

Video Digitization

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Video Digitization

- 3D data types
- Video sampling
- Progressive video sampling
- General video sampling grids
- General analog video reconstruction





3D data types: video



3D data types: video



- Moving images are spatiotemporal 3D signals of the form: $f(x, y, t): \mathbb{R}^3 \to \mathbb{R}$, having:
 - domain \mathbb{R}^3 and codomain \mathbb{R} .
 - the time *t* coordinate has a different nature than the spatial coordinates *x*, *y*.
- Video scanning: creation of an 1D analog video signal, by sampling the time-varying images (luminance or RGB channels) along the vertical axis *y* and time *t*.





3D data types: video



Analog video scanning.



3D data types: video, cinema

- Analog video signal $f(x, j\Delta y, k\Delta t)$: $\mathbb{R} \times \mathbb{Z}^2 \to \mathbb{R}$.
 - discrete along *y* and *t* axes
 - continuous along *x* axis.
- Cinema moving images $f(x, y, k\Delta t)$: $\mathbb{R}^2 \times \mathbb{Z} \to \mathbb{R}$.

• Digital video signal $f(i\Delta x, j\Delta y, k\Delta t)$: $\mathbb{Z}^3 \to \mathbb{R}$.





3D data types: video, cinema



- Spatial sampling intervals Δx , Δy define image resolution: the smaller they are, the smaller the pixel size is.
 - *HDTV image resolution* 1080p: 1080 × 1920 pixels.
- Temporal sampling interval Δt defines the video frame rate in frames per second (fps).
 - Typical fps: 25 (Europe), 30 (USA).



3D data types: volumetric images

- **3D** volumetric images: 3D signals of the form: $f(x, y, z): \mathbb{R}^3 \to \mathbb{R}.$
- Discrete versions (defined on a Euclidean grid \mathbb{Z}^3):

 $f(n_1, n_2, n_3): \mathbb{Z}^3 \to \mathbb{R}.$

- $x = n_1 \Delta x$, $y = n_2 \Delta y$, $z = n_3 \Delta z$.
- $\Delta x, \Delta y, \Delta z$: **spatial sampling intervals** define 3D image resolution.





3D data types : volumetric images



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Image gallery of a 3D volume.



3D data types: multispectral images



- *Multispectral/multichannel* (*n*-channel) images have the form: f(x, y): $\mathbb{R}^2 \to \mathbb{R}^n$.
- Color images (n = 3):

• λ wavelength.

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- $\boldsymbol{f}(\boldsymbol{x},\boldsymbol{y}) = [f_R(\boldsymbol{x},\boldsymbol{y}),f_G(\boldsymbol{x},\boldsymbol{y}),f_B(\boldsymbol{x},\boldsymbol{y})]^T \colon \mathbb{R}^2 \to \mathbb{R}^3.$
- Digital color images (assigning 8 bits per color channel to each pixel): $f(n_1, n_2)$: $\mathbb{Z}^2 \rightarrow \{0, ..., 255\}^3$.
- They are also 3D images $f(n_1, n_2, i)$, i = 1,2,3.
- Hyperspectral images (3D images): $f(x, y, \lambda)$: $\mathbb{R}^3 \to \mathbb{R}$.

3D data types: memory issues



- 3D volumetric images have relatively small number of 3D image slices.
- Video has very large number of video frames:
 - 1 hour video: $60 \times 60 \times 50 = 180,000$ video frames.
- High frame rates: 25/50 up to 1000 Hz.
 - Use in recording high-speed phenomena.
- Video buffering:
 - Only few video frames are stored in RAM.
- Buffer update in live streams.

3D data types



- *Multiview images*: images of an object or set, taken from different view points, typically using different cameras.
 - Stereo images: a special case, employing only two cameras (left and right).
 - Video synchronization issues.



3D data types



- They both carry only implicit geometrical information about the visualized 3D object.
 - They are *not* 3D data.
 - 3D object geometry can be derived using stereo or multiview 3D geometry reconstruction techniques.





3D data types



Multiview video captured by synchronized video-cameras.



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Video sampling

- Analog video signal is an image sequence f(x, y, t).
 - x: horizontal coordinate,
 - *y*: vertical coordinate,
 - *t*: time coordinate.
- Digital video signal is obtained
 by spatiotemporal sampling of analog
 video along its coordinates *x*, *y*, *t*.



Spatiotemporal video signal.





Video sampling

Analog video scanning:

- 3D luminance signal is scanned along the horizontal axis *x* and then over time *t*.
- During analog video transmission, 1D signals $f(x, j\Delta y, \kappa\Delta t)$ are concatenated to form an 1D analog signal as a function of time that can be broadcasted.
- The analog video signal is discrete along axes *y*,*t* and continuous along axis *x*.





Video sampling

• Digital video can be obtained:

CCD chip.

- by sampling analog video along the horizontal scan lines,
- by using the discrete 2D sampling grid inherent in photoelectric sensors, e.g., in CCD chips.







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2D image sampling



Rectangular image sampling grid.



2D sampled image spectrum

 Ω_2

 (α)

 Ω_1





Fourier Transform of: a) Analog image; b) Discrete image.



Progressive video sampling



Progressive sampling:

• uniform spatiotemporal analog video sampling along x, y, t.



Sampling grids for: a) Progressive; b) 2:1 interlaced video.





Discrete (sampled) video $f(n_1, n_2, n_t)$:

 $f(n_1, n_2, n_t) = f_a(n_1 \Delta x, n_2 \Delta y, n_t \Delta t).$

f_a(x, y, t): analog spatiotemporal video signal.
Δ*x*, Δ*y*, Δ*t*: sampling intervals along the horizontal/vertical spatial axis *x*, *y* and time axis *t*.



Progressive video sampling



Forward and inverse **3D** Fourier transform of analog video:

 $F_{a}(\Omega_{x},\Omega_{y},\Omega_{t}) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f_{a}(x,y,t) e^{-i\Omega_{x}x - i\Omega_{y}y - i\Omega_{t}t} dxdydt,$

$$f_{a}(x,y,t) = \frac{1}{(2\pi)^{3}} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} F_{a}(\Omega_{x},\Omega_{y},\Omega_{t}) e^{i\Omega_{x}x + i\Omega_{y}y + i\Omega_{t}t} d\Omega_{x} d\Omega_{y} d\Omega_{t}$$

• $\Omega_x = 2\pi F_x$, $\Omega_y = 2\pi F_y$, $\Omega_t = 2\pi F_t$: spatiotemporal frequencies.

They describe video content variations over space/time.





- Smooth video content is represented by low frequency content:
 - **DC term**: $\boldsymbol{\Omega}^T = [\Omega_x, \Omega_y, \Omega_t]^T = [0, 0, 0]^T$ dominates.
- Image details and video content changes along x, y axis lead to high spatial frequency Ω_x , Ω_y content:
 - Edges, lines, textured regions.
- Fast video content changes over time lead to high Ω_t frequency content:

Artificial Infelience and the case of fast moving objects.



$$F(\omega_{x}, \omega_{y}, \omega_{t}) = \sum_{n_{1}} \sum_{n_{2}} \sum_{n_{3}} f(n_{1}, n_{2}, n_{t}) e^{-i(\omega_{x}n_{1} + \omega_{y}n_{2} + \omega_{t}n_{t})}$$
(4)

and:

$$f(n_1, n_2, n_3) = \left(\frac{1}{2\pi}\right)^3 \int_{-\pi}^{\pi} \int_{-\pi}^{\pi} \int_{-\pi}^{\pi} F(\omega_x, \omega_y, \omega_t) e^{i(\omega_x n_1 + \omega_y n_2 + \omega_t n_t)} d\omega_x d\omega_y d\omega_t$$
(5)

- $\omega_x = \Omega_x \Delta x$, $\omega_y = \Omega_y \Delta y$, $\omega_t = \Omega_t \Delta t$.
- they are defined on a unit circle: ω_x , ω_y , $\omega_t \in [-\pi, \pi]$.



• The triple integral can divided into a triple sum of integrals:

 $\begin{aligned} f(n_1, n_2, n_t) &= \\ \frac{1}{(2\pi)^3} \sum_{k_x} \sum_{k_y} \sum_{k_t} \int \int \int_{RP(k_x, k_y, k_t)} \frac{1}{\Delta x \Delta y \Delta t} F_a(\frac{\omega_x}{\Delta x}, \frac{\omega_y}{\Delta y}, \frac{\omega_t}{\Delta t}) e^{i\omega_x n_1 + i\omega_y n_2 + i\omega_t n_t} d\omega_x d\omega_y d\omega_t. \end{aligned}$

• Each of them is defined on the shifted parallelepipeds $RP(k_x, k_y, k_t)$:

 $-\pi + 2\pi k_x \le \omega_x \le \pi + 2\pi k_x,$ $-\pi + 2\pi k_y \le \omega_y \le \pi + 2\pi k_y,$ $-\pi + 2\pi k_t \le \omega_t \le \pi + 2\pi k_t.$



 By interchanging summations with integrals and changing variables, the 3D discrete spatiotemporal Fourier transform is given by:

$$F(\omega_x, \omega_y, \omega_t) = \frac{1}{\Delta x \Delta y \Delta t} \sum_{k_x} \sum_{k_y} \sum_{k_t} F_a(\frac{\omega_x - 2\pi k_x}{\Delta x}, \frac{\omega_y - 2\pi k_y}{\Delta y}, \frac{\omega_t - 2\pi k_t}{\Delta t})$$

or

- $F(\Omega_x \Delta_x, \Omega_y \Delta_y, \Omega_t \Delta_t) = \frac{1}{\Delta x \Delta y \Delta t} \sum_{k_x} \sum_{k_y} \sum_{k_t} F_a(\Omega_x \frac{2\pi k_x}{\Delta x}, \Omega_y \frac{2\pi k_y}{\Delta y}, \Omega_t \frac{2\pi k_t}{\Delta x}).$
- Periodic translation of the 3D Fourier transform $F_a(\Omega_x, \Omega_y, \Omega_t)$ of the analog video $f_a(x, y, t)$.





3D periodic translation of the analog video spectrum.





If the sampling intervals Δx , Δy , Δt are sufficiently small, so that:

$$F(\Omega_{x}\Delta x, \Omega_{y}\Delta y, \Omega_{t}\Delta t) = 0, \quad |\Omega_{x}| \ge \frac{\pi}{\Delta x}, |\Omega_{y}| \ge \frac{\pi}{\Delta y}, |\Omega_{t}| \ge \frac{\pi}{\Delta t},$$

the discrete video spectrum is given by the periodic translation of the fundamental period:

$$F(\Omega_{x}\Delta x, \Omega_{y}\Delta y, \Omega_{t}\Delta t) = \frac{1}{\Delta x \Delta y \Delta t} F_{a}(\Omega_{x}, \Omega_{y}, \Omega_{t})$$
$$|\Omega_{x}| \leq \frac{\pi}{\Delta x}, |\Omega_{y}| \leq \frac{\pi}{\Delta y}, |\Omega_{t}| \leq \frac{\pi}{\Delta t}.$$



Reconstruction of analog video



Analog video can be reconstructed from the discrete video:

$$f_a(x, y, t) = \sum_{n_1} \sum_{n_2} \sum_{n_t} f(n_1, n_2, n_t) \frac{\sin\frac{\pi}{\Delta x}(x - n_1 \Delta x)}{\frac{\pi}{\Delta x}(x - n_1 \Delta x)} \frac{\sin\frac{\pi}{\Delta y}(y - n_2 \Delta y)}{\frac{\pi}{\Delta y}(y - n_2 \Delta y)} \frac{\sin\frac{\pi}{\Delta t}(t - n_t \Delta t)}{\frac{\pi}{\Delta t}(t - n_t \Delta t)}.$$

- Mathematical modeling of digital video visualization/ projection.
- In reality, much simpler approaches are used:
 - Zero-order interpolation

• Square image blobs can be visible.

Reconstruction of analog video



- If $\Delta x, \Delta y, \Delta t$ are sufficiently small, analog video can be precisely reconstructed.
- If the video signal is not low-pass:
 - too many image details and/or very strong motion.
- or the sampling intervals Δx , Δy , Δt are not sufficiently small, **spectrum aliasing** occurs.



Reconstruction of analog video



- Spectrum content overlaps in high frequency areas: $|\Omega_x \Delta x| \le \pi, \quad |\Omega_y \Delta y| \le \pi, \quad |\Omega_t \Delta x| \le \pi.$
- High frequency video content is smoothed out (details and strong motion).



Reconstruction of analog video

• Nyquist criterion:

$$\Delta x \leq \frac{\pi}{\Omega_{xmax}}, \quad \Delta y \leq \frac{\pi}{\Omega_{ymax}}, \quad \Delta t \leq \frac{\pi}{\Omega_{tmax}},$$

to avoid aliasing problems.

- Sampling frequencies $F_{sx} = \frac{1}{\Lambda x}$, $F_{sy} = \frac{1}{\Lambda y}$, $F_{st} = \frac{1}{\Lambda t}$ must satisfy:
- $F_{sx} \ge 2F_{xmax}, \quad F_{sy} \ge 2F_{ymax}, \quad F_{st} \ge 2F_{tmax}.$ • $F_{xmax}, F_{ymax}, F_{tmax}$: maximal spatial and temporal video frequencies. Artificial Intelligence & 34 nformation Analysis Lab





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- Multidimensional signal $f(\mathbf{x})$ over \mathbb{R}^{K} .
- Video: special multidimensional signal for K = 3.
- **3D** sampling grid: mathematical structure describing video sampling (also called *lattice*):

$$\Lambda = \{ \mathbf{x} \in \mathbb{R}^K \mid \mathbf{x} = \sum_{k=1}^K n_k \mathbf{v}_k, \forall n_k \in \mathbb{Z} \}.$$

- $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_K$: Linearly independent basis vectors on \mathbb{R}^K .
- $\mathbf{V} = [\mathbf{v}_1, \dots, \mathbf{v}_K]$: sampling matrix.





a) Rectangular sampling grid; b) Vertically aligned interlaced grid; c) Quincunx grid; d) Orthorhombic grid.

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• Any grid point can be represented by vector:

$$\mathbf{n} = [n_1, n_2, \dots, n_K] \in \mathbb{Z}^K.$$

- Actual pixel position in \mathbb{R}^K : $\mathbf{x} = \mathbf{V}\mathbf{n}$.
- Sampling matrix or basis vectors fully define a sampling grid.
- They are not unique:
- V and V' = EV define the same sampling grid Λ , if:
- E: integer matrix and $det(E) = \pm 1$.





- Progressive sampling grid matrix: $\mathbf{V} = \begin{bmatrix} \Delta x & 0 & 0 \\ 0 & \Delta y & 0 \\ 0 & 0 & \Delta t \end{bmatrix}.$
- Vertically aligned interlaced grid matrix : $\mathbf{V} = \begin{bmatrix} \Delta x & 0 & 0 \\ 0 & 2\Delta y & \Delta y \\ 0 & 0 & \Delta t/2 \end{bmatrix}.$





V

• Quincunx grid matrix:

$$\mathbf{V} = \begin{bmatrix} \Delta x & 0 & \Delta x/2 \\ 0 & 2\Delta y & \Delta y \\ 0 & 0 & \Delta t/2 \end{bmatrix}.$$

• Orthorhombic grid matrix:

$$= \begin{bmatrix} 2\Delta x & \Delta x & \Delta x \\ 0 & \Delta y & 0 \\ 0 & 0 & \Delta t/2 \end{bmatrix}$$





- Each grid has a unit cell, whose definition is not unique.
- Unit grid cell (Voronoi cell) $\mathcal{V}(\Lambda)$: the set of points on \mathbb{R}^K that are closest to **0**, then to any other grid point:

 $\mathcal{V}(\Lambda) = \{ \mathbf{x} \in \mathbb{R}^K | d(\mathbf{x}, \mathbf{0}) \leq d(\mathbf{x}, \mathbf{p}), \quad \forall \mathbf{p} \in \Lambda \}.$

• $d(\mathbf{x}, \mathbf{p})$: Euclidean distance of points $\mathbf{x}, \mathbf{p} \in \mathbb{R}^{K}$.



- Grid sampling density: it is the number of grid nodes that exist in the unit volume in \mathbb{R}^{K} .
- It equals the inverse of the unit cell volume:







- Inverse grid Λ^* is defined by the grid sampling matrix U: $\mathbf{V}^T \mathbf{U} = 2\pi \mathbf{I}.$
- Sampling grid density of the inverse grid Λ*:

$$D(\Lambda^*) = \frac{1}{(2\pi)^3 D(\Lambda)} = \frac{|det\mathbf{V}|}{(2\pi)^3}.$$

• It is inversely proportional to that of the sampling lattice Λ .



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- $f_a(\mathbf{x}_t), \mathbf{x}_t = [x, y, t]^T \in \mathbb{R}^3$: analog video.
- Video sampling using sampling matrix V:

$$f(\mathbf{n}) = f_a(\mathbf{x}_t) \sum_{n \in \mathbb{Z}^3} \delta(\mathbf{x}_t - \mathbf{V}\mathbf{n}).$$

• $\mathbf{n} = [n_1, n_2, n_t]^T \in \mathbb{Z}^3$.





• Discrete spatiotemporal Fourier transform:

$$F(\boldsymbol{\omega}) = \sum_{\mathbf{n}\in\mathbb{Z}^3} f(\mathbf{n})e^{i\boldsymbol{\Omega}^T\mathbf{V}\mathbf{n}}$$
 ,

- $\boldsymbol{\omega} = \mathbf{V}^T \boldsymbol{\Omega}$.
- $\mathbf{U} = 2\pi (\mathbf{V}^T)^{-1}$: sampling matrix the inverse grid Λ^* .
- If $\mathbf{n}^T \mathbf{k} = m \in \mathbb{Z}$, then $F(\boldsymbol{\omega} + 2\pi \mathbf{k}) = F(\boldsymbol{\omega})$.
- The Fourier transform of a sample video on a lattice Λ is periodic on lattice Λ^* on \mathbb{R}^3 .



• Discrete video spectrum:

$$F(\boldsymbol{\omega}) = \frac{1}{|\det \mathbf{V}|} \sum_{\mathbf{k} \in \mathbb{Z}^3} F_a(\frac{1}{2\pi} \mathbf{U}(\boldsymbol{\omega} - 2\pi \mathbf{k})).$$

- Sampling grid Λ,
 - Periodic translation of the analog video spectrum on the inverse grid Λ^* .





Spectrum aliasing:

- The denser the sampling grid Λ is, the more sparse is the inverse Λ^* grid.
- If the support area of the continuous video signal spectrum is bigger than the unit cell *P* of the grid Λ*, the nearby areas of the translated spectrum version overlap.

• Alias reduction: use a denser sampling grid.



• Progressive video frequency periodicity grid Λ^* :

$$\mathbf{U} = 2\pi (\mathbf{V}^{T})^{-1} = \begin{bmatrix} \frac{2\pi}{\Delta x} & 0 & 0\\ 0 & \frac{2\pi}{\Delta y} & 0\\ 0 & 0 & \frac{2\pi}{\Delta t} \end{bmatrix},$$

• 2:1 interlaced frequency periodicity grid Λ^* :

$$\mathbf{J} = 2\pi (\mathbf{V}^T)^{-1} = \begin{bmatrix} \frac{2\pi}{\Delta x} & 0 & 0\\ 0 & \frac{\pi}{\Delta y} & 0\\ 0 & -\frac{2\pi}{\Delta t} & \frac{4\pi}{\Delta t} \end{bmatrix}.$$

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Frequency periodicity grids Λ^* :



Inverse sampling grid Λ* for: a) Progressive video; b) 2:1 interlaced video.



General analog video reconstruction



Analog video reconstruction:

- Original analog signal $f_a(\mathbf{x}_t)$ recovery from $f(\mathbf{n})$:
 - the support area of the Fourier transform $F_a(\Omega_x, \Omega_y, \Omega_t)$ should be a subset of the unit cell of the inverse grid Λ^* .
- Perfect reconstruction by an ideal low-pass filter:
 - Its passband should be the unit cell \mathcal{P} and the Λ^* grid:

 $F_a(\mathbf{\Omega}) = \begin{cases} |\det \mathbf{V}| F(\mathbf{V}^T \mathbf{\Omega}), & \mathbf{\Omega} \in \mathcal{P} \\ 0, & \text{elsewhere.} \end{cases}$



General analog video reconstruction

• Analog video reconstruction:

$$f_a(\mathbf{x}_t) = \sum_{\mathbf{n}\in\mathbb{Z}^3} f(\mathbf{n})h(\mathbf{x}_t - \mathbf{V}\mathbf{n}).$$

 Impulse response of the ideal spatiotemporal low-pass interpolation filter:

$$h(\mathbf{x}_t) = \frac{|\det \mathbf{V}|}{(2\pi)^3} \int_{\mathbf{P}} e^{i\mathbf{\Omega}^{\mathrm{T}}\mathbf{x}_{\mathrm{t}}} d\mathbf{\Omega}.$$





General analog video reconstruction



- For progressive video sampling:
 - the impulse response of the ideal video interpolation filter is the triple sinc function:

$$h(\mathbf{x}_t) = \frac{\sin(\frac{\pi}{\Delta x}x)}{\frac{\pi}{\Delta x}x} \frac{\sin(\frac{\pi}{\Delta y}y)}{\frac{\pi}{\Delta y}y} \frac{\sin(\frac{\pi}{\Delta t}t)}{\frac{\pi}{\Delta t}t}$$



Bibliography



[PIT2017] I. Pitas, "Digital video processing and analysis", China Machine Press, 2017 (in Chinese).

[PIT2013] I. Pitas, "Digital Video and Television", Createspace/Amazon, 2013.
[PIT2021] I. Pitas, "Computer vision", Createspace/Amazon, in press.
[NIK2000] N. Nikolaidis and I. Pitas, "3D Image Processing Algorithms", J. Wiley, 2000.
[PIT2000] I. Pitas, "Digital Image Processing Algorithms and Applications", J. Wiley, 2000.







Thank you very much for your attention!

More material in http://icarus.csd.auth.gr/cvml-web-lecture-series/

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