

# Digital Image Processing and Pattern Recognition

E1528

Fall 2022-2023

Lecture 9



## Image Restoration

**INSTRUCTOR**

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## ➤ Contents

- Image Restoration
- Noise Types
- Recovering From Noise
- Estimating the degradation function
- Image Reconstruction from Projections

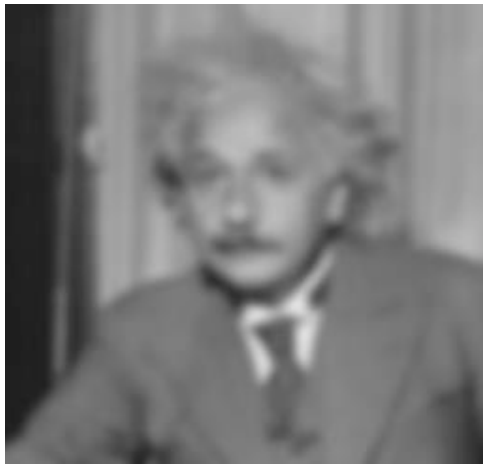


# Image Restoration

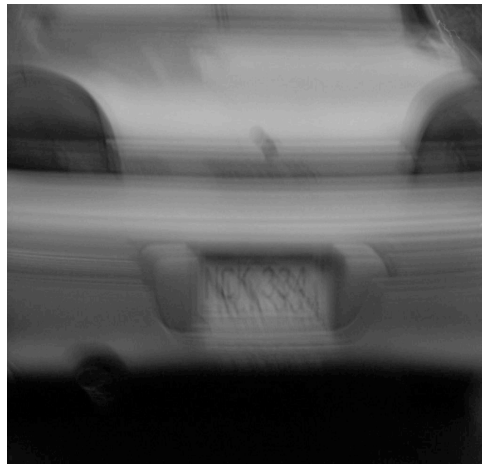
- ❑ Image restoration aims to improve an image that has suffered from linear degradation.
- ❑ Degradation considered noise in the acquisition, transmission problems, etc.
- ❑ The purpose of image restoration is to reconstruct the original image from a degraded observation.

# Distortion may arise from

**Out of focus blur**



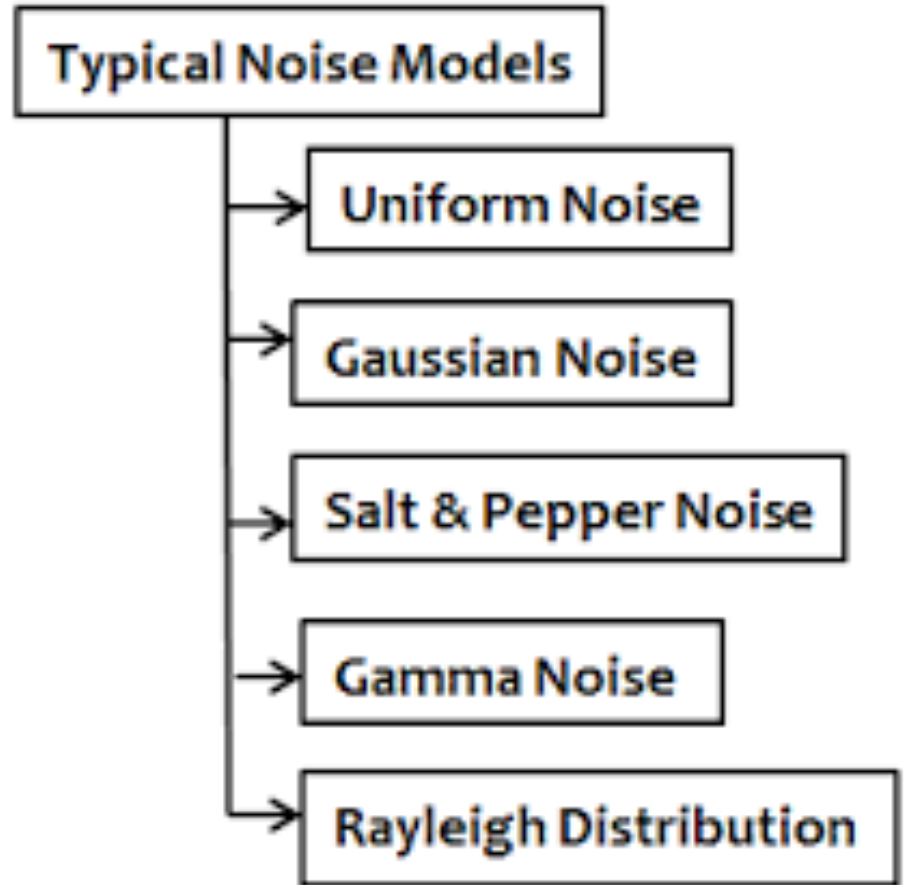
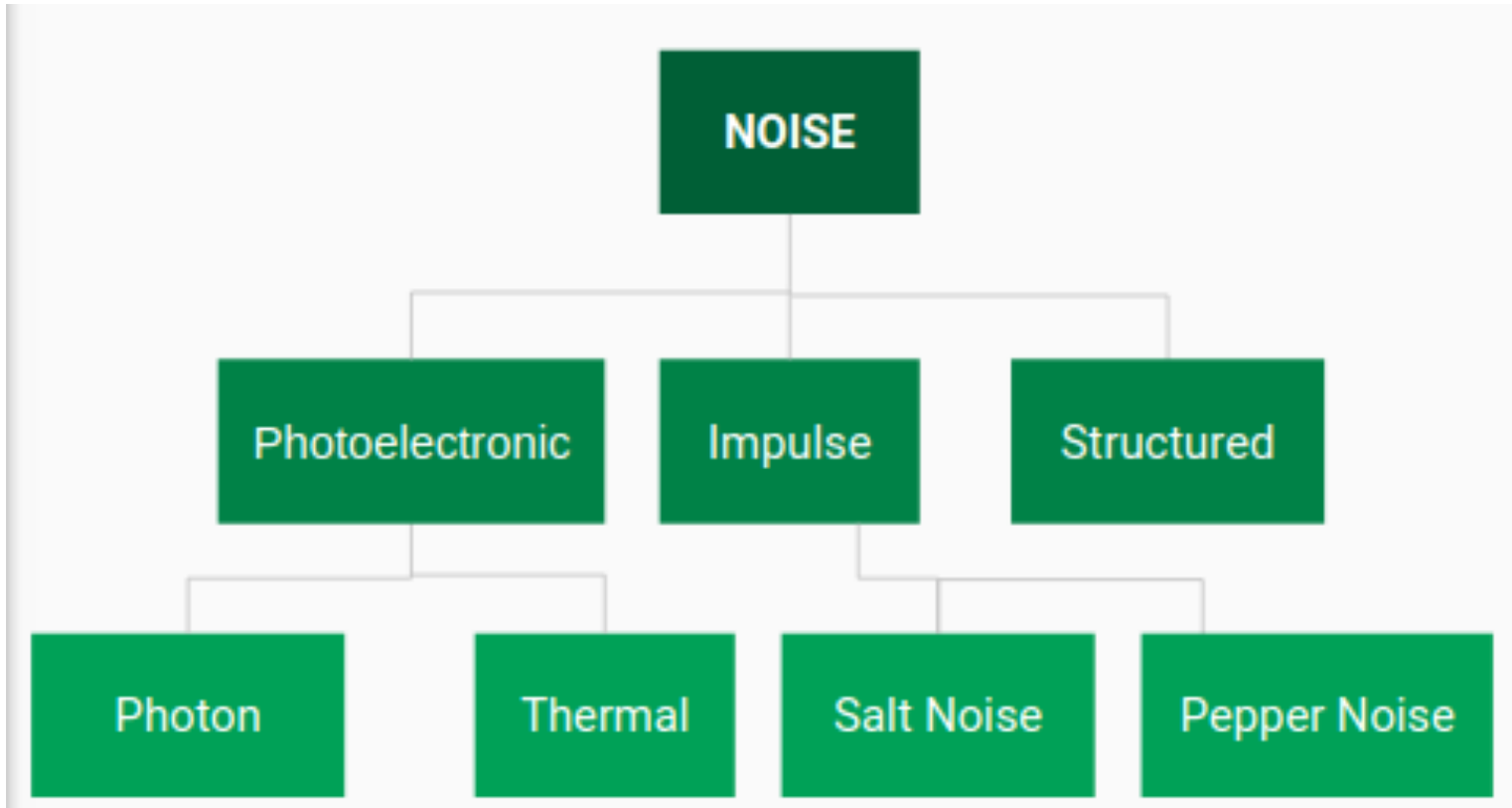
**Motion Blur**



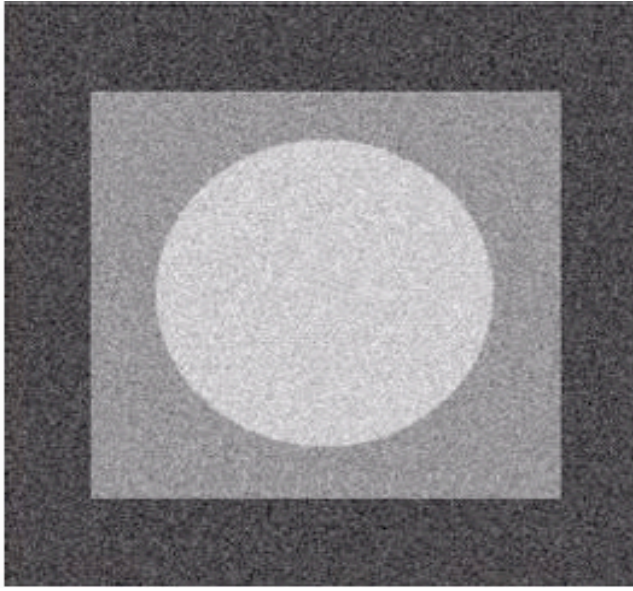
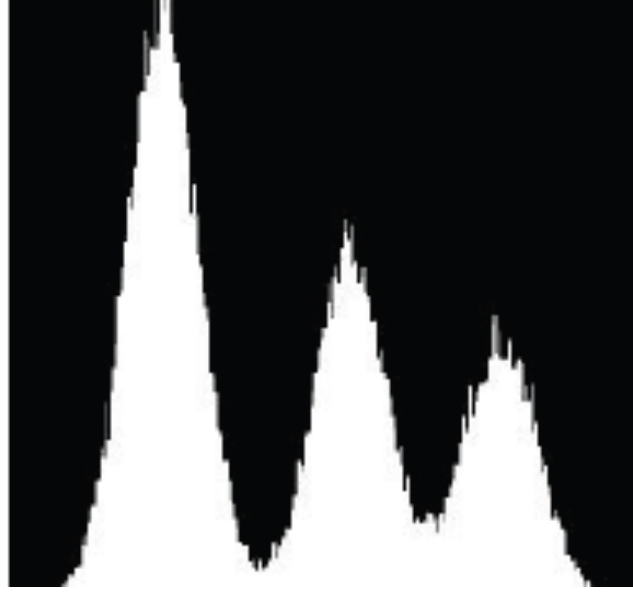
**Additive Noise**



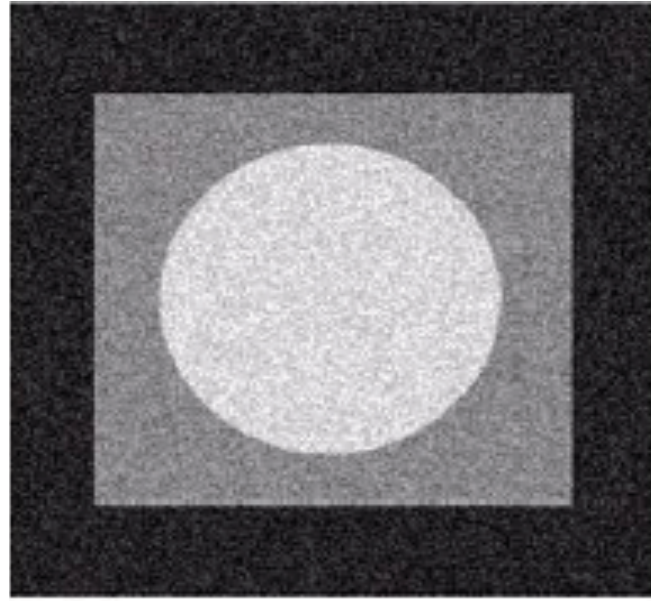
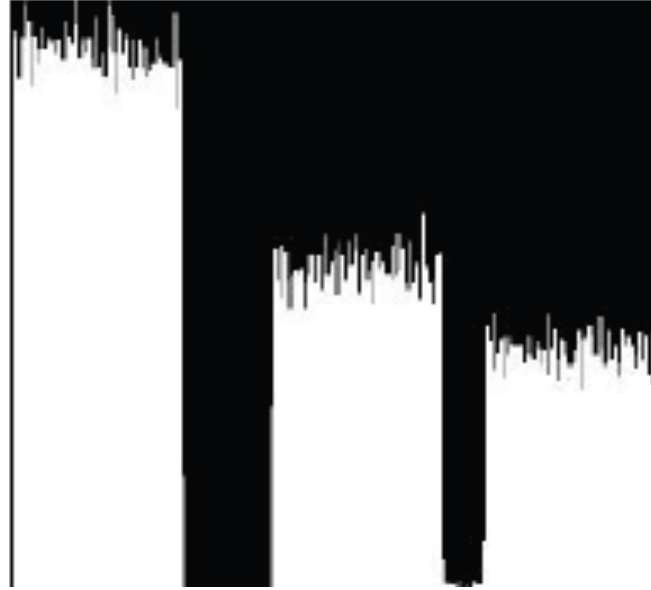
# Noise Types



**Gaussian  
Noise**



**Uniform  
Noise**



# Recovering From Noise

## Spatial Filters

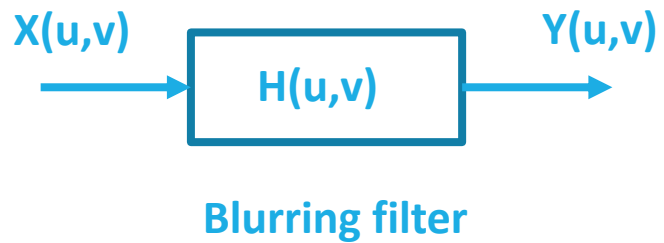
- Gaussian Filter
- Median Filter
- Mean Filter

## Frequency Domain Filters

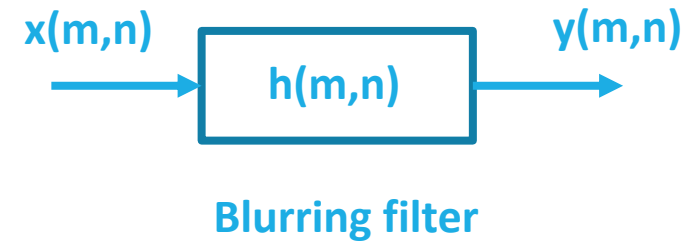
- Notch Filter
- LPF
- HPF

# Blur Model

Frequency Domain

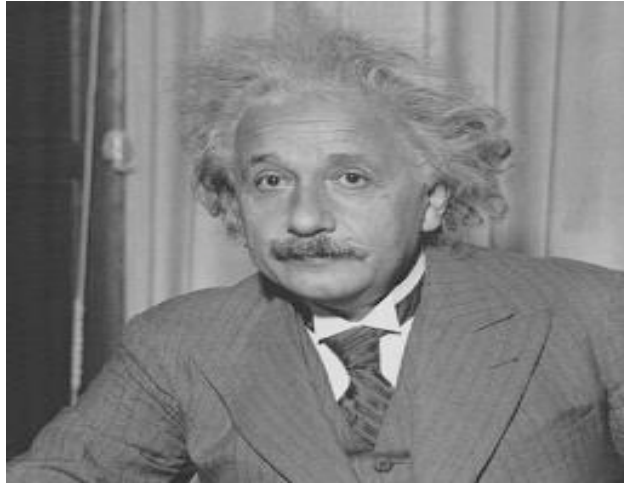


Spatial Domain





# Blurring Effect



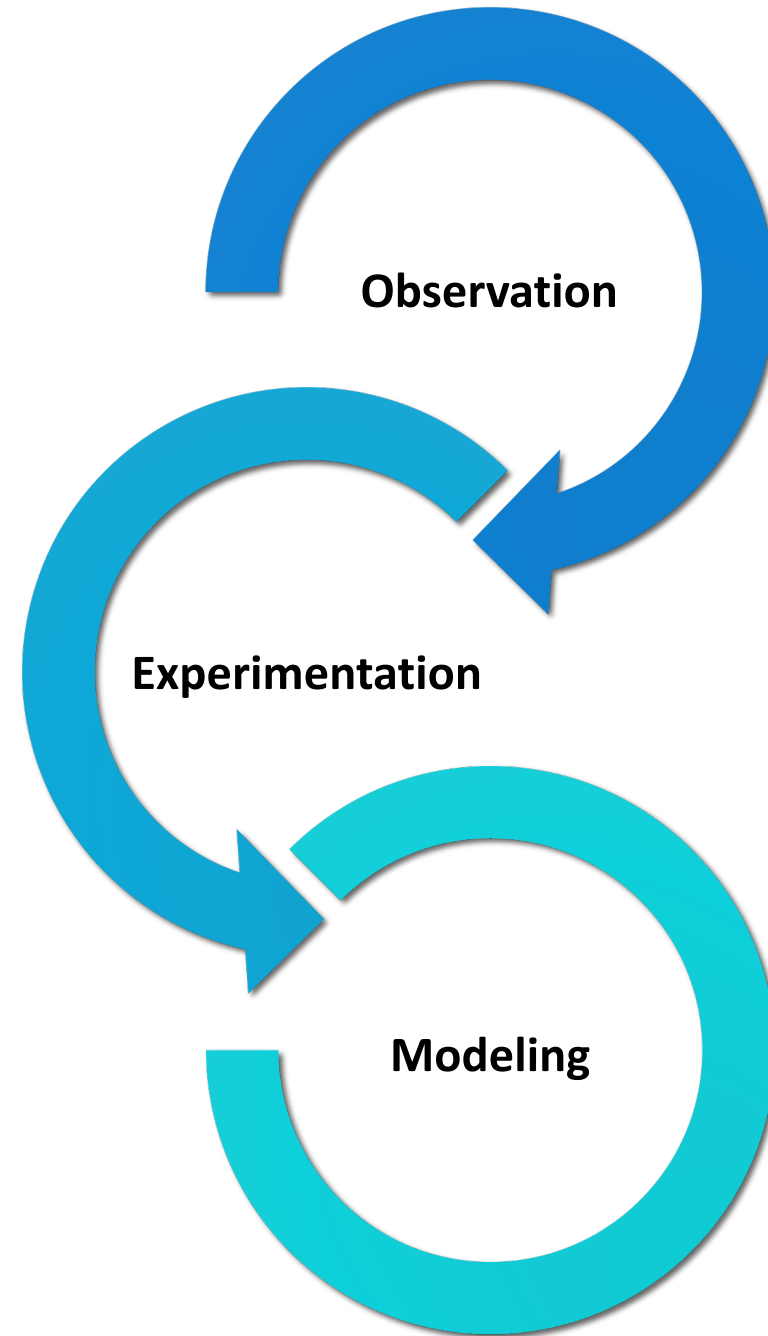
Gaussian Blur



Motion Blur



# Estimating the degradation function



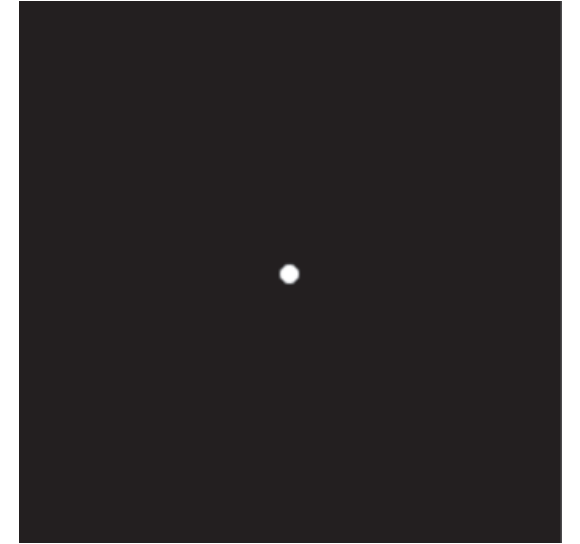
## Estimating by observation

- ❑ Finding the information from the observed image.
- ❑ Identify a portion of the image that is visually unblurred  $[k(x,y)]$  and observed image  $[g(x,y)]$ .
- ❑ The degradation function can be estimated by applying an **inverse Fourier transform** to the ratio of the Fourier transform of the **observed image** and the **sub image**.

$$H(U, V) = \frac{G(U, V)}{K(U, V)}$$

# Estimating by Experimentation

- ❑ Obtain the impulse response by imaging small dot of light.
- ❑ Knowing that the Fourier transform of the impulse is a constant (**A**).
- ❑ The degradation function can be estimated by applying an **inverse Fourier transform** to the ratio of the Fourier transform of the **observed image** and the **impulse function**.



$$H(U, V) = \frac{G(U, V)}{A}$$

# Estimating by Modeling

- ❑ A set of equations that approximate the real system.
- ❑ Scenario 1: Complete knowledge about the blur available.
- ❑ Scenario 2: There is only a partial knowledge of the blurring function available.
- ❑ Scenario 3: There is no knowledge about the blurring function (**Blind Restoration**).

$$H(u, v) = e^{-k(u^2 + v^2)^{5/6}}$$

# Image Restoration methods can be divided into two classes

## Blind

- In which the blurring operator is **unknown**

## Non-blind

- In which the blurring operator is **known**

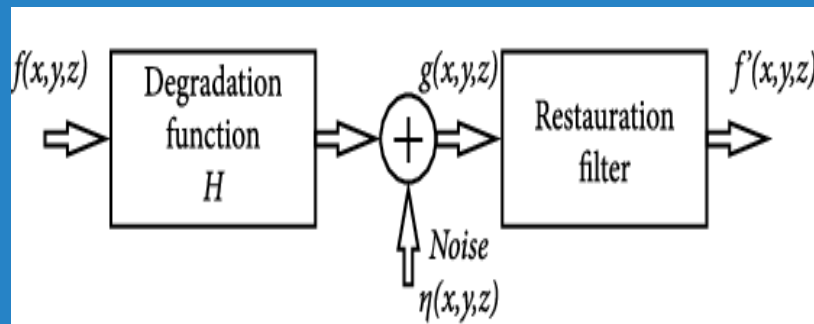
# Non-Blind Deblurring

- Inverse Filter
- Pseudo inverse Filter
- Wiener Filter

# Blind Deblurring

- Iterative Blind Deblurring
- Non-Iterative Blind Deblurring

# Degradation Model



- ❑ The image can be degraded using Filter and Noise.
- ❑ The degraded image can be described by the following

equation:

$$g = H \times f + \eta$$

**Where:**

**g..... Degraded or blurred image**

**H..... Degradation Function**

**f..... The Original image**

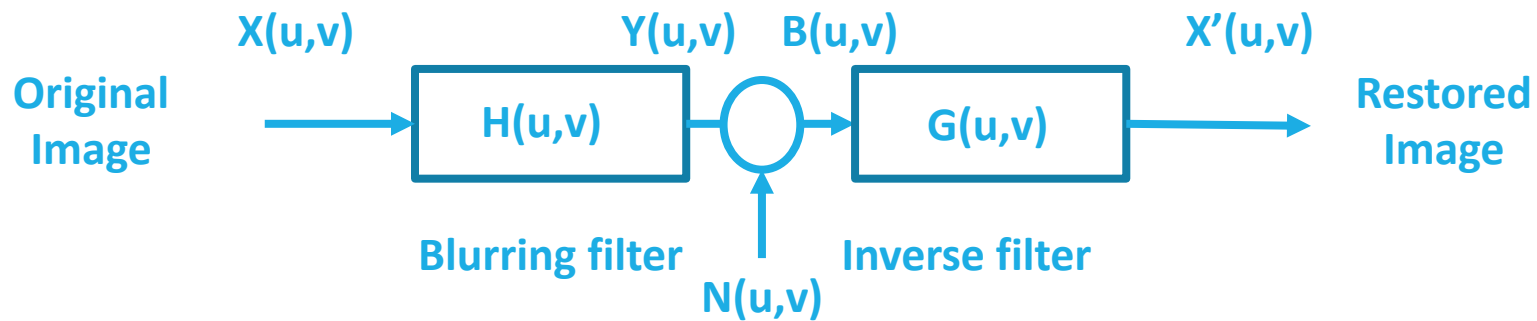
**η ..... Additive Noise**



# Non-Blind Deblurring

## i. Inverse Filter

$$G(\mathbf{u}, \mathbf{v}) = \frac{1}{H(\mathbf{u}, \mathbf{v})}$$

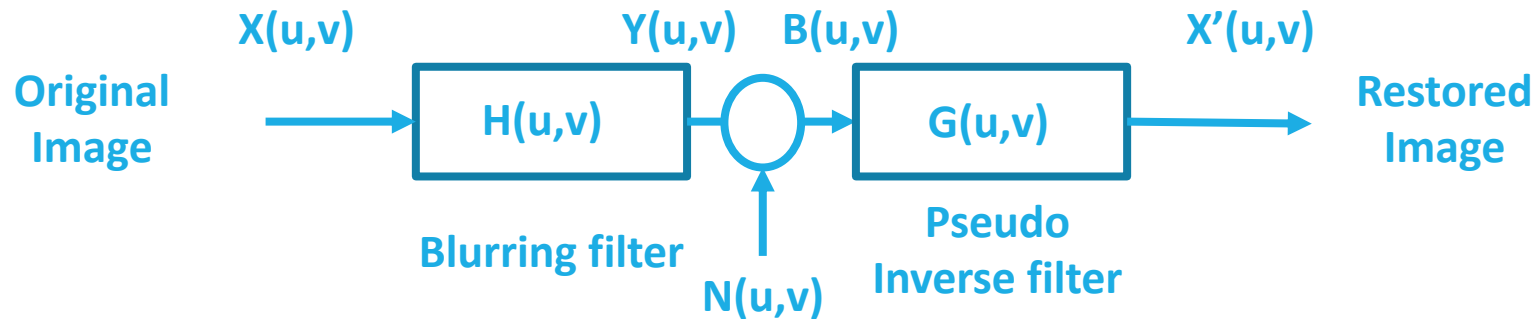


$$\therefore B(\mathbf{u}, \mathbf{v}) = X(\mathbf{u}, \mathbf{v})H(\mathbf{u}, \mathbf{v}) + N(\mathbf{u}, \mathbf{v})$$

$$\therefore X'(\mathbf{u}, \mathbf{v}) = X(\mathbf{u}, \mathbf{v}) + \frac{N(\mathbf{u}, \mathbf{v})}{H(\mathbf{u}, \mathbf{v})}$$

## ii. Pseudo Inverse Filter

$$G(\mathbf{u}, \mathbf{v}) = \begin{cases} \frac{1}{H(\mathbf{u}, \mathbf{v})} & |H(\mathbf{u}, \mathbf{v})| \geq \delta \\ \mathbf{0} & |H(\mathbf{u}, \mathbf{v})| < \delta \end{cases}$$



$$\therefore X'(\mathbf{u}, \mathbf{v}) = X(\mathbf{u}, \mathbf{v}) + \frac{N(\mathbf{u}, \mathbf{v})}{H(\mathbf{u}, \mathbf{v})}$$

### iii. Wiener Filter

$$G(\mathbf{u}, \mathbf{v}) = \frac{H^*(\mathbf{u}, \mathbf{v})}{|H(\mathbf{u}, \mathbf{v})|^2 + K}$$

$$K = \frac{\delta_W^2}{\delta_X^2}$$

→ Noise Power

→ Signal Power

For  $H(\mathbf{u}, \mathbf{v}) = 1$ :

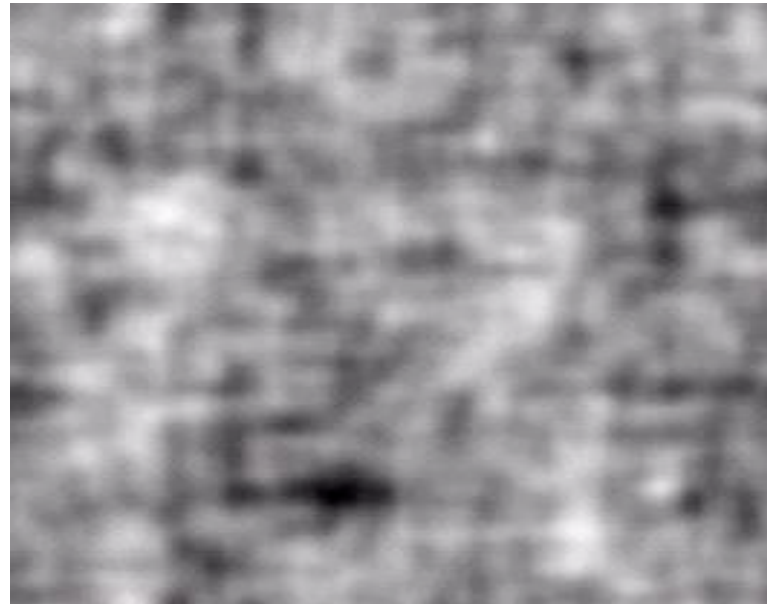
### Wiener denoising Filter

$$G(\mathbf{u}, \mathbf{v}) = \frac{1}{1 + K} = \frac{\delta_X^2}{\delta_X^2 + \delta_W^2}$$

**Blurred Image**



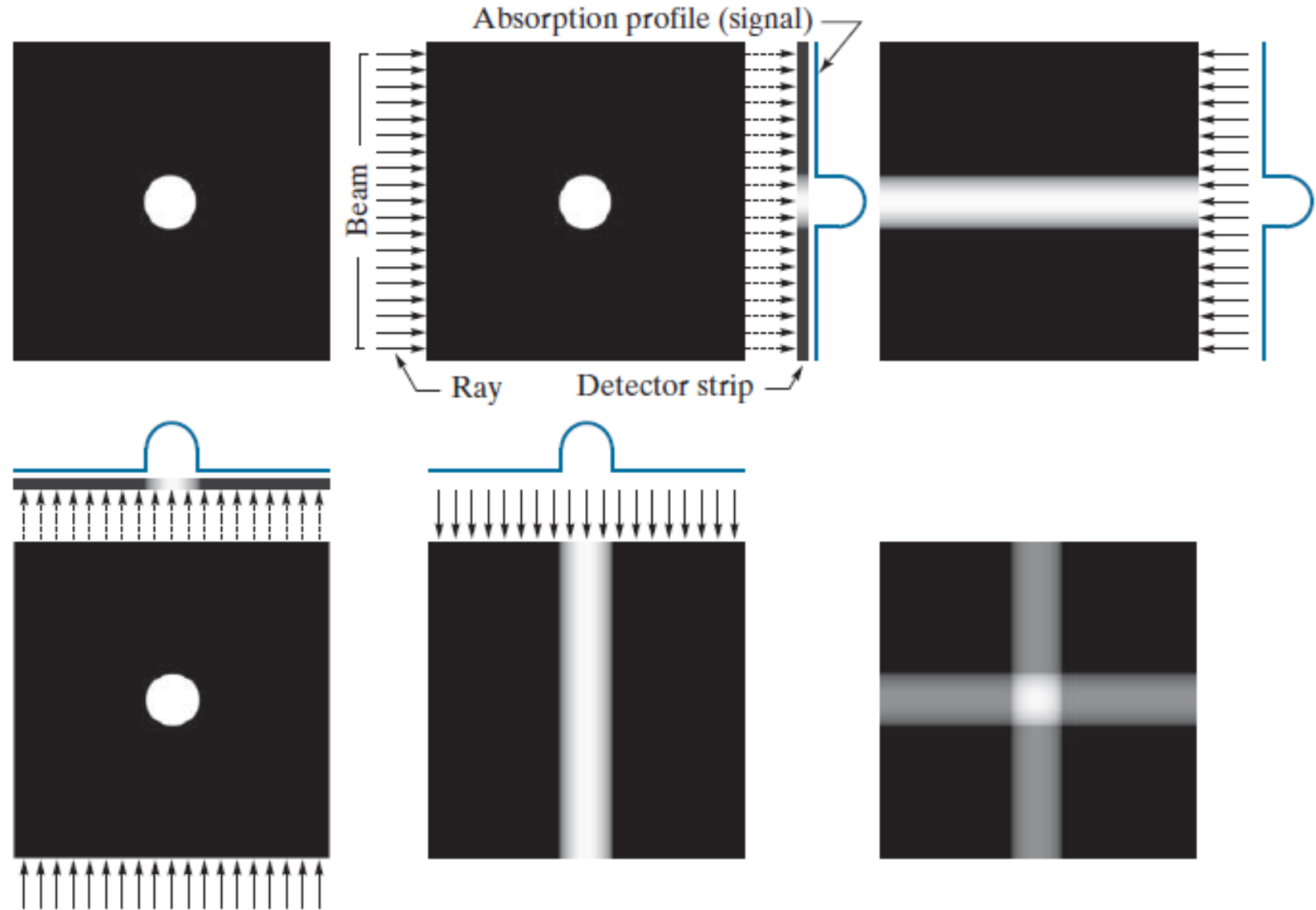
**Inverse Filtering**



**Wiener Filtering**

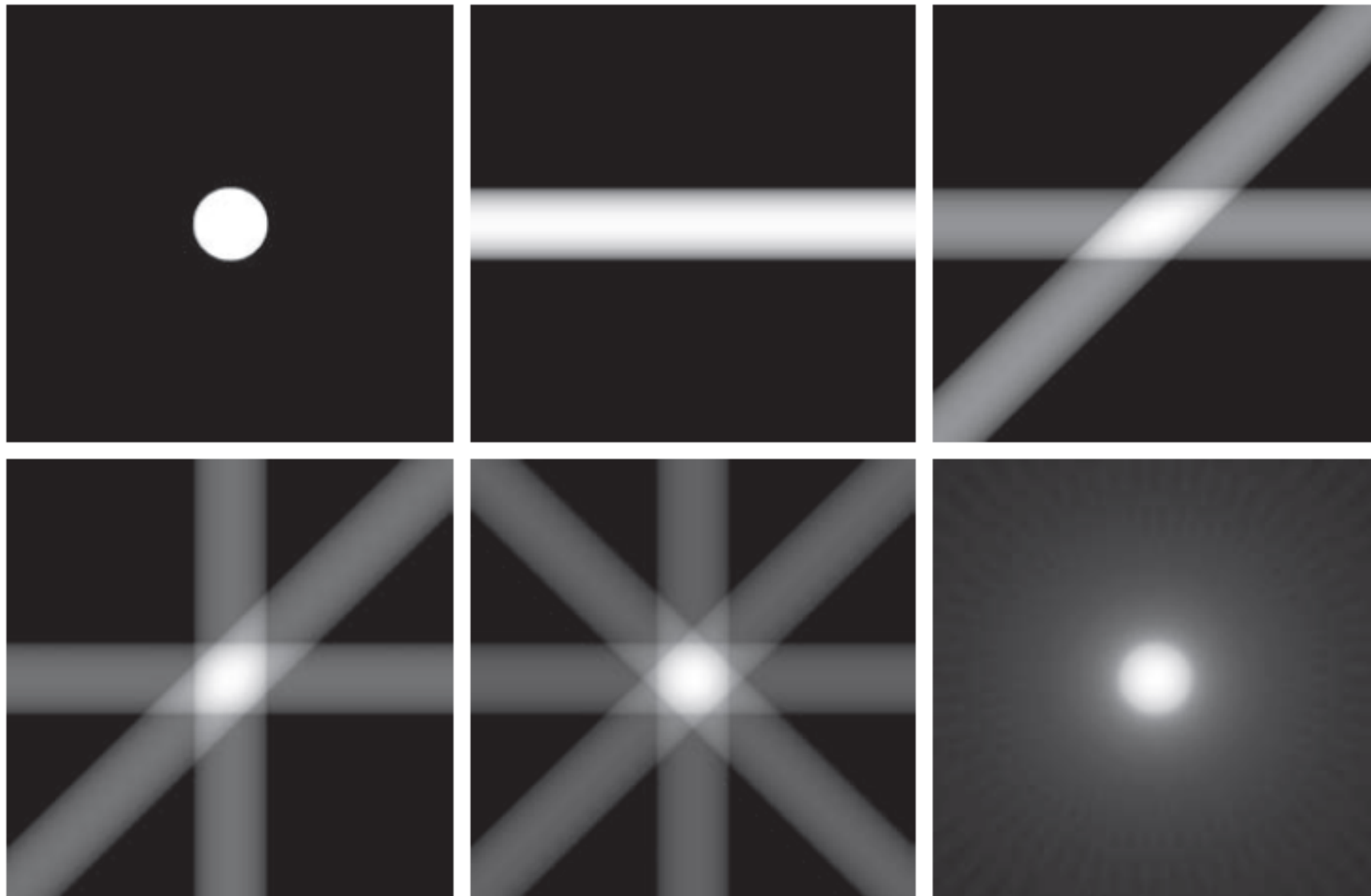


# Image Reconstruction from Projections



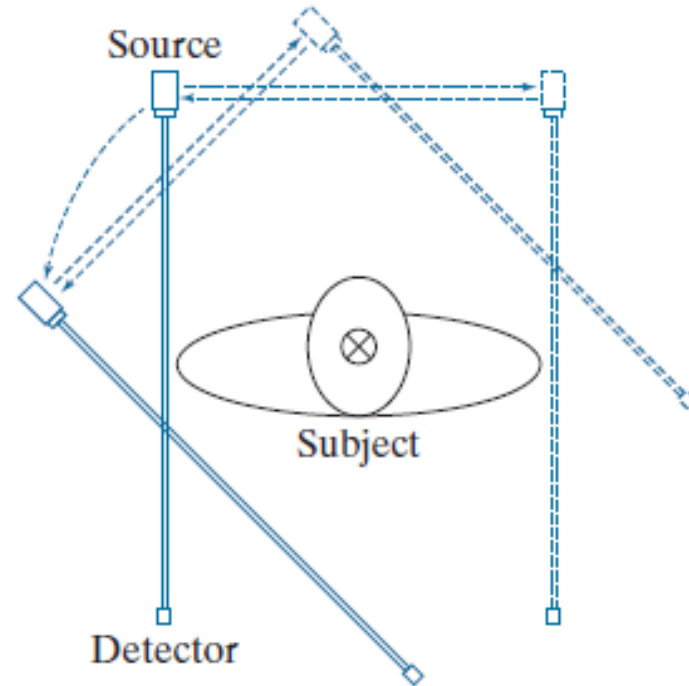
**As the number of projections increases, the amplitude strength of non-intersecting back projections decreases**

**Reconstruction with  
32 back projections  
5.625° apart.**



# CT Generations

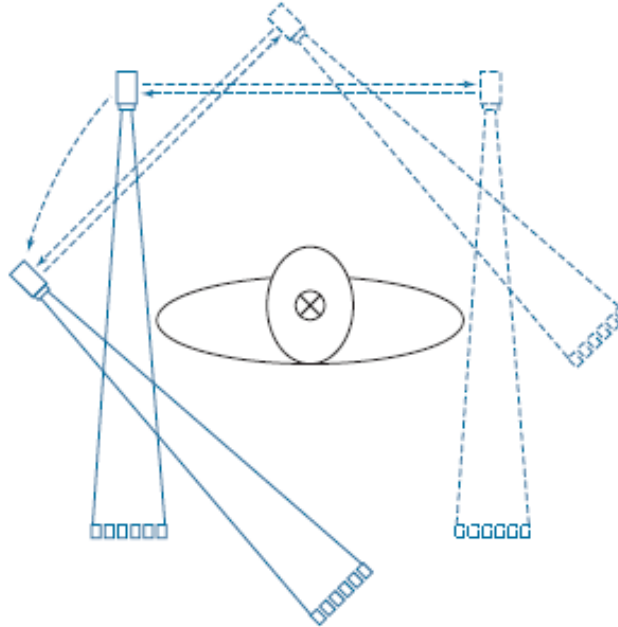
## First Generation



### Pencil Beam and single detector

- ❑ A projection is generated by measuring the output of the detector at each increment of translation.

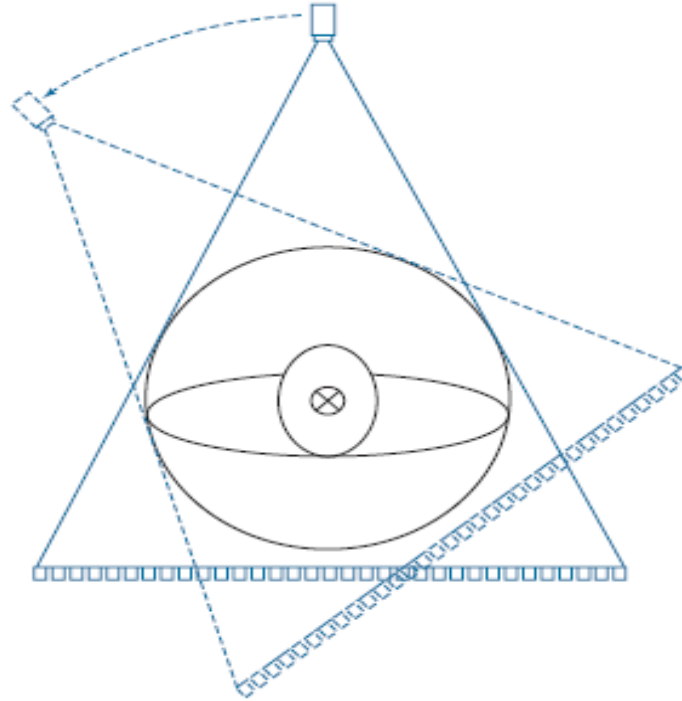
# Second Generation



- ❑ Operate on the same principle as G1 scanners, but the beam used is in the shape of a fan. This allows the use of multiple detectors.

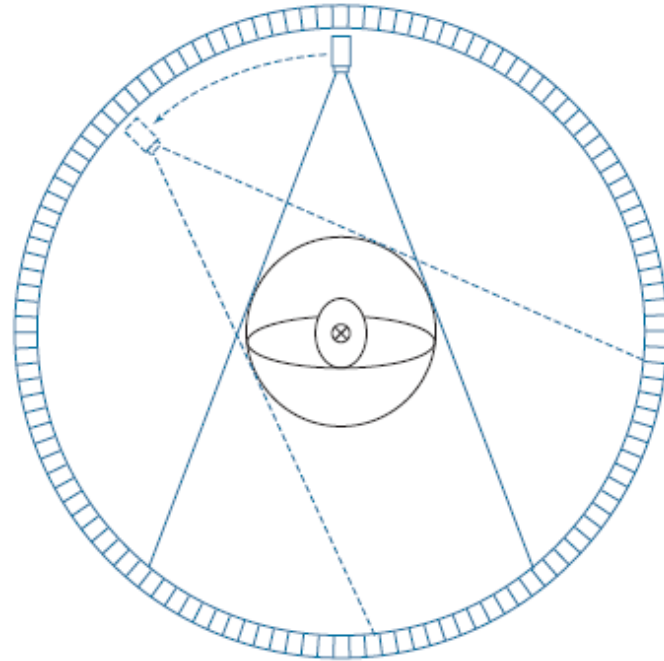


# Third Generation



- ❑ **G3 scanners employ a bank of detectors to cover the entire field of view of a wider beam.**
- ❑ **Each increment of angle produces an entire projection, eliminating the need to translate the source/detector pair, as in G1 and G2 scanners**

# Fourth Generation



- ❑ **Employing a circular ring of detectors (on the order of 5000 individual detectors), only the source must rotate.**
- ❑ **The key advantage of G3 and G4 scanners is speed;**
- ❑ **The key disadvantages are cost and greater X-ray scatter.**

*Thank  
you*

