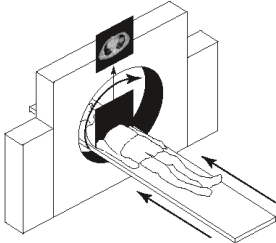



Bioengineering 280A  
 Principles of Biomedical Imaging  
  
 Fall Quarter 2010  
 CT Lecture 1

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## Computed Tomography

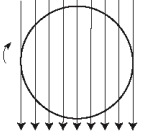



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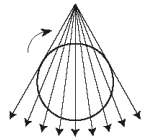
## Computed Tomography

Parallel Beam

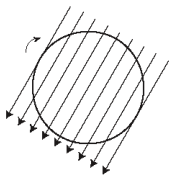


(a)

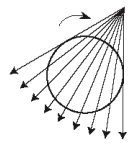
Fan Beam



(b)



(c)

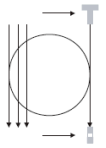


(d)

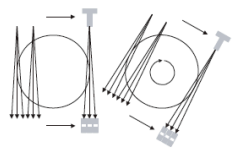
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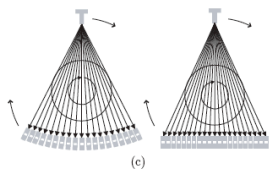
## Scanner Generations



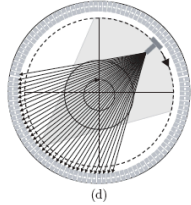
(a)



(b)



(c)

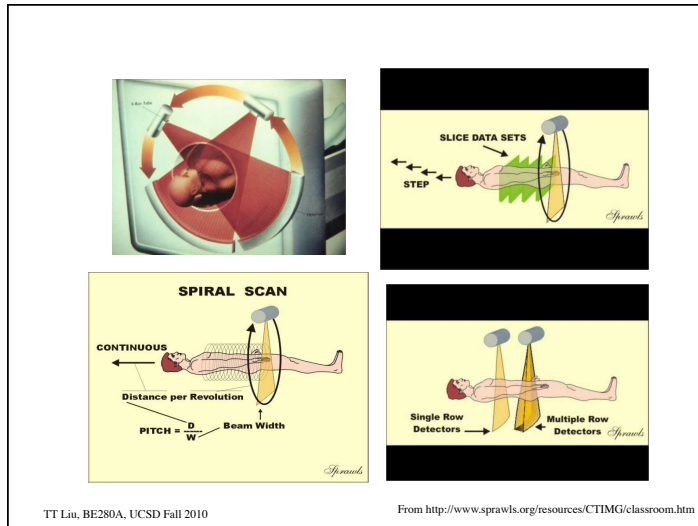


(d)

Figure 5.19: Subsequent scanner generations: (a) first generation, (b) second generation, (c) third generation and (d) fourth generation CT scanner.

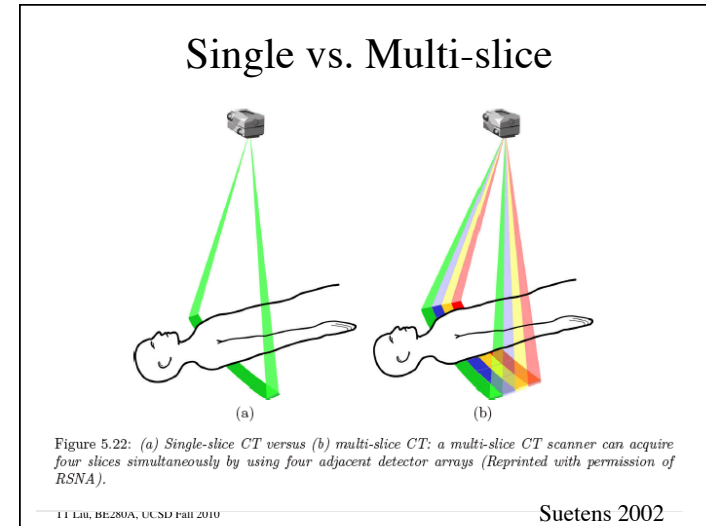
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From <http://www.sprawls.org/resources/CTIMG/classroom.htm>



### Scanner Generations

TABLE 6.1  
Comparison of CT Generations

| Generation         | Source                                    | Source Collimation                | Detector                     | Detector Collimation           | Source-Detector Movement               | Advantages  | Disadvantages   |
|--------------------|---|-----------------------------------|------------------------------|--------------------------------|--|---|---|
| 1G                 | Single x-ray tube                         | Pencil beam                       | Single                       | None                           | Move linearly and rotate in unison     | Scattered energy is undetected                                  | Slow  |
| 2G                 | Single x-ray tube                         | Fan beam, not enough to cover FOV | Multiple                     | Collimated to source direction | Move linearly and rotate in unison     | Faster than 1G  | Lower efficiency and larger noise because of the collimation in detectors |
| 3G                 | Single x-ray tube                         | Fan beam, enough to cover FOV     | Many                         | Collimated to source direction | Rotate in synchrony                    | Faster than 2G, continuous rotation using a slip ring.          | More expensive than 2G, low efficiency                                    |
| 4G                 | Single x-ray tube                         | Fan beam covers FOV               | Stationary ring of detectors | Cannot collimate detectors     | Detectors are fixed, source rotates    | Higher efficiency than 3G                                       | High scattering since detectors are not collimated                        |
| 5G (EBCT)          | Many tungsten anodes in single large tube | Fan beam                          | Stationary ring of detectors | Cannot collimate detectors     | No moving parts                        | Extremely fast, capable of stop-action imaging of beating heart | High cost, difficult to calibrate   |
| 6G (Spiral CT)     | 3G/4G                                     | 3G/4G                             | 3G/4G                        | 3G/4G                          | 3G/4G plus linear patient table motion | Fast 3D images  | A bit more expensive  |
| 7G (Multislice CT) | Single x-ray tube                         | Cone beam                         | Multiple arrays of detectors | Collimated to source direction | 3G/4G/6G motion                        | Fast 3D images  | Expensive   |

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### 1G vs. 2G scanner

Example 6.1 from Prince and Links.  
Compare 1G vs. 2G scanner whose source - detector apparatus can move linearly at speed of 1 m/sec; FOV 0.5m; 360 projections over 180 degrees; 0.5 s for apparatus to rotate one angular increment, regardless of angle.

Question : Scan time for 1 G scanner? Scan time for 2G scanner with 9 detectors space 0.5 degrees apart?

Answer :

1G scanner :  $0.5m/(1m/s) = 0.5s$  per projection.  
 $360 * 0.5 = 180s$  scan time  
 $360 * 0.5 = 180s$  for rotation of apparatus.  
 Total time = 360 s or 6 minutes.

2G scanner : Required angular resolution is  $180/360 = 0.5$  degrees -- agrees with spacing.  
 $360/9 = 40$  rotations required.  
 $40 * 0.5 = 20s$  for scanning  
 $40 * 0.5 = 20s$  for rotations.  
 Total time = 40s.

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## 3G, 6G, and 7G scanners

3G scanner: Typical scanner acquires 1000 projections with fanbeam angle of 30 to 60 degrees; 500 to 700 detectors; 1 to 20 seconds.

6G: Spiral/Helical CT

60 cm torso scan: 30s.

24 cm lung scan: 12s

15 cm angio: 30s

7G: Multislice CT

64 or more parallel 1D projections.

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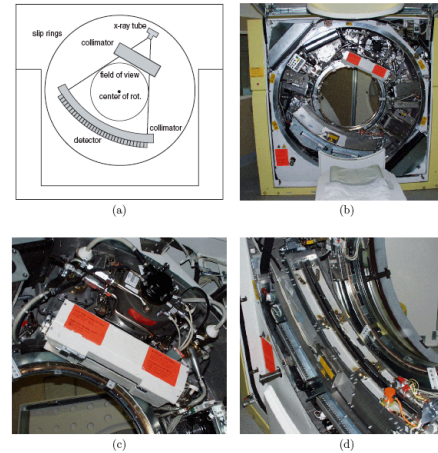


Figure 5.20: (a-b) The basic internal geometry of a third generation spiral CT scanner. (c) X-ray tube with adjustable collimating split. (d) Detector array with post-patient collimator.

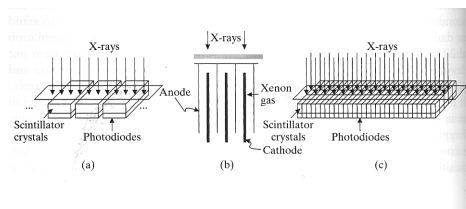
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## Detectors

Figure 6.7

(a) Solid-state detectors, (b) xenon gas detectors, and (c) multiple (solid-state) detector array.



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Prince and Links 2005

## CT Line Integral

$$I_d = \int_0^{E_{\max}} S_0(E) E \exp\left(-\int_0^d \mu(s; E) ds\right) dE$$

Monoenergetic Approximation

$$I_d = I_0 \exp\left(-\int_0^d \mu(s; \bar{E}) ds\right)$$

$$g_d = -\log\left(\frac{I_d}{I_0}\right) \\ = \int_0^d \mu(s; \bar{E}) ds$$

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## CT Number

$$\text{CT\_number} = \frac{\mu - \mu_{\text{water}}}{\mu_{\text{water}}} \times 1000$$

Measured in Hounsfield Units (HU)

Air: -1000 HU

Soft Tissue: -100 to 60 HU

Cortical Bones: 250 to 1000 HU

Metal and Contrast Agents: > 2000 HU

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## CT Display

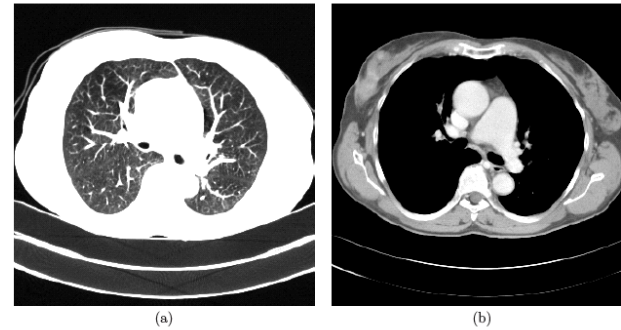


Figure 5.4: CT-image of the chest with different window/level settings:(a) for the lungs (window 1500 and level -500) and (b) for the soft tissues (window 350 and level 50).

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## Direct Inverse Approach

|         |         |
|---------|---------|
| $\mu_1$ | $\mu_2$ |
| $\mu_3$ | $\mu_4$ |

$$\begin{array}{l}
 p_1 \\
 p_2 \\
 p_3 \\
 p_4
 \end{array}
 =
 \begin{array}{l}
 \mu_1 + \mu_2 \\
 \mu_3 + \mu_4 \\
 \mu_1 + \mu_3 \\
 \mu_2 + \mu_4
 \end{array}
 \begin{array}{l}
 \begin{bmatrix} p_1 \\ p_2 \\ p_3 \\ p_4 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \end{bmatrix} \\
 \mu_1 \\
 \mu_2 \\
 \mu_3 \\
 \mu_4
 \end{array}$$

4 equations, 4 unknowns.

Are these the correct equations to use?

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## Direct Inverse Approach

|         |         |
|---------|---------|
| $\mu_1$ | $\mu_2$ |
| $\mu_3$ | $\mu_4$ |

$$\begin{array}{l}
 p_1 \\
 p_2 \\
 p_3 \\
 p_4
 \end{array}
 =
 \begin{array}{l}
 \mu_1 + \mu_2 \\
 \mu_3 + \mu_4 \\
 \mu_1 + \mu_3 \\
 \mu_2 + \mu_4
 \end{array}
 \begin{array}{l}
 \begin{bmatrix} p_1 \\ p_2 \\ p_3 \\ p_4 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \end{bmatrix} \\
 \mu_1 \\
 \mu_2 \\
 \mu_3 \\
 \mu_4
 \end{array}$$

4 equations, 4 unknowns.

Are these the correct equations to use?

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## Direct Inverse Approach

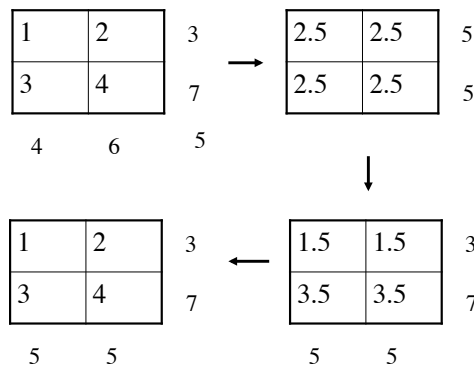
|         |         |
|---------|---------|
| $\mu_1$ | $\mu_2$ |
| $\mu_3$ | $\mu_4$ |

$$\begin{matrix} p_1 & p_1 = \mu_1 + \mu_2 \\ p_2 & p_2 = \mu_3 + \mu_4 \\ p_3 & p_3 = \mu_1 + \mu_3 \\ p_5 & p_5 = \mu_1 + \mu_4 \end{matrix} \quad \begin{matrix} p_1 \\ p_2 \\ p_3 \\ p_5 \end{matrix} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \end{bmatrix}$$

4 equations, 4 unknowns. These are linearly independent now.  
 In general for a  $N \times N$  image,  $N^2$  unknowns,  $N^2$  equations.  
 This requires the inversion of a  $N^2 \times N^2$  matrix  
 For a high-resolution  $512 \times 512$  image,  $N^2 = 262144$  equations.  
 Requires inversion of a  $262144 \times 262144$  matrix!  
 Inversion process sensitive to measurement errors.

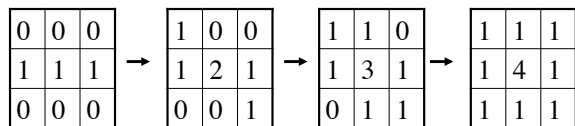
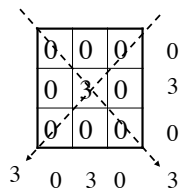
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## Iterative Inverse Approach Algebraic Reconstruction Technique (ART)



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## Backprojection



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## In-Class Exercise

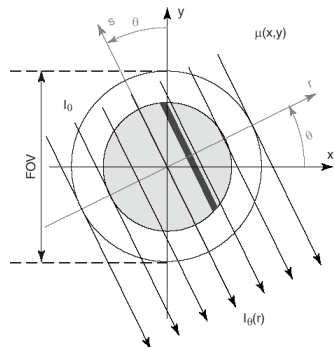
|         |         |
|---------|---------|
| $\mu_1$ | $\mu_2$ |
| $\mu_3$ | $\mu_4$ |

$$\begin{matrix} 5.7 \\ 11.3 \\ 8.2 & 8.8 & 10.1 \end{matrix}$$

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## Projections



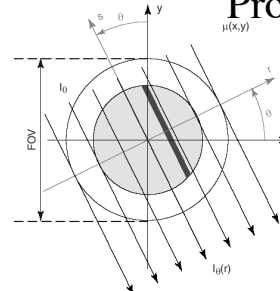
$$\begin{bmatrix} r \\ s \end{bmatrix} = \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} r \\ s \end{bmatrix}$$

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## Projections

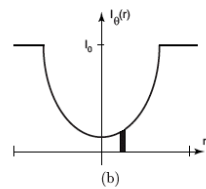


$$\begin{aligned} I(r, \theta) &= I_0 \exp\left(-\int_{L_r, \theta} \mu(x, y) ds\right) \\ &= I_0 \exp\left(-\int_{L_r, \theta} \mu(r \cos\theta - s \sin\theta, r \sin\theta + s \cos\theta) ds\right) \end{aligned}$$

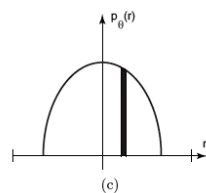
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## Projections



$$\begin{aligned} I(r, \theta) &= I_0 \exp\left(-\int_{L_r, \theta} \mu(r \cos\theta - s \sin\theta, r \sin\theta + s \cos\theta) ds\right) \\ p(r, \theta) &= -\ln \frac{I_\theta(r)}{I_0} \\ &= \int_{L_r, \theta} \mu(r \cos\theta - s \sin\theta, r \sin\theta + s \cos\theta) ds \end{aligned}$$

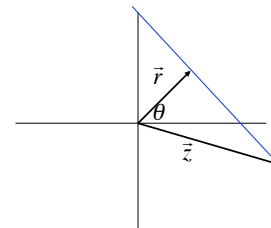


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## Radon Transform

$$\begin{aligned} g(r, \theta) &= \int_{-\infty}^{\infty} \mu(x(s), y(s)) ds \\ &= \int_{-\infty}^{\infty} \mu(r \cos\theta - s \sin\theta, r \sin\theta + s \cos\theta) ds \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mu(x, y) \delta(x \cos\theta + y \sin\theta - r) dx dy \end{aligned}$$



$$\begin{aligned} \vec{z} \cdot \frac{\vec{r}}{r} &= r \\ (x\hat{x} + y\hat{y}) \cdot (\cos\theta\hat{x} + \sin\theta\hat{y}) &= r \\ x \cos\theta + y \sin\theta &= r \end{aligned}$$

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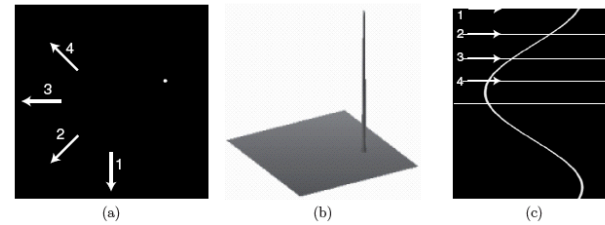
## Example

$$f(x,y) = \begin{cases} 1 & x^2 + y^2 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} g(l, \theta = 0) &= \int_{-\infty}^{\infty} f(l,y) dy \\ &= \int_{-\sqrt{1-l^2}}^{\sqrt{1-l^2}} dy \\ &= \begin{cases} 2\sqrt{1-l^2} & |l| \leq 1 \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

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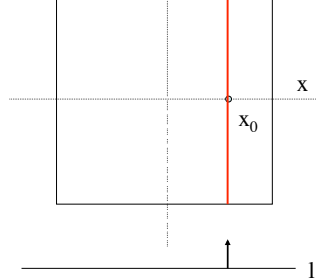
## Sinogram



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## Backprojection



$$\begin{aligned} b(x_0, y) &= p(l, \theta = 0) \Delta\theta \\ &= p(x_0, 0) \Delta\theta \end{aligned}$$

$$b_\theta(x, y) = g(x \cos \theta + y \sin \theta, \theta) \Delta\theta$$

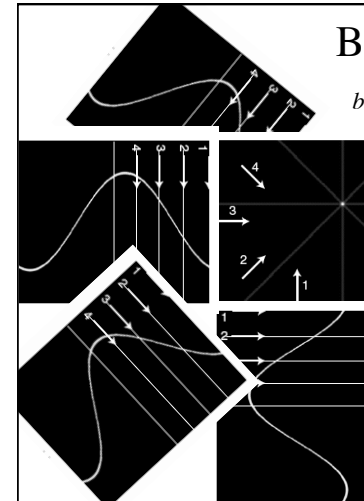
$$b(x, y) = B\{g(l, \theta)\}$$

$$= \int_0^\pi g(x \cos \theta + y \sin \theta, \theta) d\theta$$

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## Backprojection



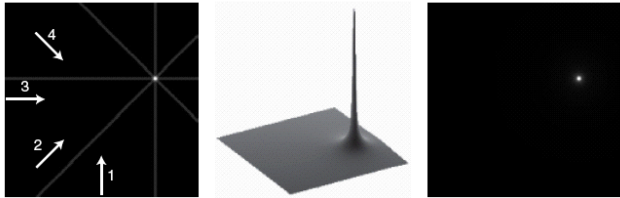
$$b(x, y) = B\{p(l, \theta)\}$$

$$= \int_0^\pi p(x \cos \theta + y \sin \theta, \theta) d\theta$$

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## Backprojection

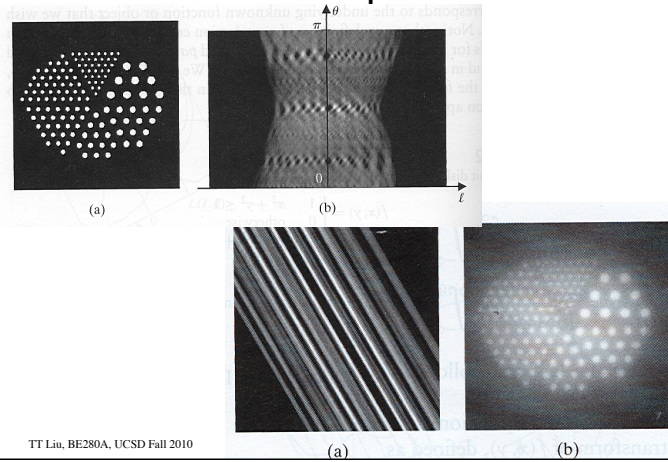


$$b(x, y) = B\{p(l, \theta)\} = \int_0^\pi p(x \cos \theta + y \sin \theta, \theta) d\theta$$

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## Example



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