

# Point Lattices in Sampling Theory

## Multidimensional Sampling Theory

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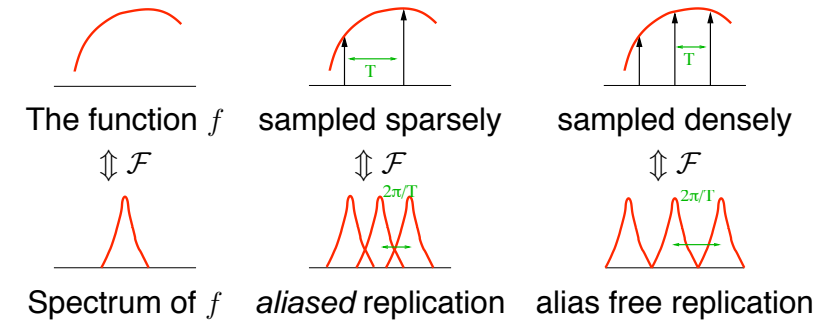
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# Sampling Theory in 1D

- Continuous-domain function
- Fourier domain representation



# Sampling in 1D

- Space Domain: Periodic sampling with period  $T$

$$f_s[k] = f(Tk)$$

- Fourier Domain: Periodic replication of spectra with period  $\frac{2\pi}{T}$

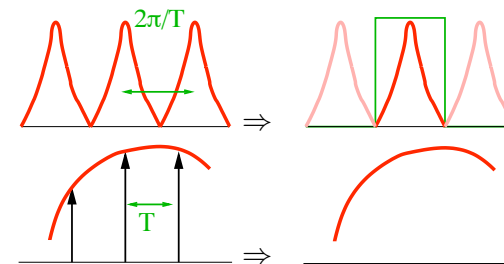
$$\hat{f}_s(\omega) = \sum_{k \in \mathbb{Z}} \hat{f}\left(\omega - \frac{2\pi k}{T}\right)$$

- Discrete samples correspond to periodic replication with a reciprocal period



# Reconstruction in 1D

- Best approximation in the space of bandlimited functions
- Space domain: Get a continuous function back from discrete samples
- Fourier domain: Remove the periodicity of the spectrum. Diminish the replica

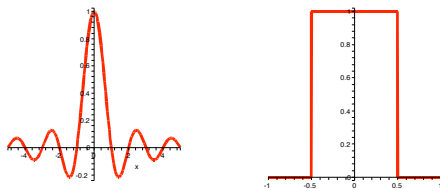


## Ideal reconstruction in 1D

- Best convolution kernel for reconstructing *into the space of bandlimited functions*

$$\text{sinc}(x) = \frac{\sin \pi x}{\pi x} \quad (1)$$

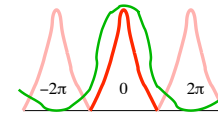
- Infinite support in space domain!
- Fourier domain: Box function



The sinc function    the Box function

## Compact support kernels

- Compact support means infinite support in Fourier domain
- Diminish the replicas by zeroing out the replicas located at aliasing frequencies:  $\dots, \frac{-4\pi}{T}, \frac{-2\pi}{T}, \frac{2\pi}{T}, \frac{4\pi}{T}, \dots$
- Strang-Fix Vanishing moments: The order of zeros at the aliasing frequencies determines the smoothness of reconstruction

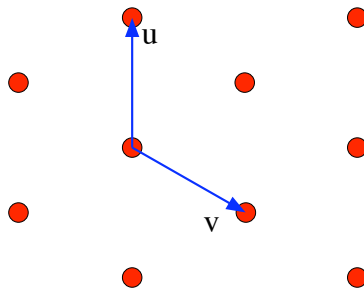


The zero crossings are important at *aliasing* frequencies

## Multidimensional Sampling Theory

- Space Domain: Periodic sampling with period *matrix*  $M = [u \ v]$

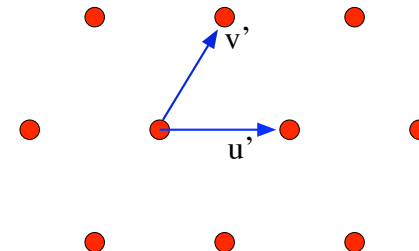
$$f_s[\mathbf{k}] = f(M\mathbf{k}), \mathbf{k} \in \mathbb{Z}^n$$



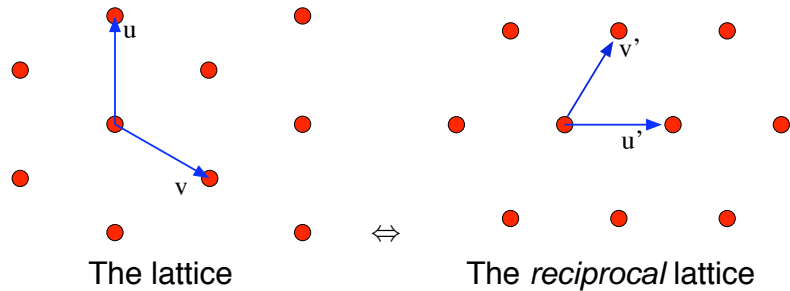
## Multidimensional Sampling Theory

- Fourier Domain: Periodic replication of spectra with the period matrix  $2\pi M^{-T} = [u' \ v']$ .

$$\hat{f}_s(\omega) = \sum_{\mathbf{k} \in \mathbb{Z}^n} \hat{f}(\omega - 2\pi M^{-T} \mathbf{k})$$



## Multidimensional Sampling Theory



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## Replication of Spectrum

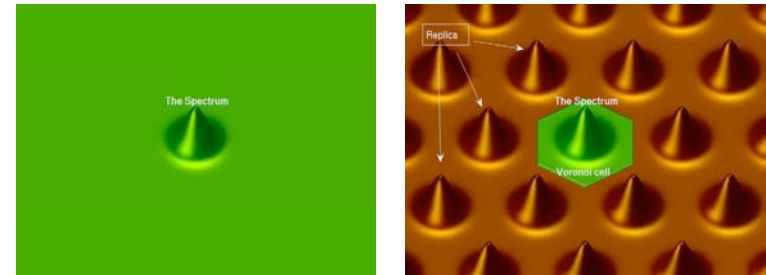


Figure 1: The Effect of sampling with  $M = [uv]$  in the Fourier domain is replication on  $2\pi M^{-T}$ , i.e. the *reciprocal lattice*

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## Reconstruction/Interpolation

- Reconstruction in the space of bandlimited functions
- Sampling results in a periodic replication in the Fourier domain
- Reconstruction = inverse sampling
- Fourier domain: Removing the replica

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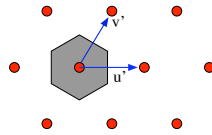
## Convolution based reconstruction

- Convolution based reconstruction = multiplication by a function that eliminates all but the main copy of the spectrum
- The replica are located on the *reciprocal, dual or polar* lattice.
- The Voronoi cell of the reciprocal lattice is where the primary copy spectrum is located (also known as Brillouin Zones).

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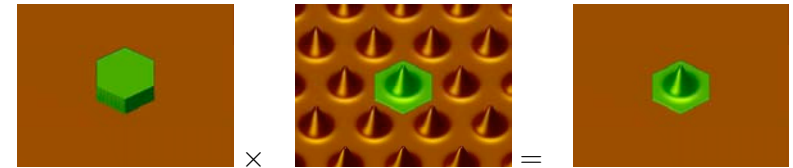
## The Ideal Bandlimited Reconstruction

- sinc function for the lattice  $M$  is defined to be the function whose Fourier transform is the characteristic function of the Voronoi cell of the lattice dual to  $M$ .
- The Fourier transform of the sinc function is constant on the interior of the Voronoi cell of the dual lattice and is zero everywhere else.
- Fourier transform of  $\text{sinc}_M$ : Characteristic function of the Voronoi cell of the reciprocal lattice



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## Ideal Bandlimited Reconstruction



- Fourier domain:  $\mathcal{F}\{\text{sinc}_M\}(\omega) \times \hat{f}_s(\omega) = \hat{f}_r(\omega)$
- Space domain:  $\text{sinc}_M(\mathbf{x}) * f_s(\mathbf{x}) = f_r(\mathbf{x})$

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## Other reconstruction kernels

- Reconstruction kernels with compact support need to have unbounded support in the Fourier domain
- One measure of accuracy of such kernels is by how much they diminish the influence of the replicas
- Strang-Fix result in the multidimensional case: the smoothness of reconstruction is determined by the number of vanishing moments.
- Vanishing moments = the order of zero roots at the centers of the replicas = at the dual lattice points

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## Compact support reconstruction kernels

- Tensor product of one dimensional compact support reconstruction kernels
- Radial based interpolation
- Multidimensional B-splines: successive convolutions of Characteristic function of the Voronoi cell of the sampling lattice, [See Hexsplines].
- Box splines: Projection of higher dimensional boxes
- Simplex and Polyhedral splines: Projection of higher dimensional simplices and polyhedra.

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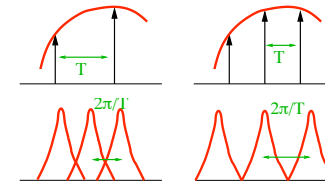
# Point Lattices in Sampling Theory

## *From Sphere Packing to the Optimal Sampling Lattice*

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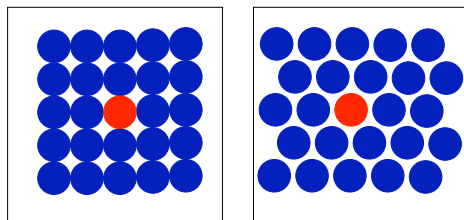
# The uncertainty principle

- The closer the samples are together, the farther replicas are from each other
- The farther the samples are from each other, the closer the replicas are together
- Optimal uniform sampling in 1D: bring the replicas of the spectrum as close to each other as possible



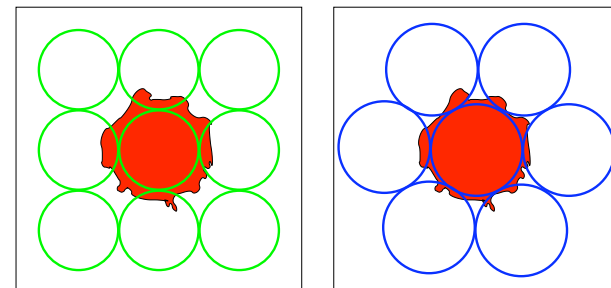
# Optimal lattice sampling

- Isotropic spectrum  $\Rightarrow$  uniform resolution on all orientations
- Spherical support of the Fourier representation
- Optimal lattice sampling in multidimensions: pack the replicas of the spectrum as tight as possible to each other
- Pack the spherical shape of the spectrum as dense as possible  $\Rightarrow$  Sphere packing!



# Optimal lattice sampling

- Same sampling density  $\Rightarrow$  Capture more information!
- In the Fourier domain cover larger part of the spectrum
- Hexagonal sampling captures higher frequencies with equal sampling density



## Kepler's problem

- What is the tightest arrangement of equal sized spheres in 3D?
- Motivated by dense packing of cannon balls on a battleship
- Grocer's method: Face Centered Cubic packing



Images Courtesy of Max Wardetzky <http://www.zib.de/wardetzky/kepler/index.html>

- Packing density:  $\frac{\text{volume of inscribing sphere}}{\text{volume of the Voronoi cell}}$



Optimal Sampling Lattice – p.5f

## Kepler Conjecture

- Can the packing density of FCC be bettered by another arrangement?
- Gauß proved that we can not if we want the arrangement still to be a lattice
- Harder to prove without the assumption of lattice packing
- Abstracted in the Hilbert's 18<sup>th</sup> problem
- Hales' found a computer aided proof in 1998.

The Face Centered Cubic packing:



Optimal Sampling Lattice – p.6f

## Sphere packing in 3D

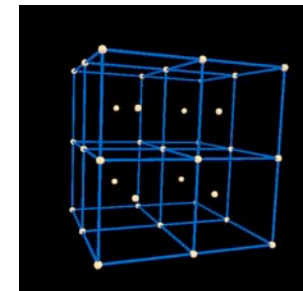
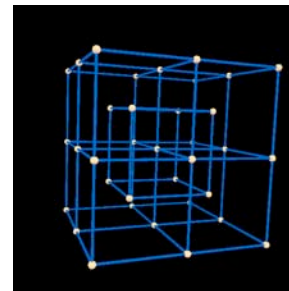
- The FCC packing is as dense as it gets!
- The study of point lattices and sphere packing are integral parts of each other
- The theory has fruited various other fields:
  - Quantization
  - Numerical integration
  - Design of error correcting codes, coding theory, Leech lattice
  - Sampling Theory [Covered in this course]



Optimal Sampling Lattice – p.7f

## Sphere packing in Sampling Theory

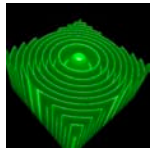
- Densest packing in Fourier domain
- Sparsest pattern for distribution of samples in space domain
- The Dual lattice!
- The Body Centered Cubic (BCC) lattice:



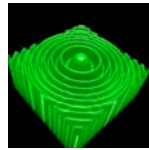
Optimal Sampling Lattice – p.8f

# Optimality of the BCC

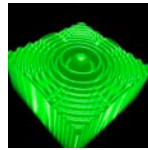
- Given the same number of samples with a Cartesian sampling, the BCC sampling captures 30% more information:



Original



Body Centered Cubic



Cartesian

- Information contained in the Cartesian sampled data can be represented with 70% of the samples, if sampled on the BCC lattice

