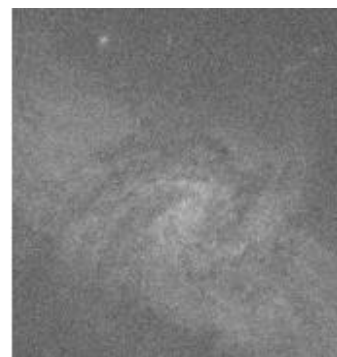
图像降噪 —— 噪声图像相加（平均）

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

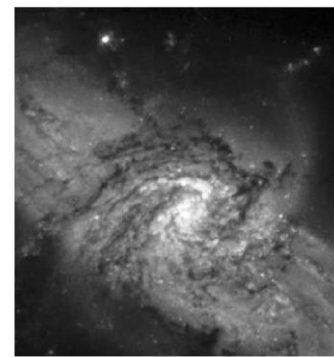
退化的图像

理想图像

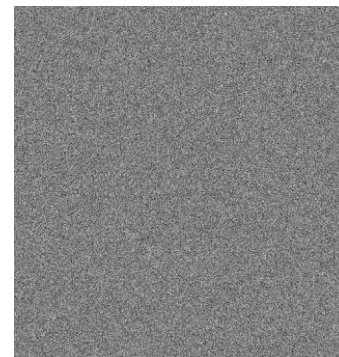
噪声



=



+

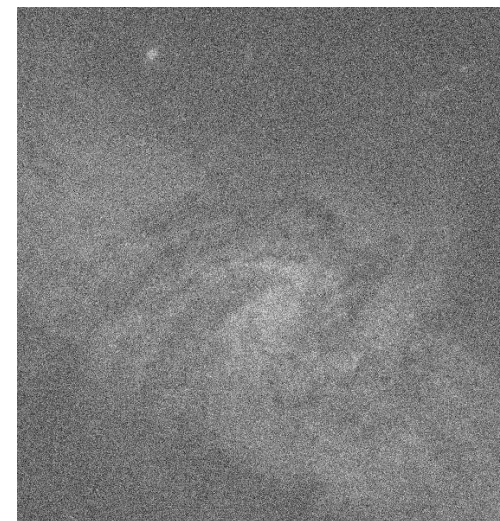


假设噪声服从零均值的高斯分布 $\eta(x, y) \sim N(0, \sigma_{\eta(x, y)}^2)$

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

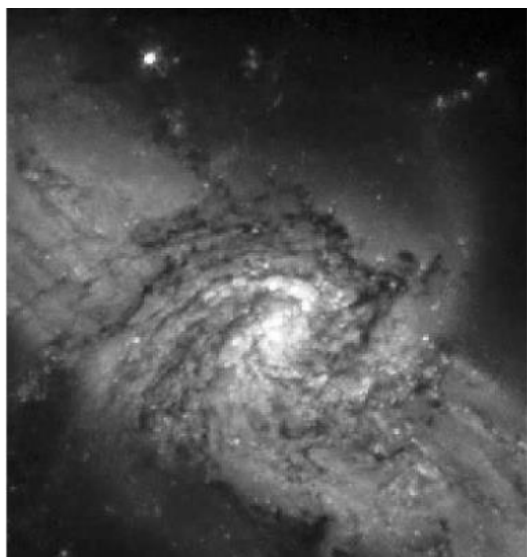
输出图像为 $\bar{g}(x, y) = \frac{1}{K} \sum_{i=1}^K g_i(x, y)$

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

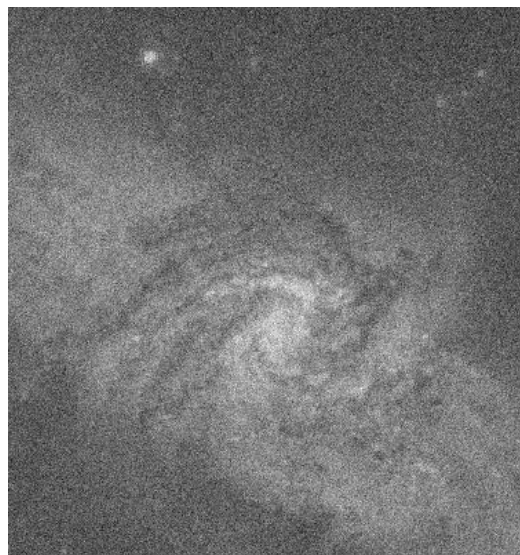
理想图像 (8 bit)



噪声图像 (8 bit)



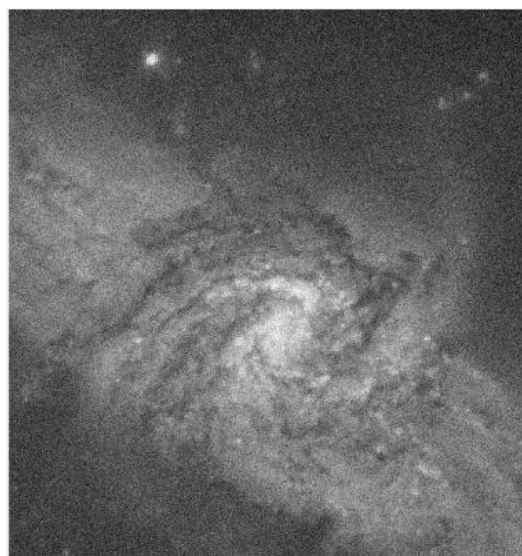
NEX=5



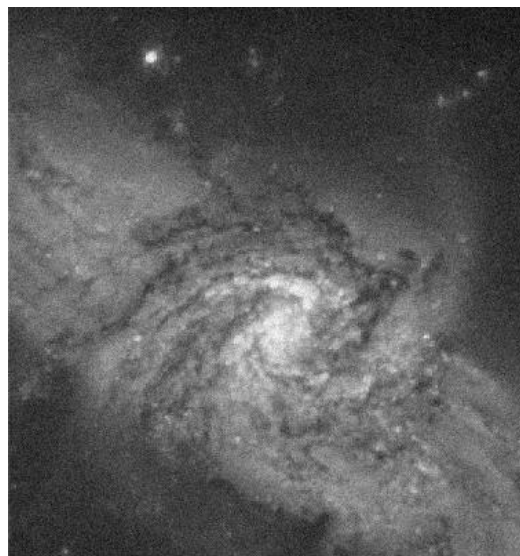
NEX=10



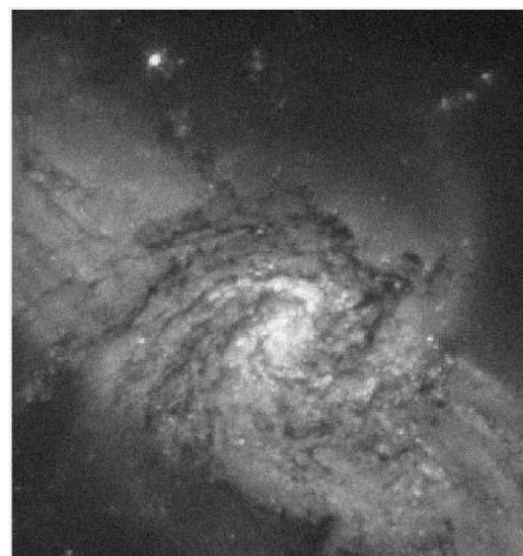
NEX=20

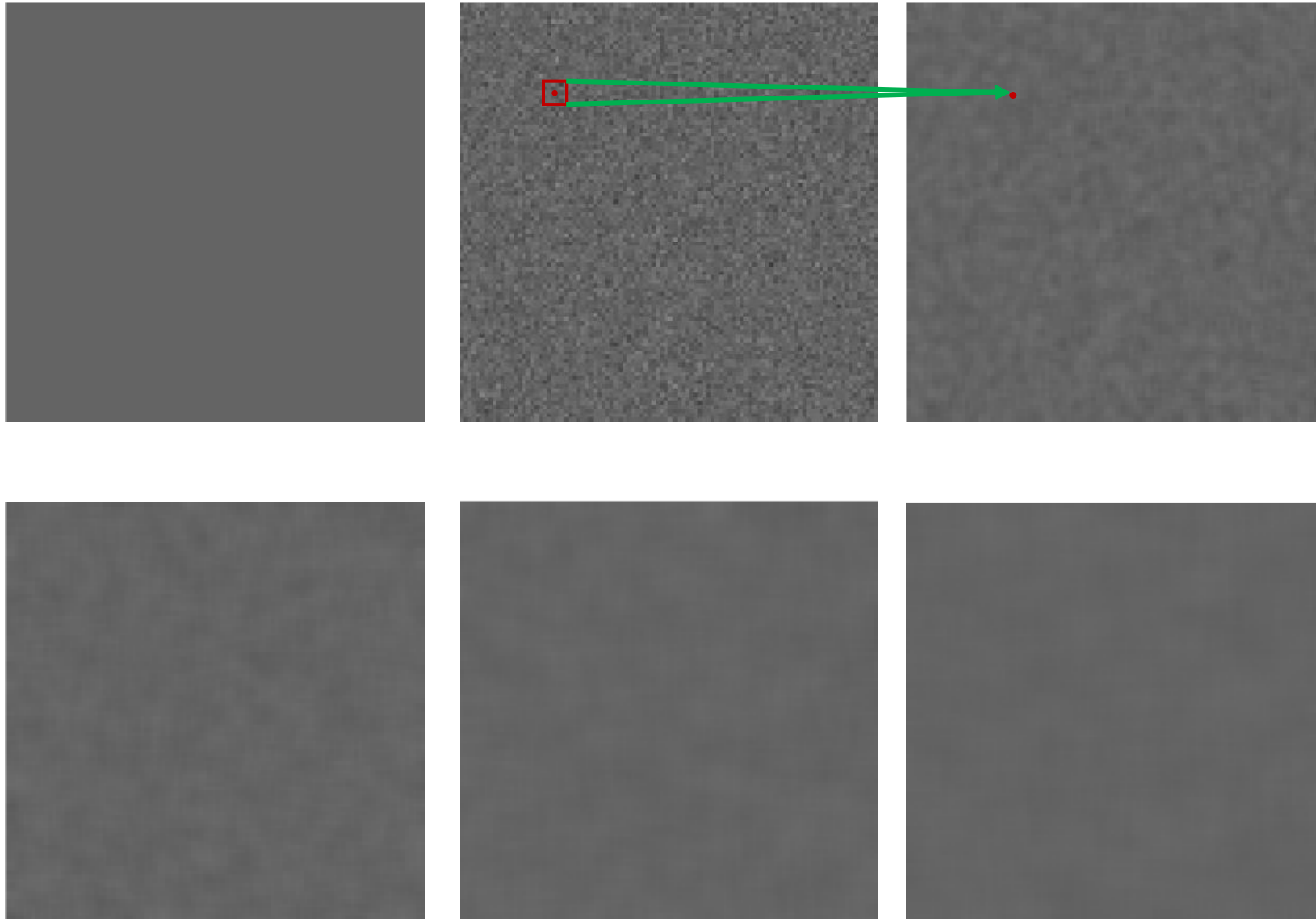


NEX=50

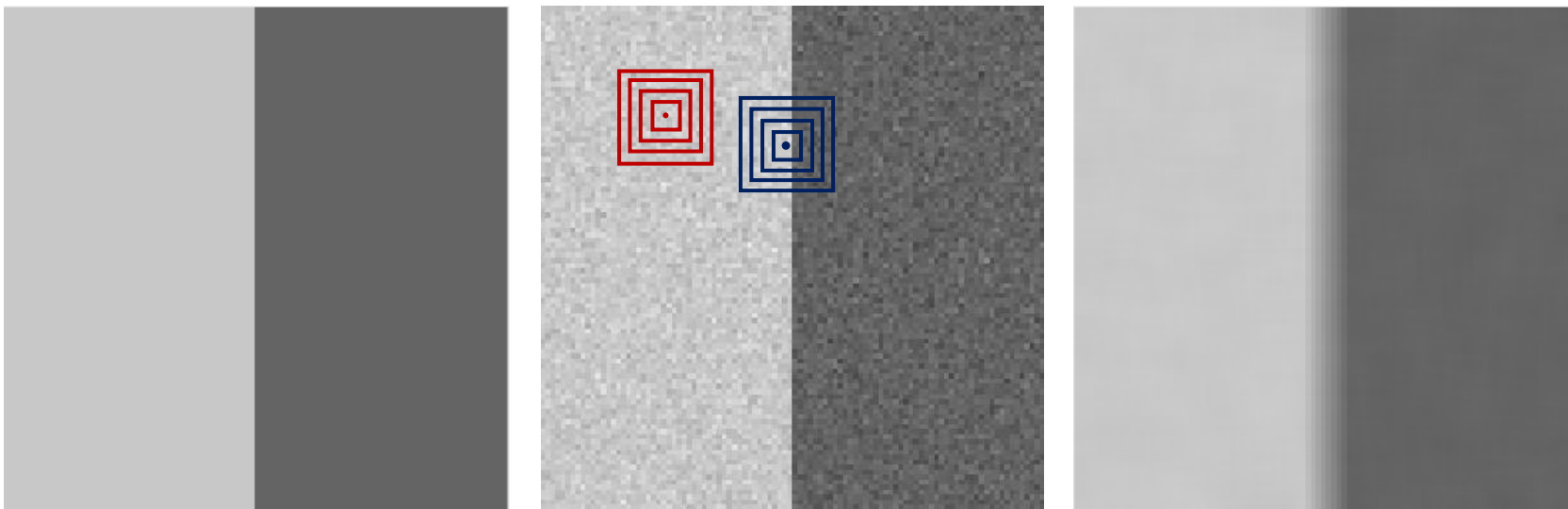


NEX=100





$I=100, \sigma=10$

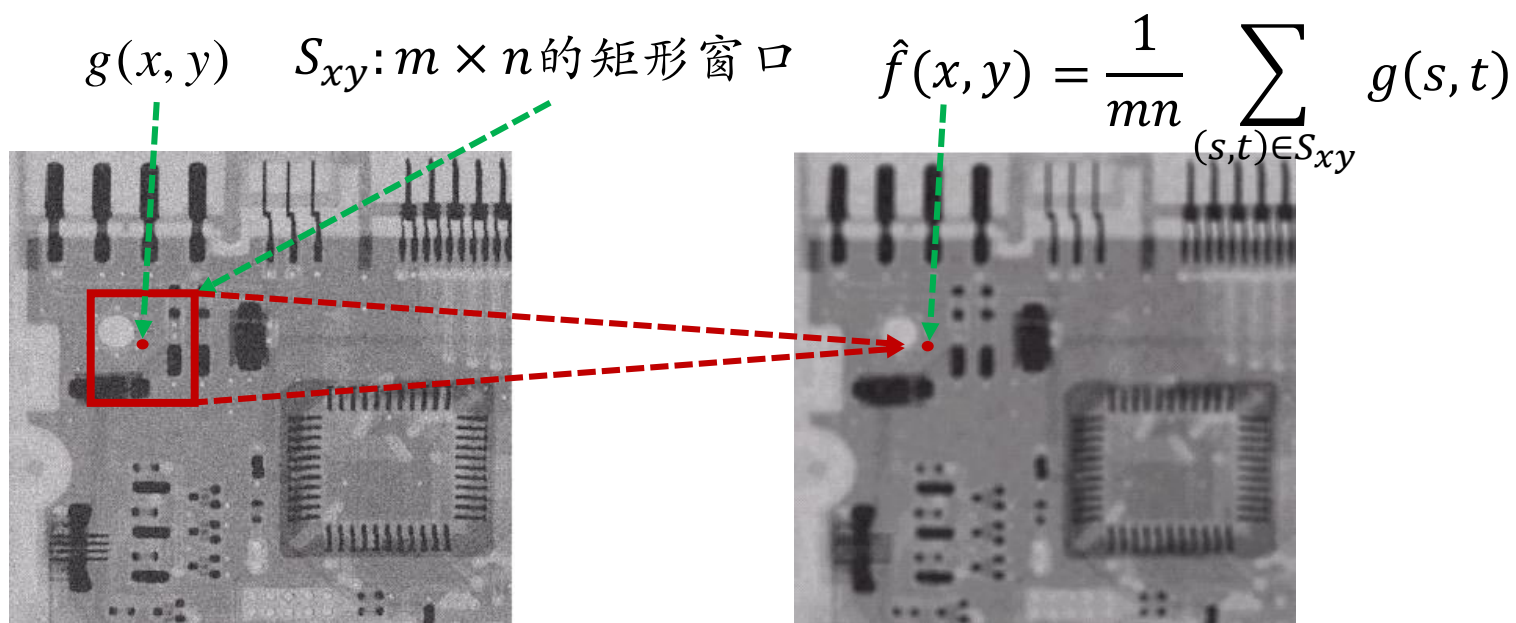


$\sigma = 10$

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

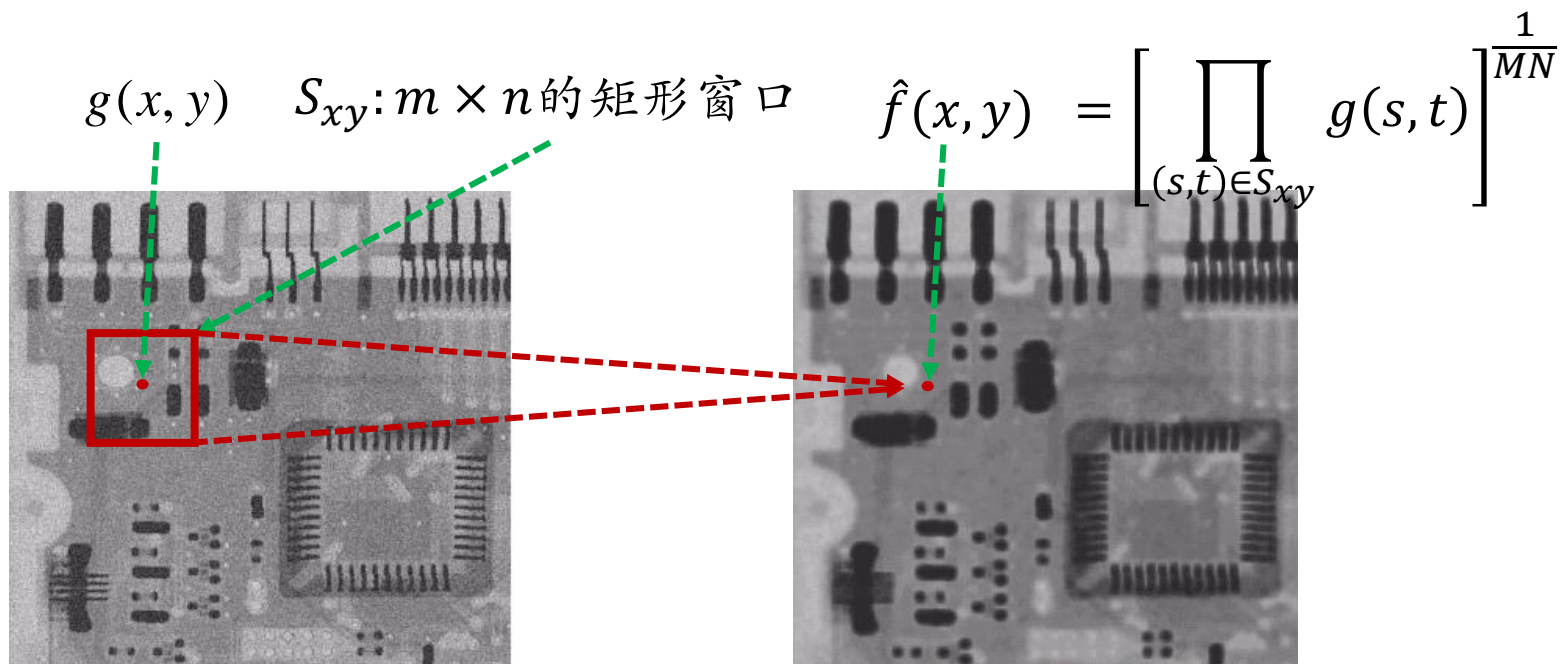
图像细节模糊

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

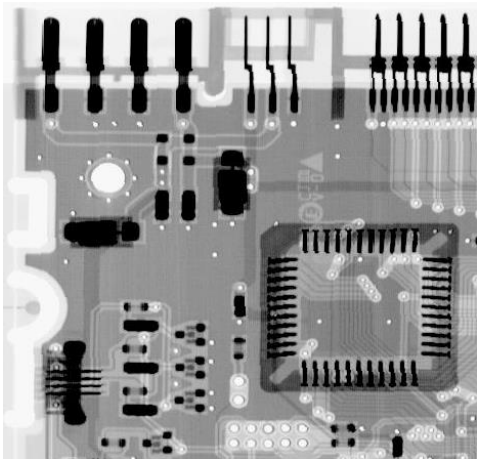
- 几何均值 $\hat{f}(x, y) = \left[\prod_{(s,t) \in S_{xy}} g(s, t) \right]^{\frac{1}{MN}}$



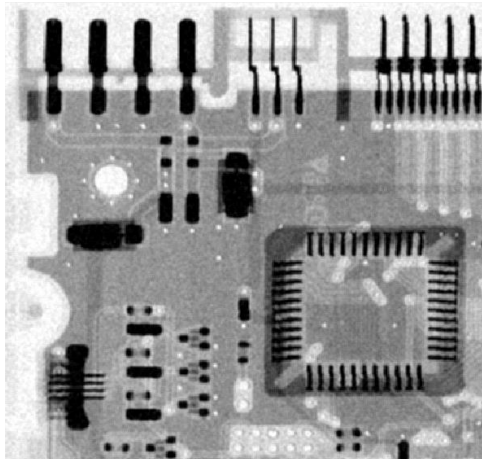
The filter size is 7×7

• 算数均值vs几何均值

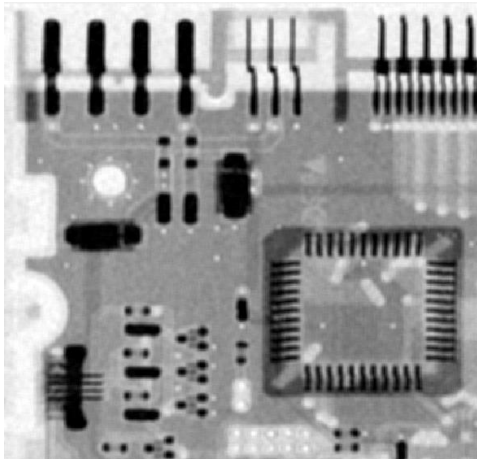
无噪图像



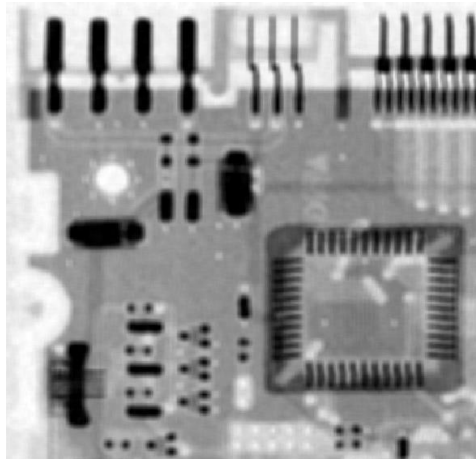
3×3



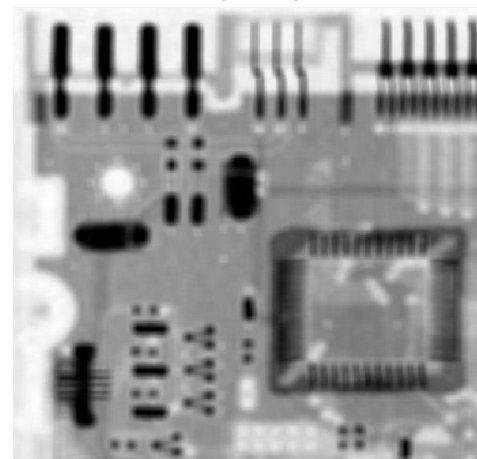
5×5



7×7

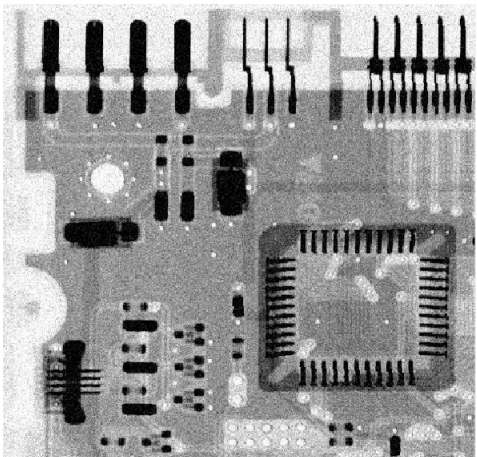


9×9

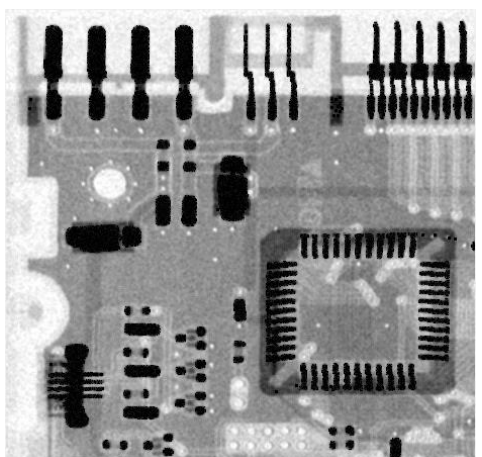


算数均值

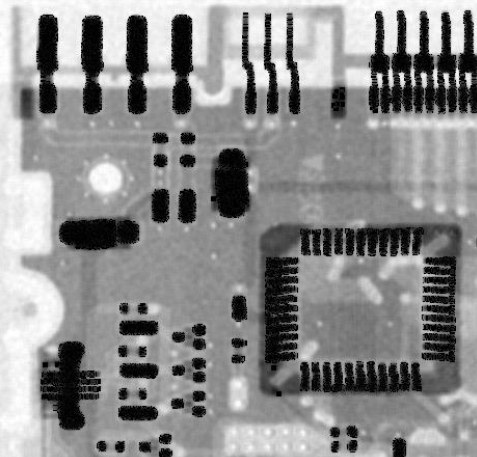
噪声图像



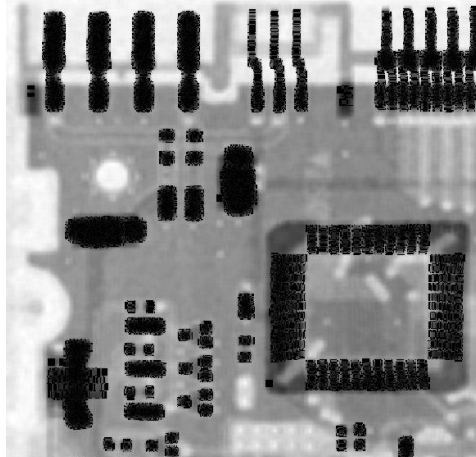
3×3



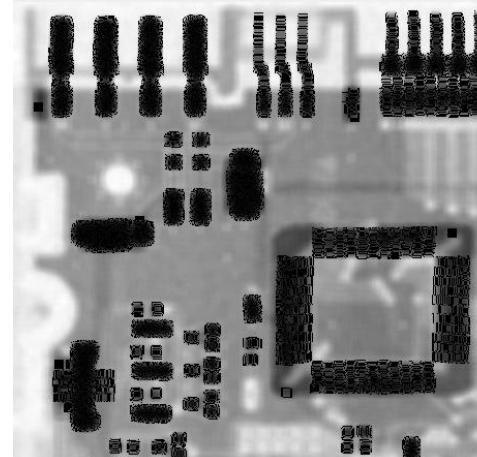
5×5



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
 S_{xy} 内所有像素值的方差

分析

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

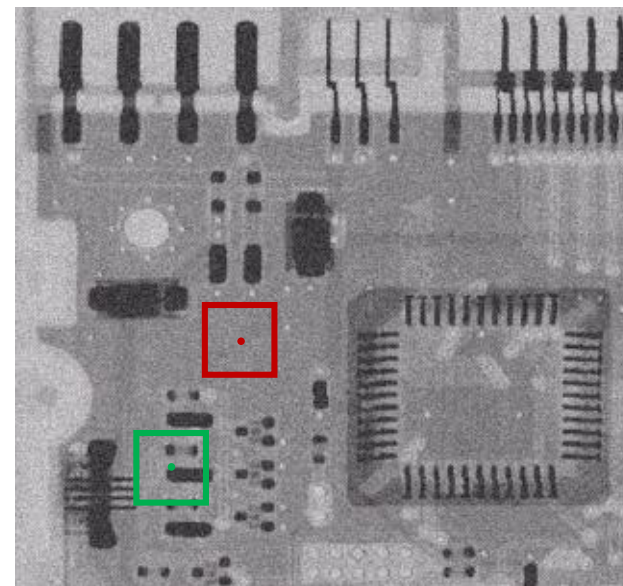
均匀区域 $\hat{f}(x, y) = m_L$

$$\sigma_{\eta}^2 < \sigma_L^2$$

细节或是边缘

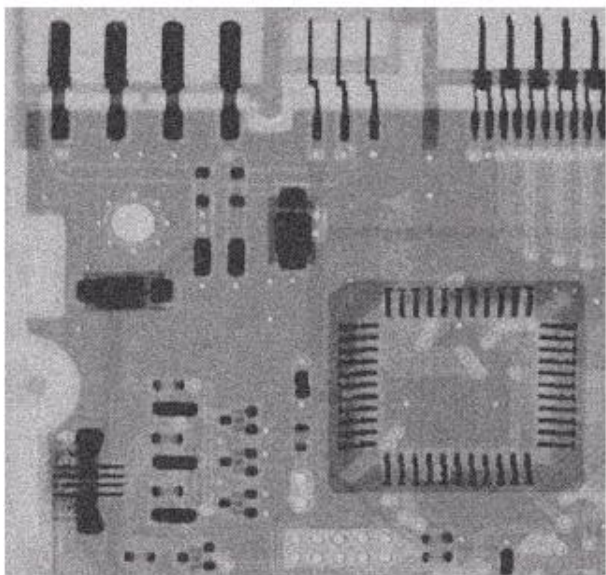
$$\sigma_{\eta}^2 > \sigma_L^2$$

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

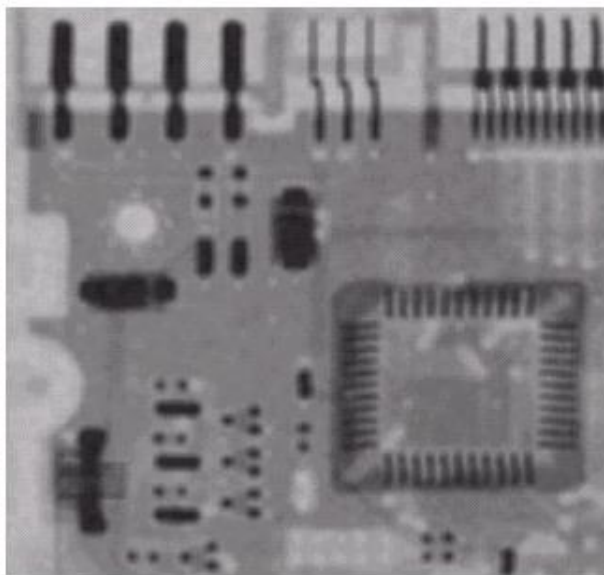


5.3 只存在噪声的复原 —— 空间滤波

高斯噪声图像

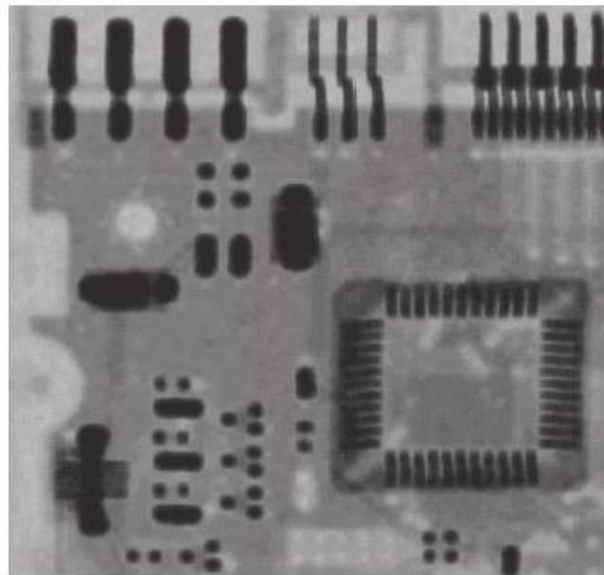


算术均值



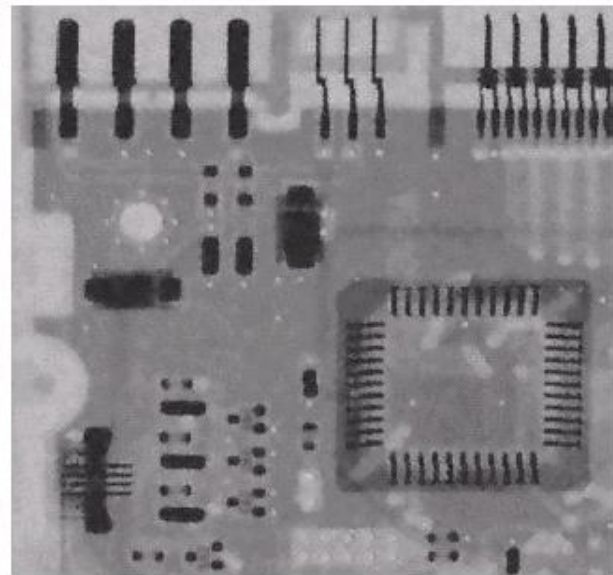
7×7

几何均值



7×7

自适应均值



7×7

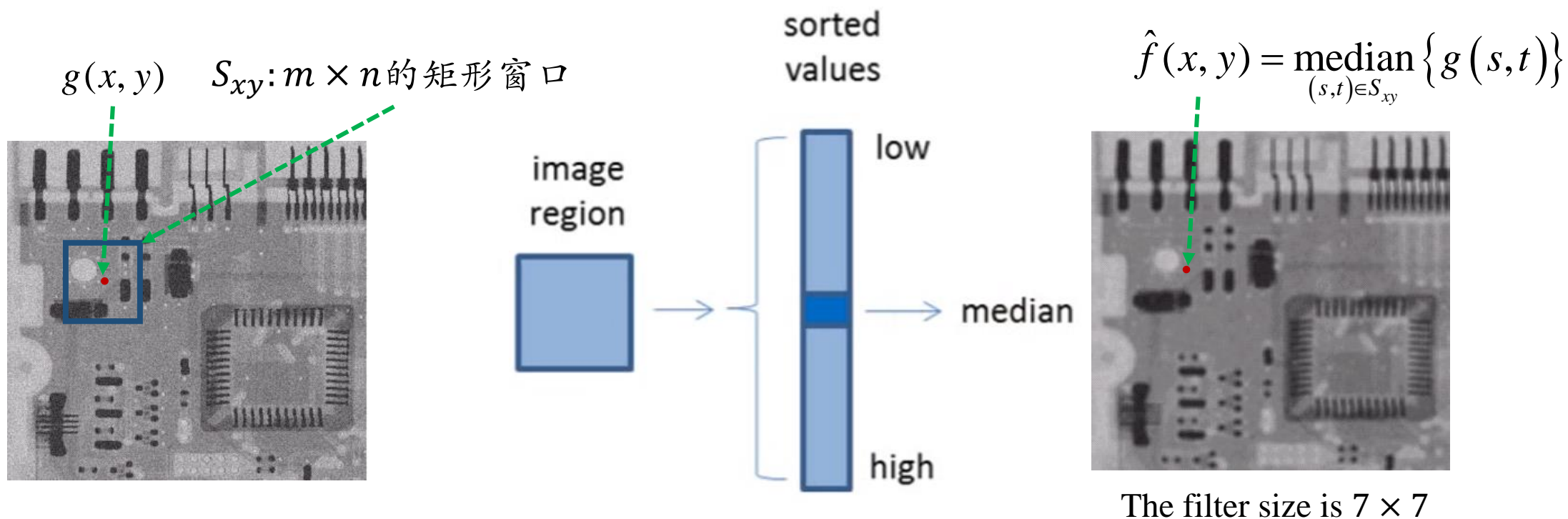
5.3 只存在噪声的复原 —— 空间滤波

• 统计排序滤波器

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

对脉冲噪声尤其有效!

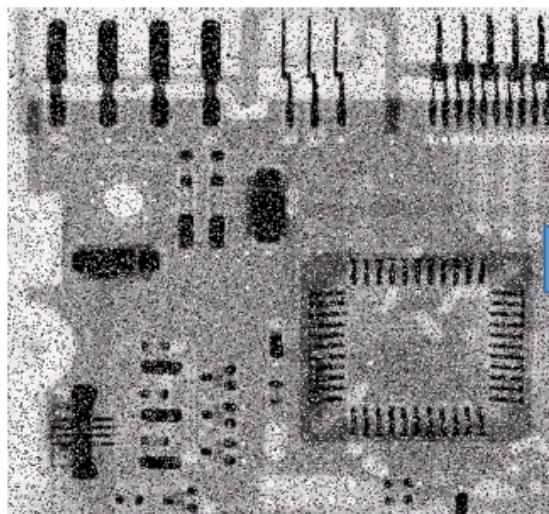
比相同尺寸的均值滤波器模糊要轻



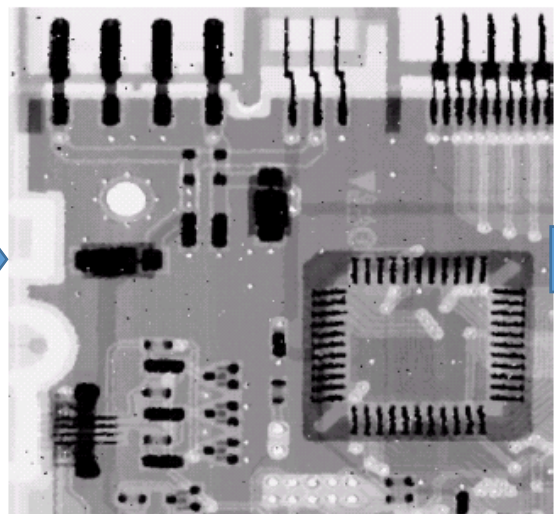
5.3 只存在噪声的复原 —— 空间滤波

• 中值滤波

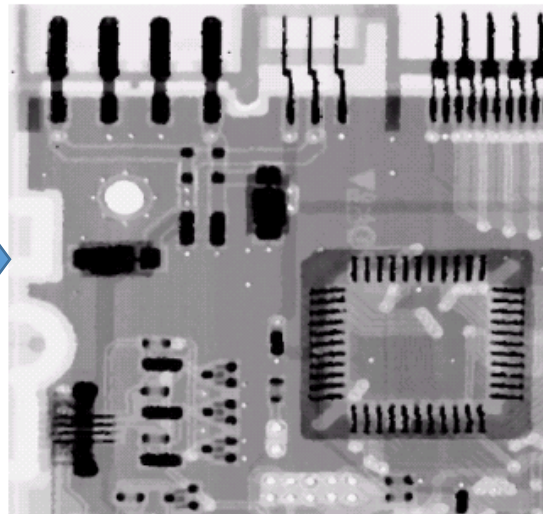
椒盐噪声图像



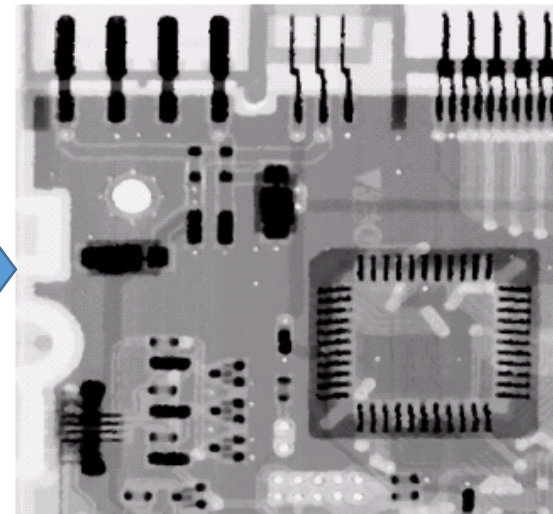
3×3



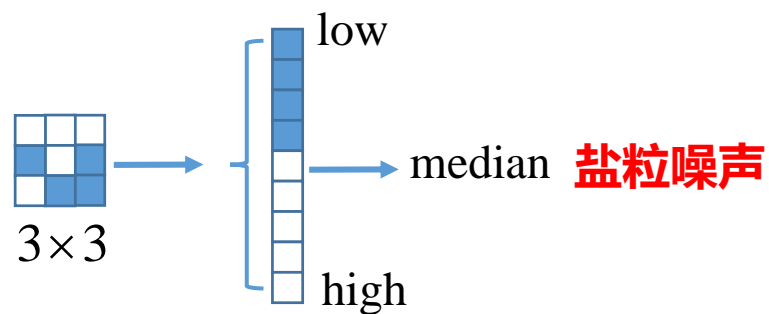
3×3



3×3



$$P_a = P_b = 0.1$$



5.3 只存在噪声的复原 —— 空间滤波

- 统计排序滤波器

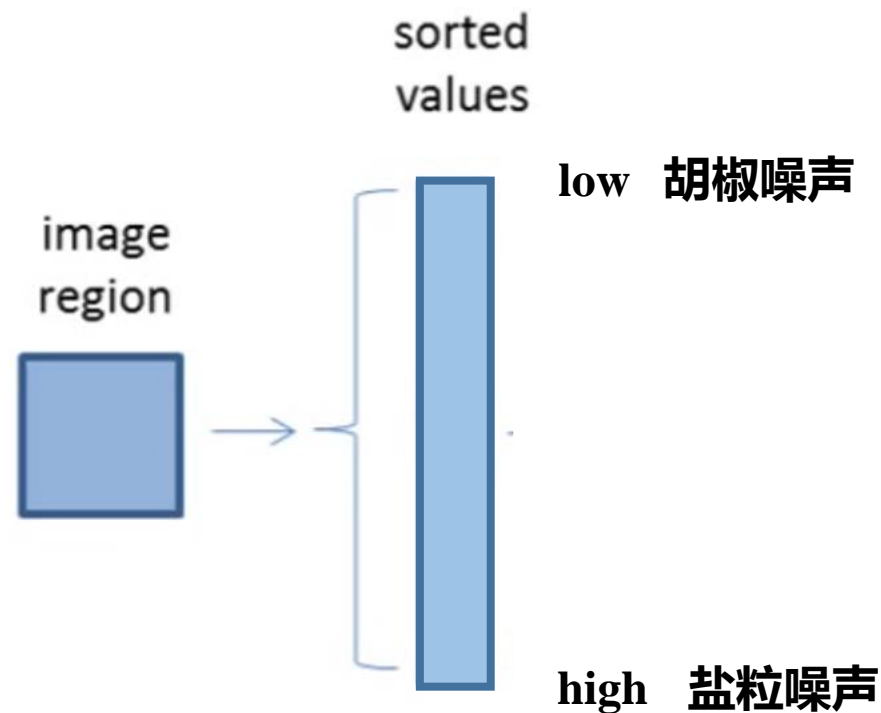
- 最大值和最小值滤波器

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

■ 适用于胡椒噪声

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

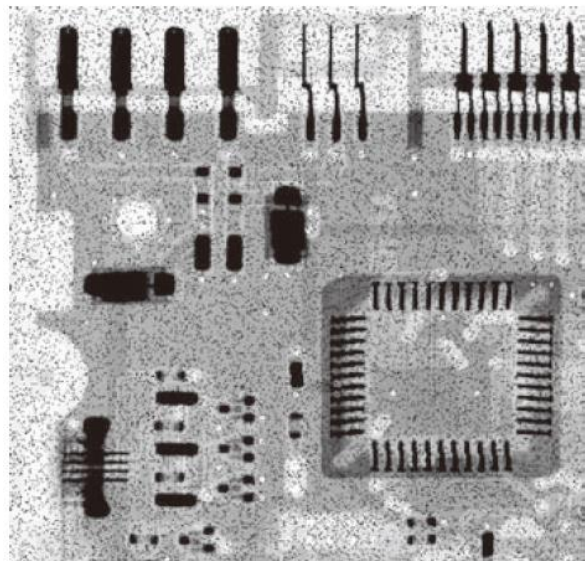
■ 适用于盐粒噪声



5.3 只存在噪声的复原 —— 空间滤波

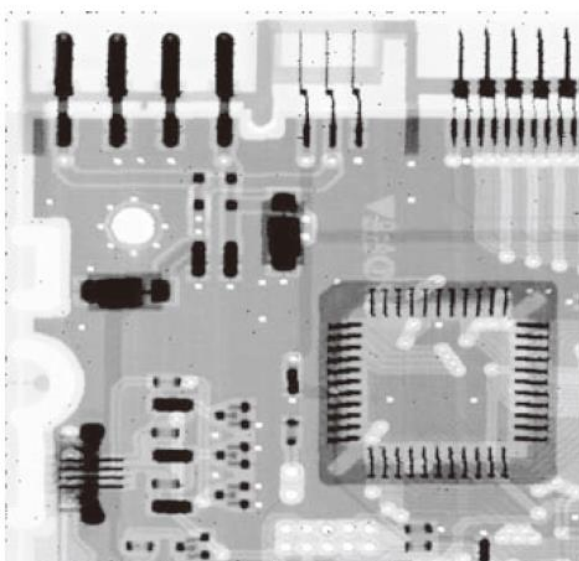
- 最大值和最小值滤波

胡椒噪声图像

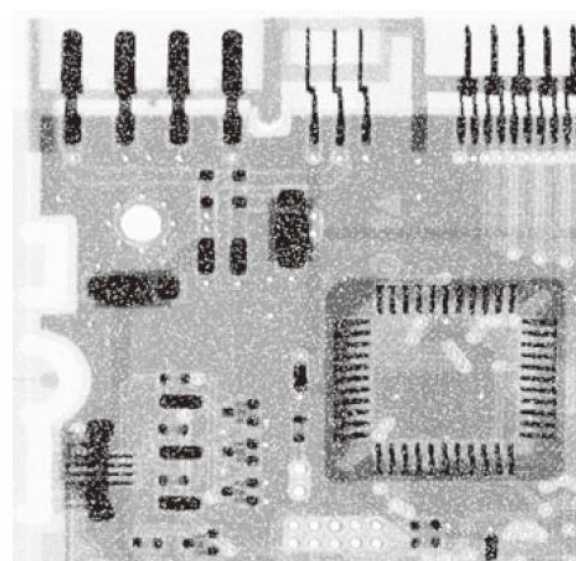


$$P_a = 0.1$$

最大滤波器

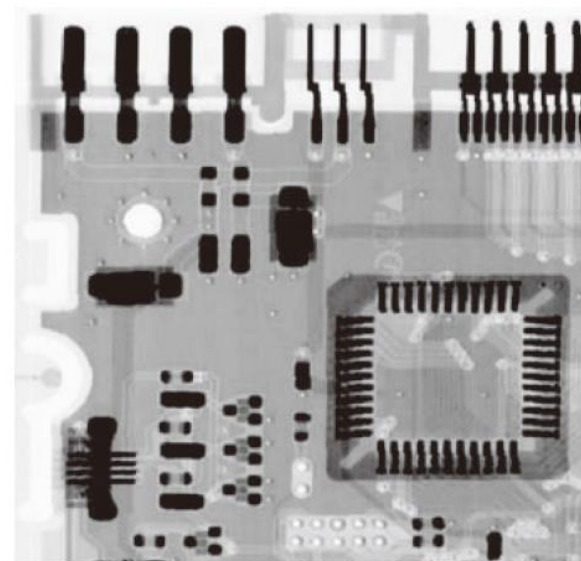


盐粒噪声图像



$$P_b = 0.1$$

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

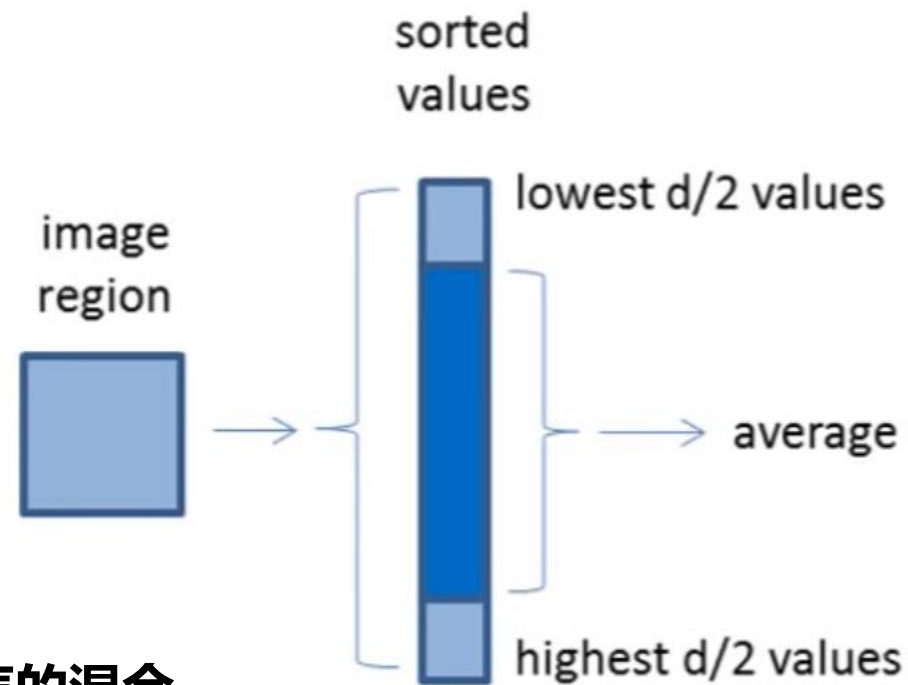
- Alpha-trimmed 均值滤波

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

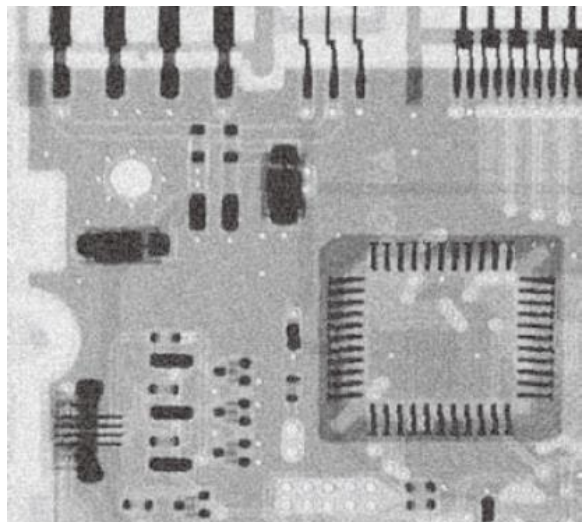
其中 $d \in [0, mn - 1]$ 高斯噪声和椒盐噪声的混合

当 $d=0$ 时 \longrightarrow **算数均值滤波器** 高斯噪声

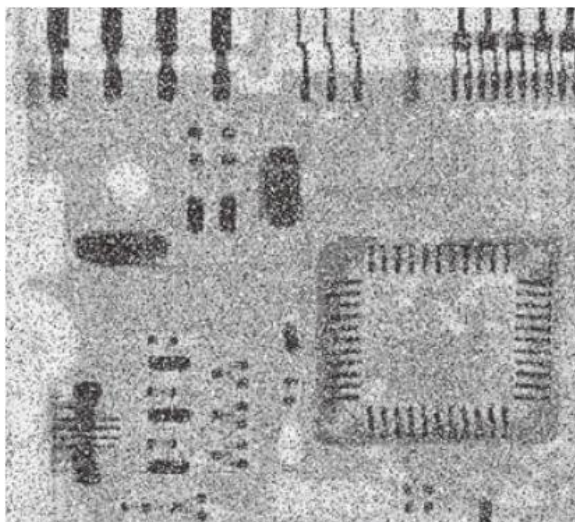
当 $d=mn-1$ 时 \longrightarrow **中值滤波器** 椒盐噪声



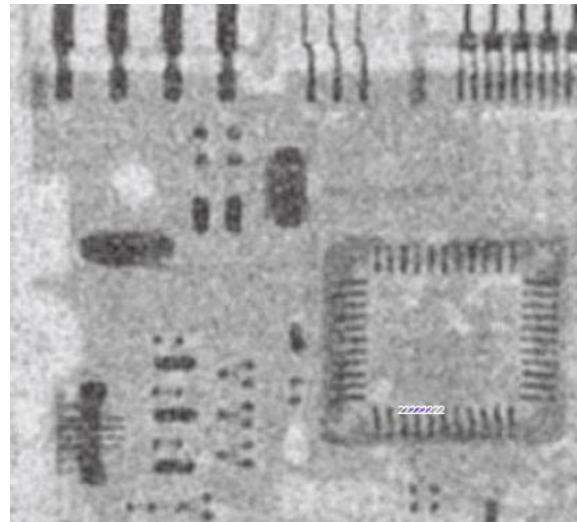
引入均匀噪声的图像



均匀噪声+椒盐噪声

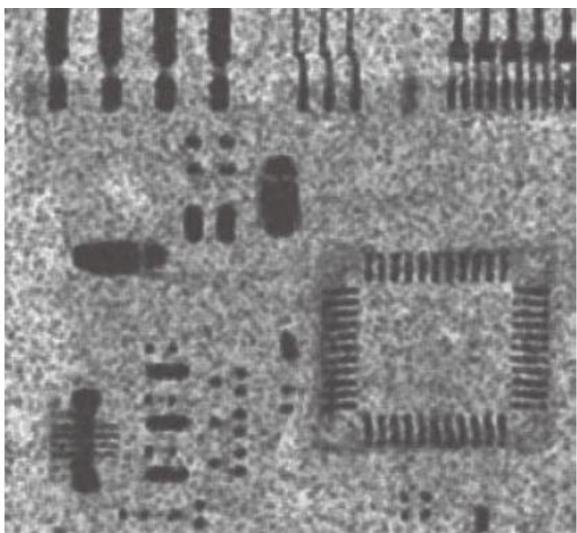


算数均值

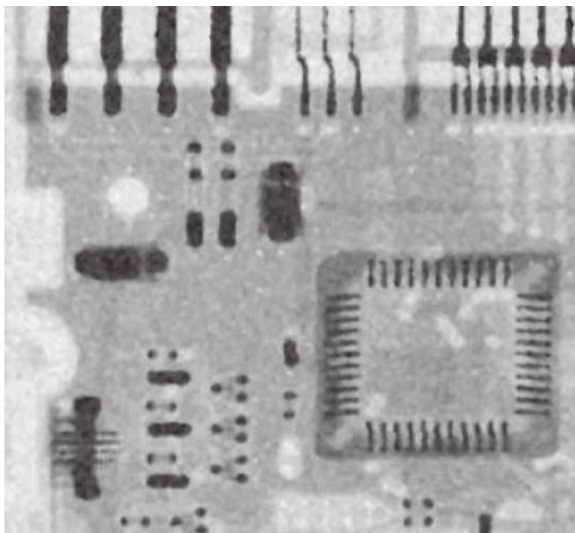


5×5 filter

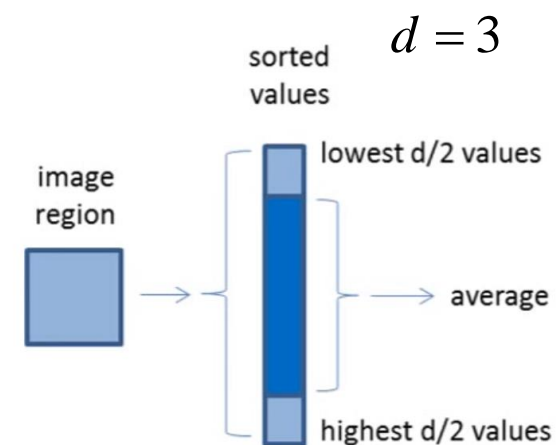
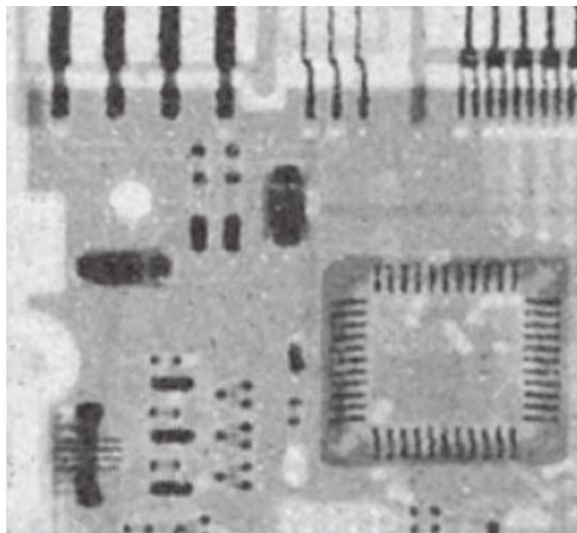
几何均值



中值滤波

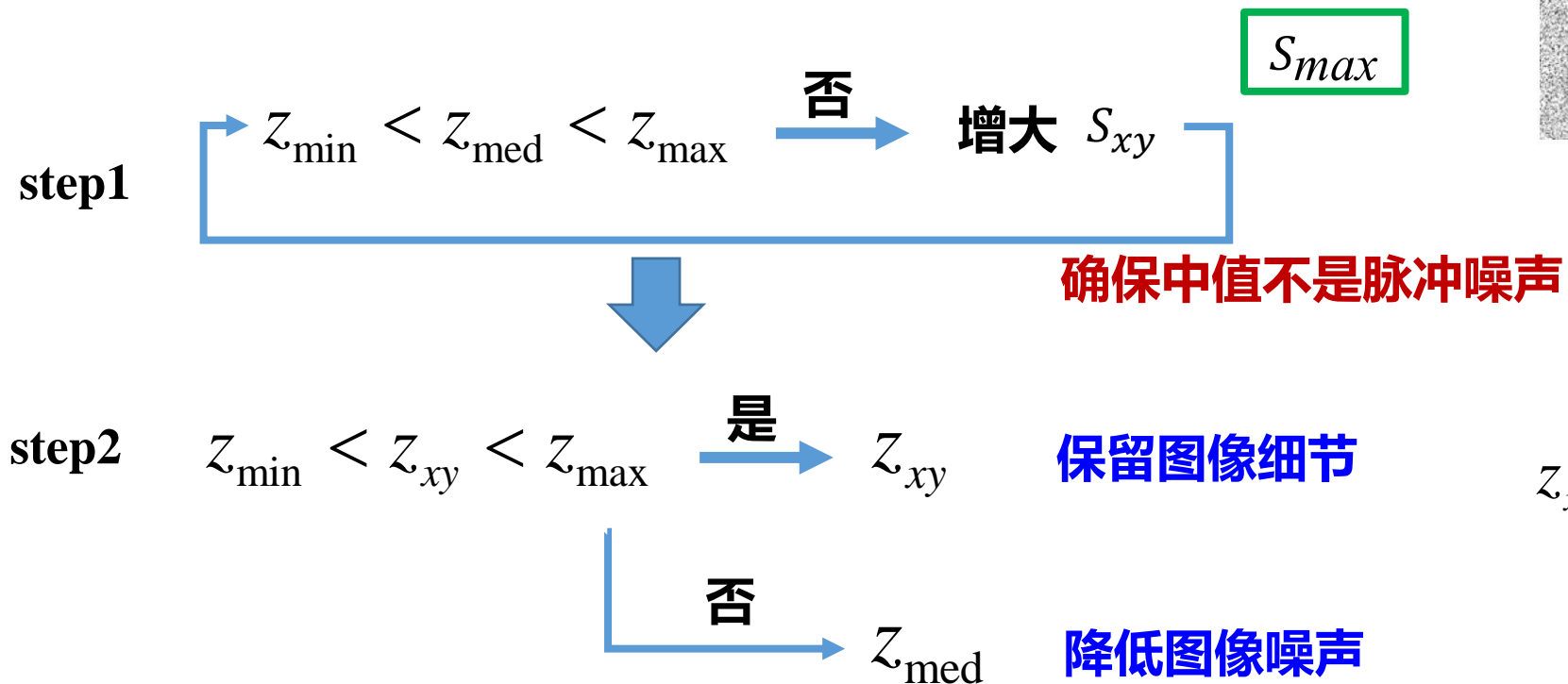
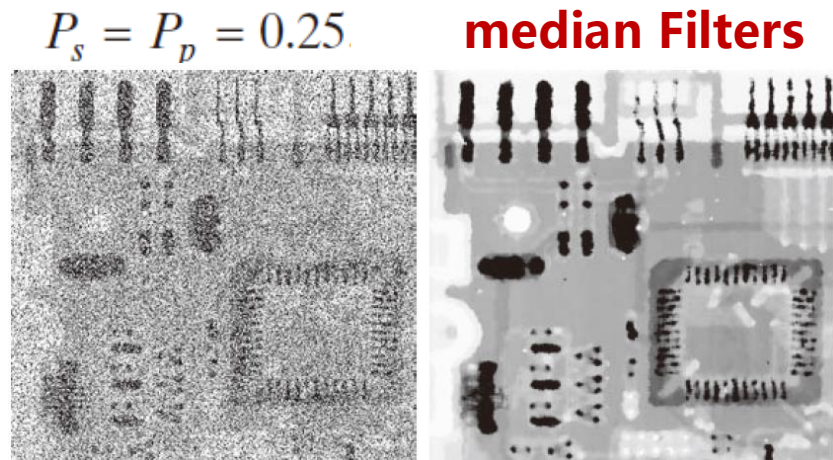


Alpha-trimmed 均值滤波

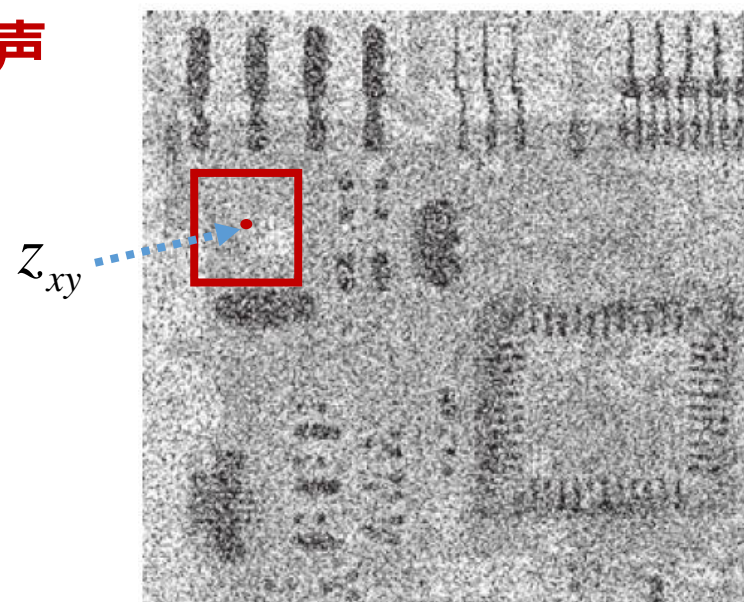


5.3 只存在噪声的复原 —— 空间滤波

• 自适应中值滤波 Adaptive median Filters

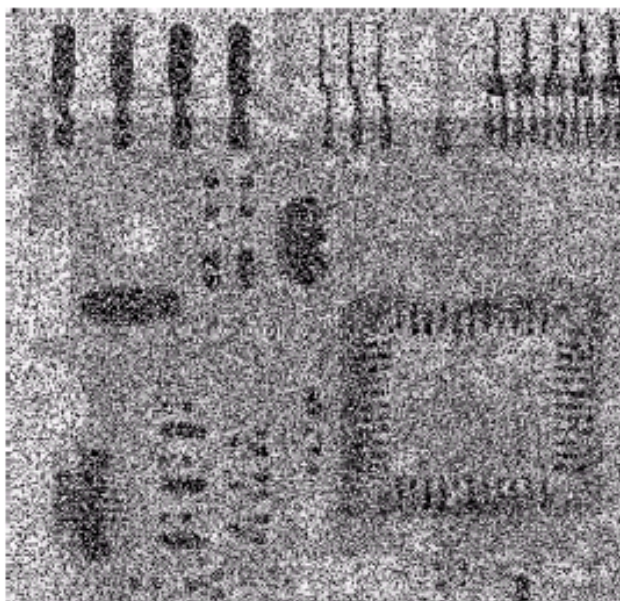


可以处理更大概率脉冲噪声！

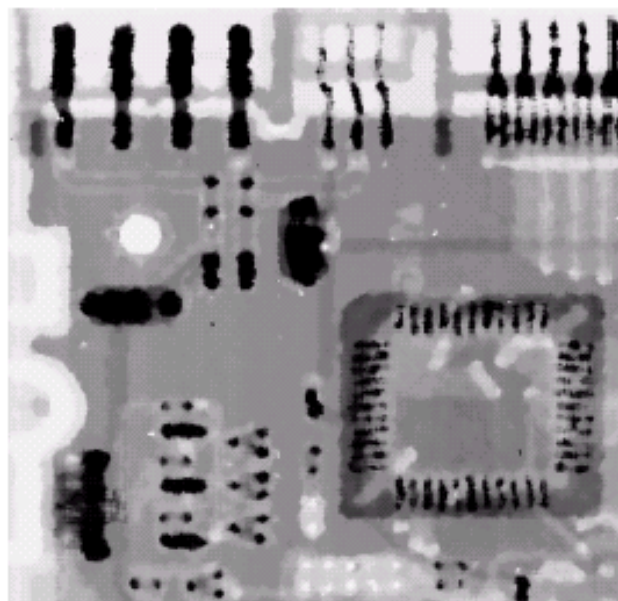


5.3 只存在噪声的复原 —— 空间滤波

椒盐噪声 $P_s = P_p = 0.25$

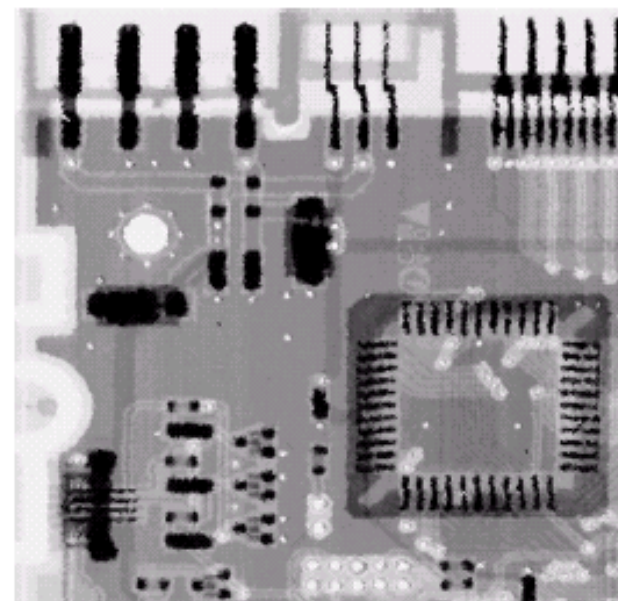


中值滤波



7×7

自适应中值滤波



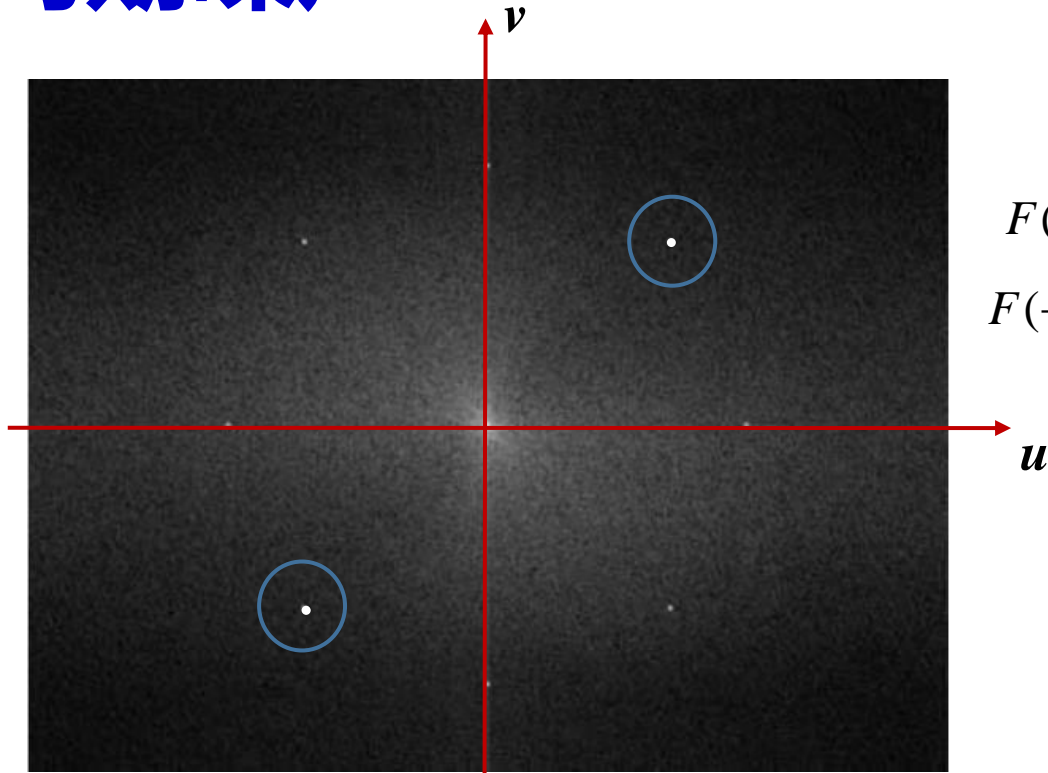
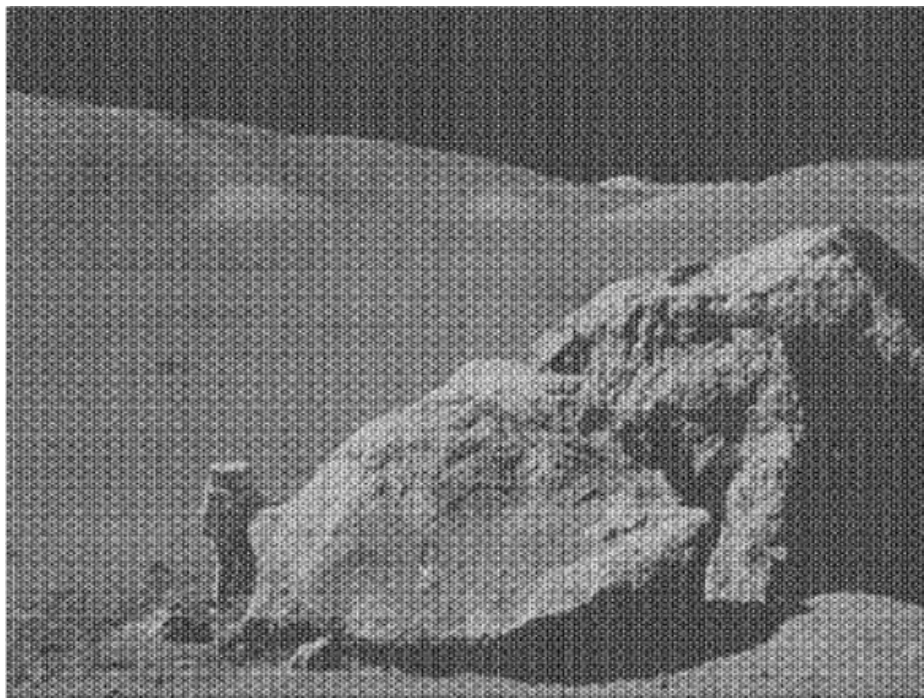
$S_{max} = 7 \times 7$

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



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

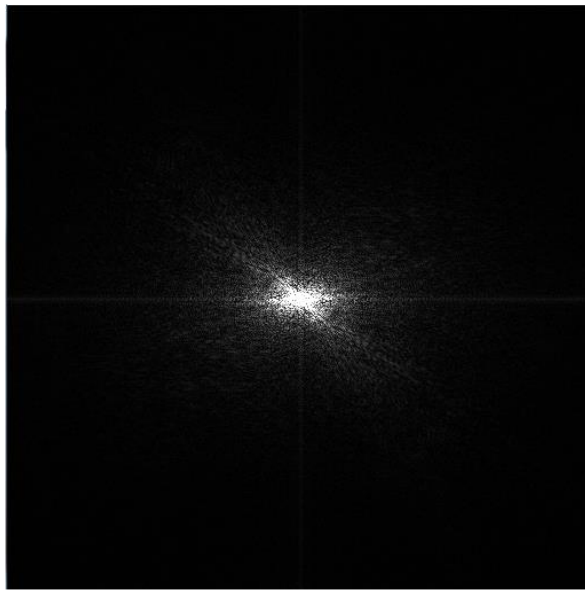
$$F(-u_0, -v_0) = a - bi$$

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

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

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

5.4 用频率域滤波消除周期噪声

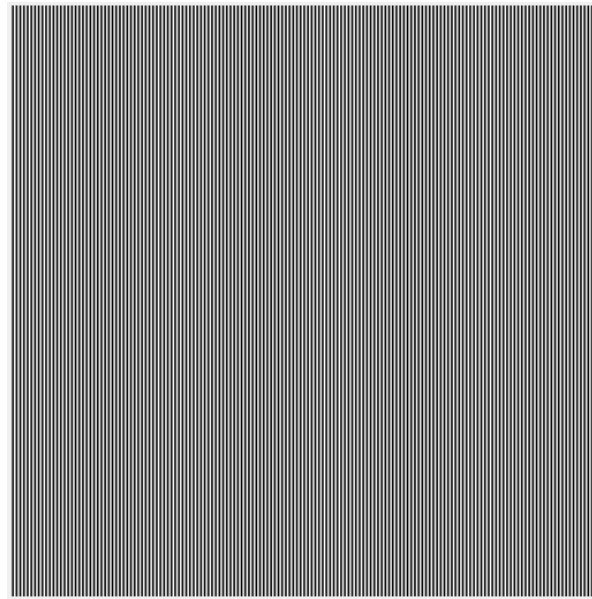
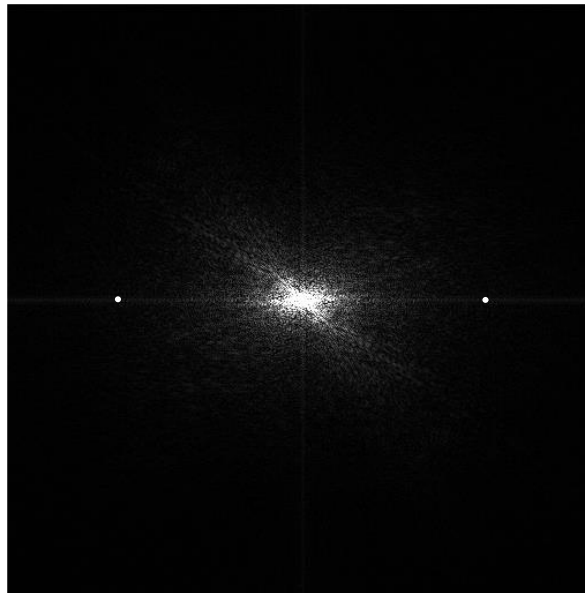


$$F(u_0, v_0) = a + bi$$

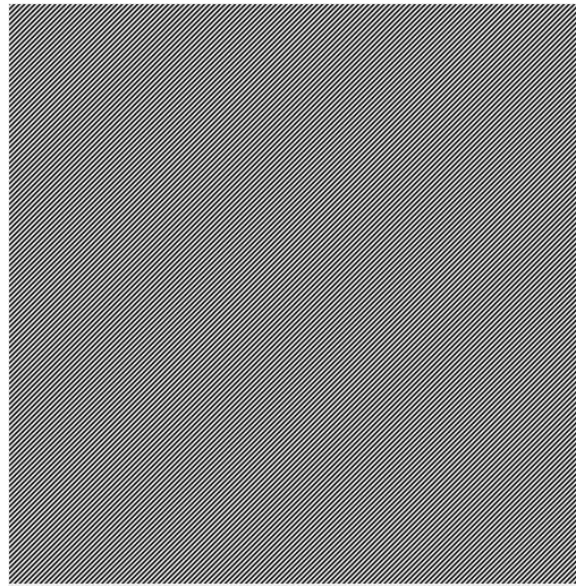
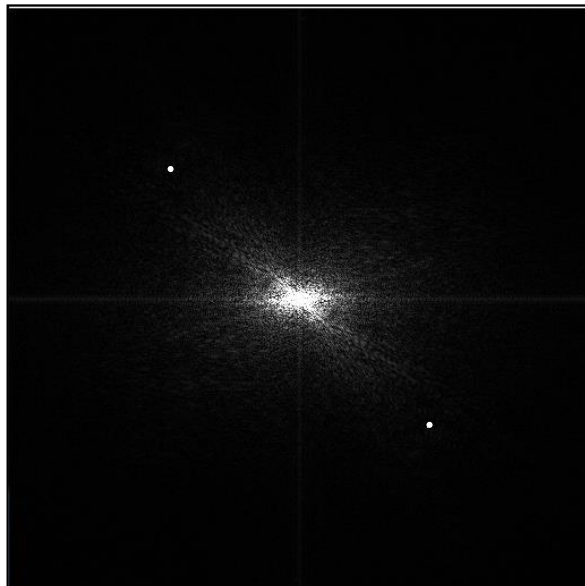
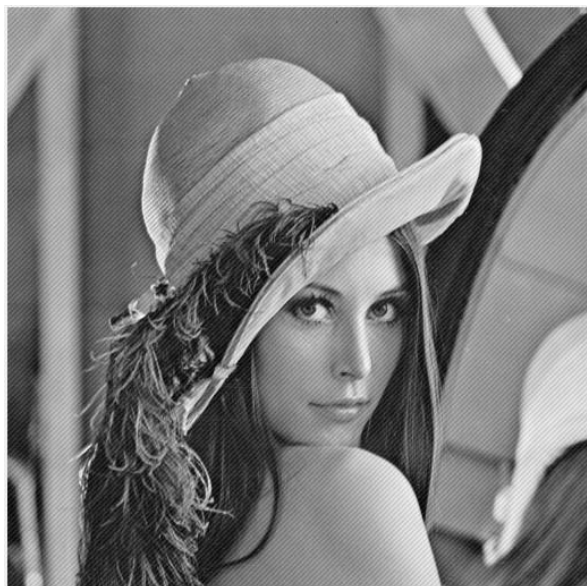
$$F(-u_0, -v_0) = a - bi$$



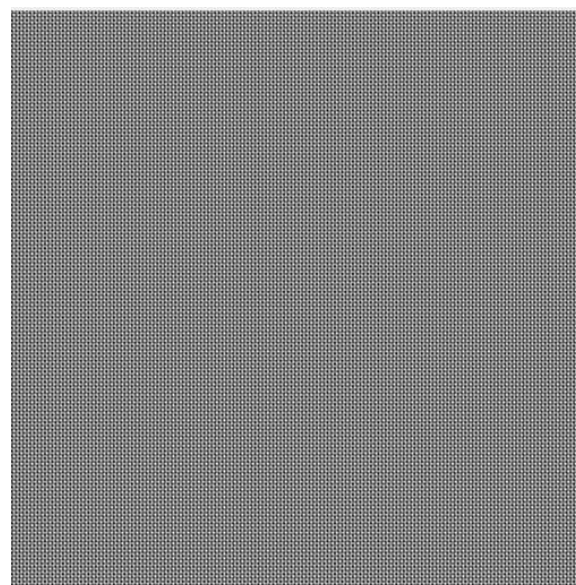
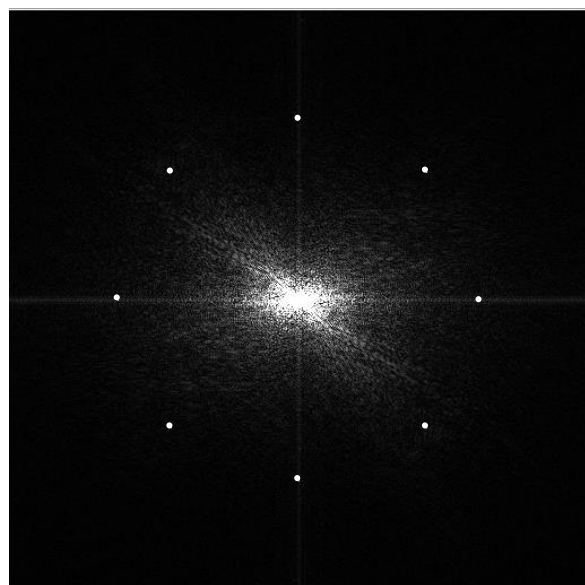
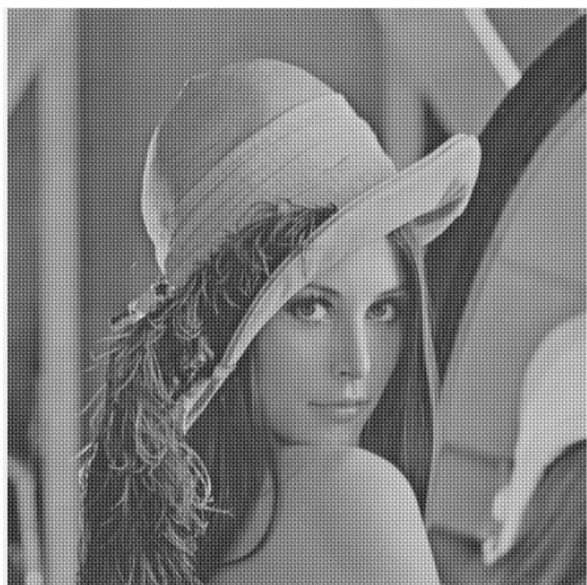
$$f(x, y) = 2a \cos\left(\frac{u_0 x}{M} + \frac{v_0 y}{N}\right) - 2b \sin\left(\frac{u_0 x}{M} + \frac{v_0 y}{N}\right)$$



5.4 用频率域滤波消除周期噪声

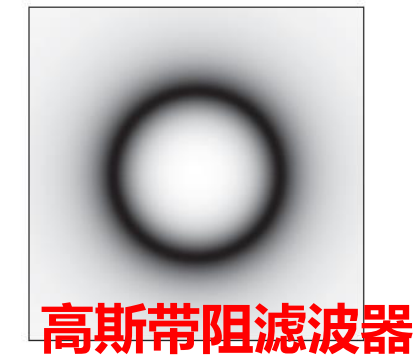
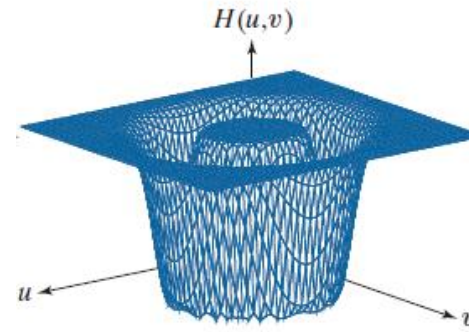
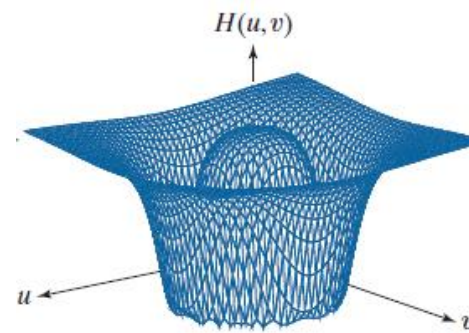
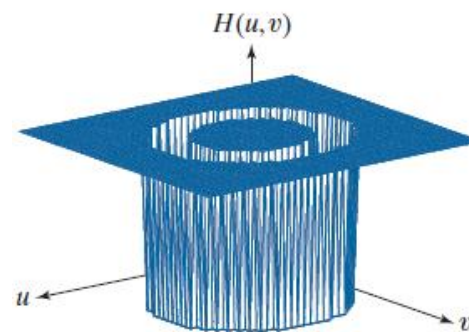
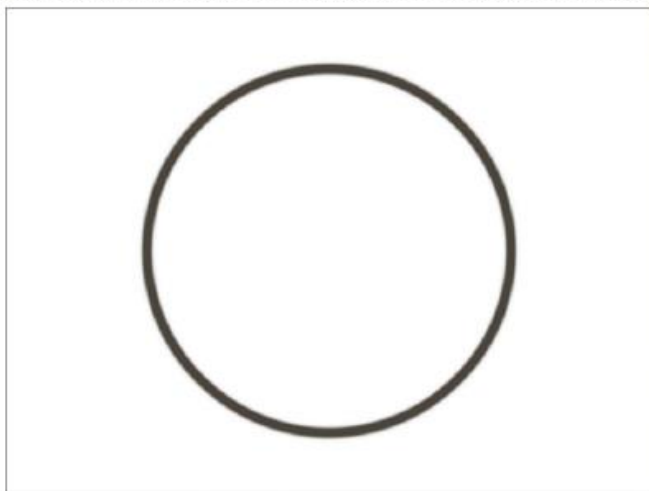
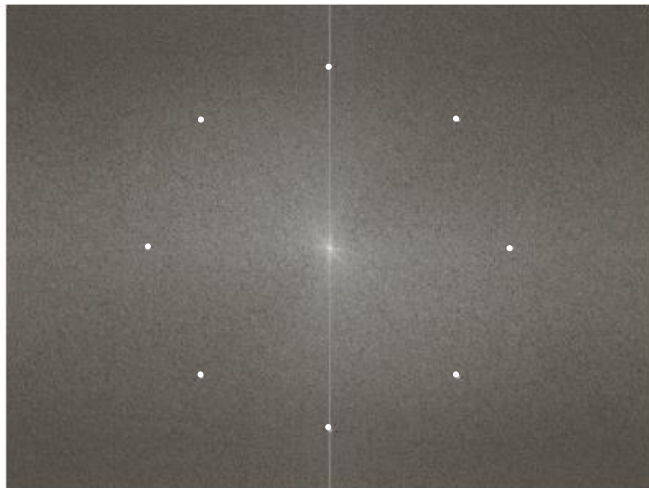
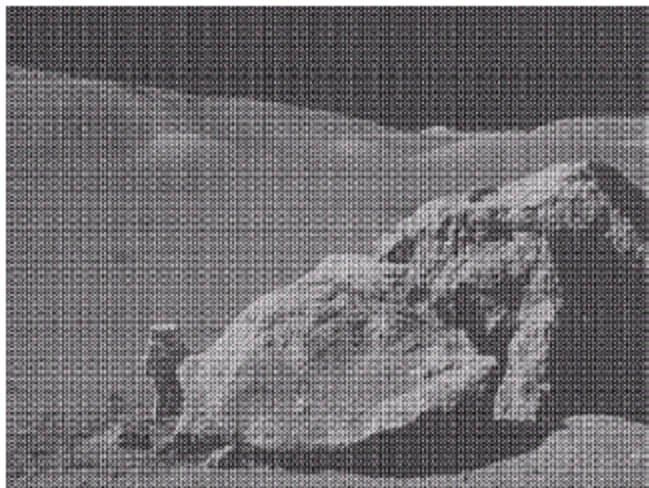


用频率域技术可以有效
分析并滤除周期噪声



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



5.4 用频率域滤波消除周期噪声——限波滤波器

