#### 1-D Sampling



- Let  $f_s = 1/T$  be the sampling frequency.
- What is the relationship between  $S(e^{j\omega})$  and G(f)?

$$S(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} G\left(\frac{\omega - 2\pi k}{2\pi T}\right)$$

- Intuition
  - Scale frequencies

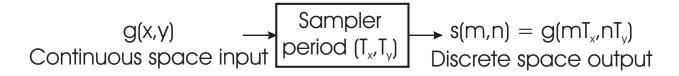
$$f = 0 \Leftrightarrow \omega = 0$$

$$f = \frac{1}{2T} = \frac{1}{2}f_s \Leftrightarrow \omega = \pi$$

$$f = \frac{1}{T} = f_s \Leftrightarrow \omega = 2\pi$$

- Replicate at period  $2\pi$
- Apply gain factor of  $\frac{1}{T}$ .

#### 2-D Sampling



- Let  $T_x$  and  $T_y$  be the sampling period in the x and y dimensions.
- Then

$$S(e^{j\mu}, e^{j\nu}) = \frac{1}{T_x T_y} \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} G\left(\frac{\mu - 2\pi k}{2\pi T_x}, \frac{\nu - 2\pi l}{2\pi T_y}\right)$$

- Intuition
  - Scale frequencies

$$(u, v) = (0, 0) \Leftrightarrow (\mu, \nu) = (0, 0)$$

$$(u, v) = (\frac{1}{2T_r}, 0) \iff (\mu, \nu) = (\pi, 0)$$

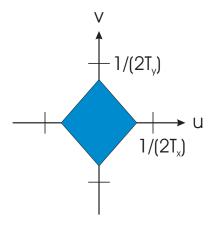
$$(u, v) = (0, \frac{1}{2T_y}) \Leftrightarrow (\mu, \nu) = (0, \pi)$$

$$(u,v) = (\frac{1}{2T_x}, \frac{1}{2T_y}) \Leftrightarrow (\mu, \nu) = (\pi, \pi)$$

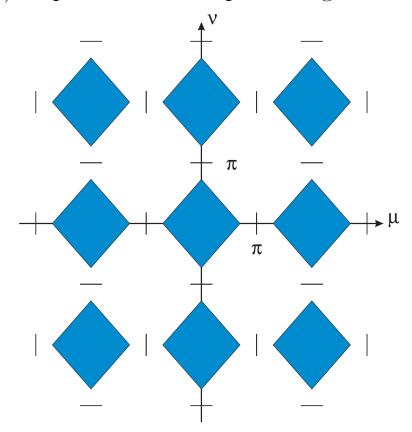
- Replicate along both  $\mu$  and  $\nu$  with period  $2\pi$
- Apply gain factor of  $\frac{1}{T_x T_y}$ .

## Example 1: 2-D Sampling Without Aliasing

G(u,v) - Spectrum of continuous space image.

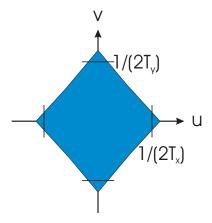


 $S(e^{j\mu},e^{j\nu})$  - Spectrum of sampled image.

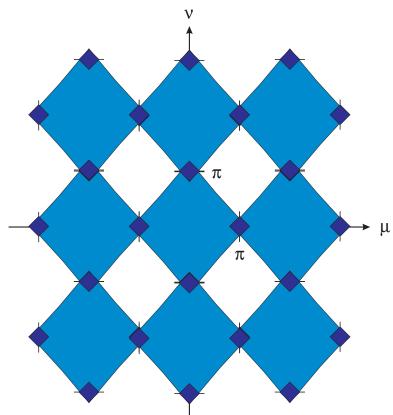


# Example 2: 2-D Sampling With Aliasing

G(u, v) - Spectrum of continuous space image.



 $S(e^{j\mu},e^{j\nu})$  - Spectrum of sampled image.



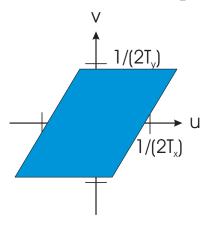
#### **Nyquist Condition**

• A continuous-space signal, g(x,y), may be uniquely reconstructed from its sampled version, s(m,n), if G(u,v)=0 for all  $|u|>\frac{1}{2T_x}$  and  $|v|>\frac{1}{2T_y}$ .

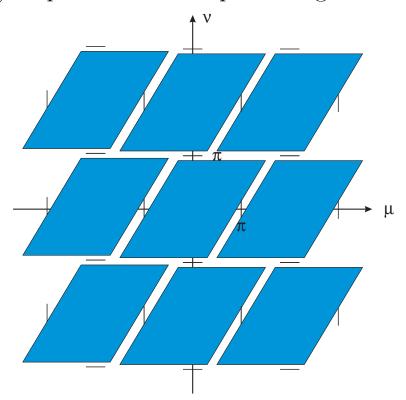
• This condition is sufficient, but not necessary.

### Example 3: Nonrectangular Spectral Support

G(u, v) - Spectrum of continuous space image.

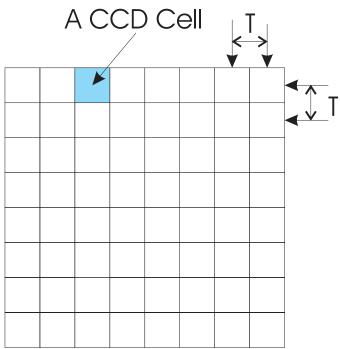


 $S(e^{j\mu},e^{j\nu})$  - Spectrum of sampled image.



#### Focal Plain Arrays

• Typical Charged Coupled Devices (CCD) Imaging array



- Solid state device used in video and still cameras.
- Each cell collects photons in a square  $T \times T$  region.
- Response of each cell is linear with energy (photons).
- Signal is "shifted out" after captured.
- Cell should be large for best sensitivity.
- Finite cell size violates sampling assumptions.

#### Mathematical Model for CCD

• Let s(m,n) be the output of cell (m,n), then  $s(m,n) = \int_{\mathbb{R}^2} h(x-mT,y-nT) \, g(x,y) \, dx dy$  where h(x,y) is the rectangular window for each cell.

$$h(x,y) = \frac{1}{T^2} rect(x/T, y/T)$$

• Define  $\tilde{g}(x,y)$  so that

$$\begin{split} \tilde{g}(\xi,\eta) &= \int_{\mathbb{R}^2} h(x-\xi,y-\eta) \, g(x,y) \, dx dy \\ &= h(-x,-y) * g(x,y) \end{split}$$

and then we have that

$$s(m,n) = \tilde{g}(mT, nT)$$

#### CCD Model in Space Domain

• Filter signal with space reversed cell profile

$$\begin{split} \tilde{g}(x,y) &= h(-x,-y) * g(x,y) \\ &= \frac{1}{T^2} \mathrm{rect}(x/T,y/T) * g(x,y) \end{split}$$

• Sample filtered image

$$s(m,n) = \tilde{g}(mT, nT)$$

• Cell aperture blurs image.

#### CCD Model in Frequency Domain

• Filter signal with cell profile

$$\tilde{G}(u, v) = H^*(u, v)G(u, v)$$

$$= \operatorname{sinc}(uT, vT)G(u, v)$$

• Sample filtered image

$$S(e^{j\mu}, e^{j\nu}) = \frac{1}{T^2} \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \tilde{G}\left(\frac{\mu - 2\pi k}{2\pi T}, \frac{\nu - 2\pi l}{2\pi T}\right)$$

• Complete model

$$S(e^{j\mu}, e^{j\nu}) = \frac{1}{T^2} \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \operatorname{sinc}\left(\frac{\mu - 2\pi k}{2\pi}, \frac{\nu - 2\pi l}{2\pi}\right) \cdot G\left(\frac{\mu - 2\pi k}{2\pi T}, \frac{\nu - 2\pi l}{2\pi T}\right)$$

• Sinc function filters image.

## Sampled Image Display or Rendering (Reconstruction)

- CRT's and LCD displays convert discrete-space images to continuous-space images.
- Notation:

$$s(m,n)$$
 - sampled image  $p(x,y)$  - point spread function (PSF) of display  $f(x,y)$  - displayed image

- Model:
  - In space domain:

$$f(x,y) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} s(m,n)p(x-mT,y-nT)$$

- In frequency domain:

$$F(u,v) = P(u,v)S(e^{j2\pi Tu}, e^{j2\pi Tv})$$

$$\mu \to 2\pi Tu$$

$$\nu \to 2\pi Tv$$

• Monitor PSF further "softens" image.

### Model for Sampling and Reconstruction

• Combining models for sampling and reconstruction results in:

$$F(u,v) =$$

$$\frac{P(u,v)}{T^2} \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} H^* \left( u - \frac{K}{T}, v - \frac{l}{T} \right) G \left( u - \frac{K}{T}, v - \frac{l}{T} \right)$$

• When no aliasing occurs, this reduces to

$$F(u,v) = \frac{P(u,v)H^*(u,v)}{T^2}G(u,v)$$
$$= \frac{P(u,v)\operatorname{sinc}(uT,vT)}{T^2}G(u,v)$$

#### Effect of Sampling and Reconstruction

• The image is effectively filtered by the transfer function

$$\frac{1}{T^2}P(u,v)H^*(u,v) = \frac{1}{T^2}P(u,v)\operatorname{sinc}(uT,vT)$$

• Scanned image normally must be "sharpened" to remove the effect of softening produced in the scanning and display processes.

### Raster Scan Ordering

- Specific scan pattern for mapping 2-D images to 1-D.
- Order pixels from top to bottom and left to right.
- Example: Consider the discrete-space image f(m, n)

$$\begin{bmatrix}
f(0,0) & \cdots & f(M-1,0) \\
\vdots & \ddots & \vdots \\
f(0,N-1) & \cdots & f(M-1,N-1)
\end{bmatrix}$$

• Raster ordering produces a 1-D signal  $x_n$ 

$$\begin{bmatrix} x_0 & x_1 & \cdots & x_{M-1} \\ x_M & x_{M+1} & \cdots & x_{2*M-1} \\ \vdots & \vdots & \ddots & \vdots \\ x_{(N-1)*M} & x_{(N-1)*M+1} & \cdots & x_{N*M-1} \end{bmatrix}$$

#### Vector Representation of Images

- An image is **not a matrix**. (A Matrix specifies a linear function.)
- Vectorizing images
  - Often image must be converted to a vector (data).
  - Vector looks like

$$x = \begin{bmatrix} f(0,0) \\ \vdots \\ f(M-1,0) \\ \vdots \\ f(0,N-1) \\ \vdots \\ f(M-1,N-1) \end{bmatrix}$$

- Mapping from vector to image is  $f(m, n) = x_{n*M+m}$ .