

Moving Image Perception

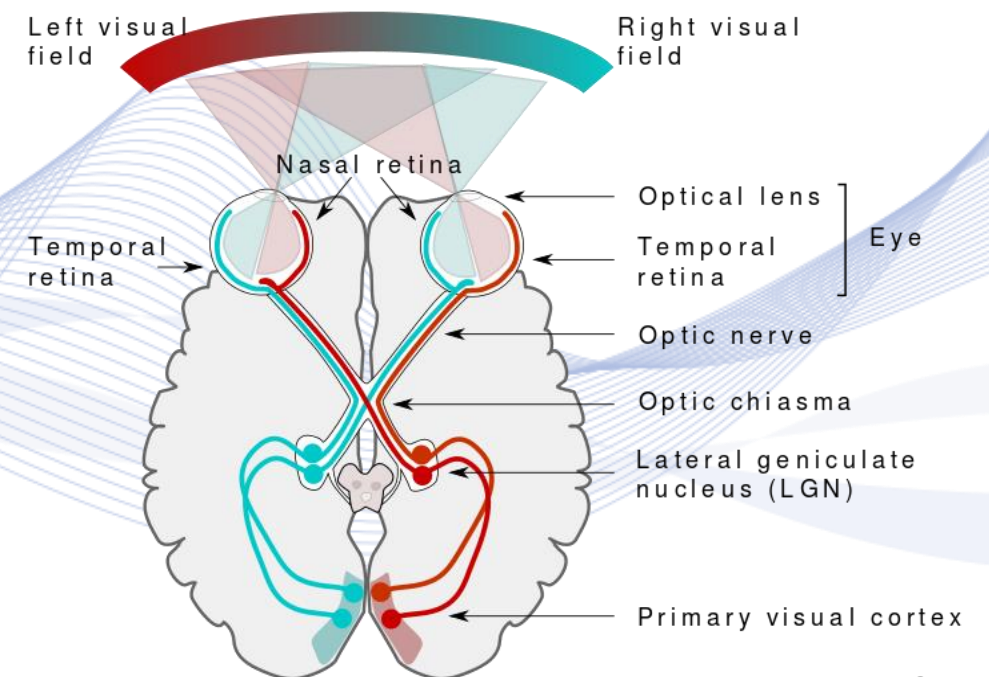
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Moving Image Perception

- **Human Vision Modeling**
- Video Frequency Content
- Spatiotemporal HVS Models
- Video Quality Assessment

Human Vision Modeling

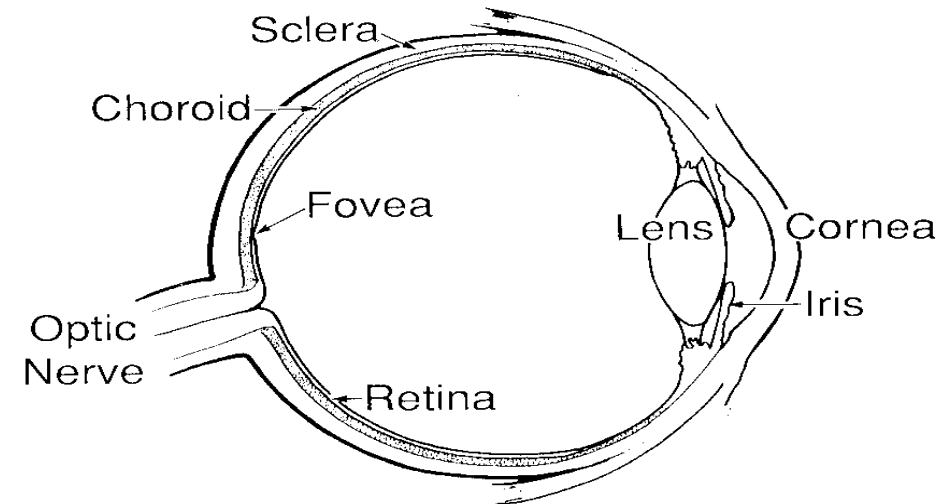
- One of digital image and video processing aims is image quality improvement.
- **Human Visual System (HVS)** modeling is difficult, because of its complex structure.



Human visual system [HVP].

Human Vision Modeling

- **Human eye:** spherical shape with a diameter of 20 mm.
- Light enters the **pupil** of the **iris** (diameter 2 - 8 mm).
- It passes through **lens**, **vitreous humour** and on the **retina**.

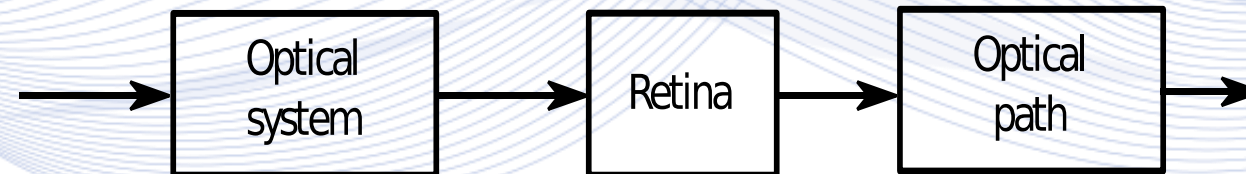


Human eye.

Human Vision Modeling

Retina light detectors: cones and rods.

- **Cones**: sensitive to color.
 - **Photopic** (high brightness, daylight) vision.
- **Rods**: sensitive to light intensity, not color.
 - They create a general idea of the visual content.
- **Scotopic** (night) vision.



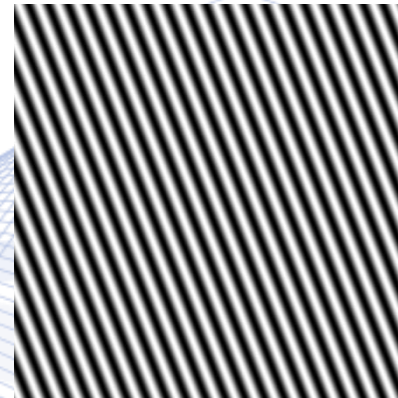
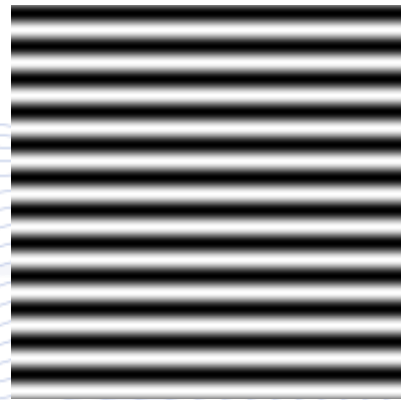
Human visual system model.

Moving Image Perception

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- **Video Frequency Content**
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Video Frequency Content

- A frequency F is linked with angular frequency $\Omega = 2\pi F$.
- Spatial frequencies (video content changes along x, y axes):
 - $\Omega_x = 2\pi F_x$ and $\Omega_y = 2\pi F_y$.



2D sinusoidal signals: a) $(F_x, F_y) = (0,6)$; b) $(F_x, F_y) = (10,4)$.

Video Frequency Content

Spatial frequencies F_x, F_y :

- They show spatial luminance changes on the image plane.
- Local frequency vector $\boldsymbol{\Omega} = [\Omega_x, \Omega_y]^T$ is colinear to local image content change direction (perpendicular to edge direction).
- Spatial frequencies can be defined along different orthogonal axes than (x, y) .
- They are measured in cycles per unit length:
 - e.g., a 2D sinusoidal spatial pattern $f(x, y) = \sin(20\pi y)$ has a frequency $(0, 10)$.

Video Frequency Content

Temporal Frequency F_t :

- Video signal: moving image (2D video frames changing over time).
- The temporal frequency F_t depends on image content motion.
- The video content motion is due to:
 - camera motion and/or
 - object(s) motion.

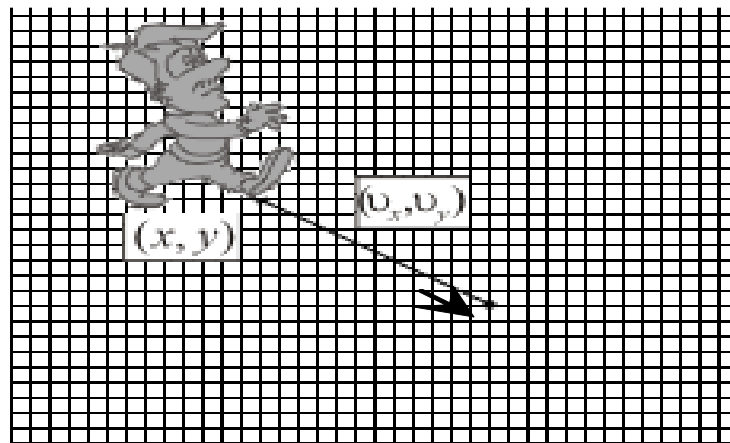
Video Frequency Content

Constant velocity **2D linear object motion**:

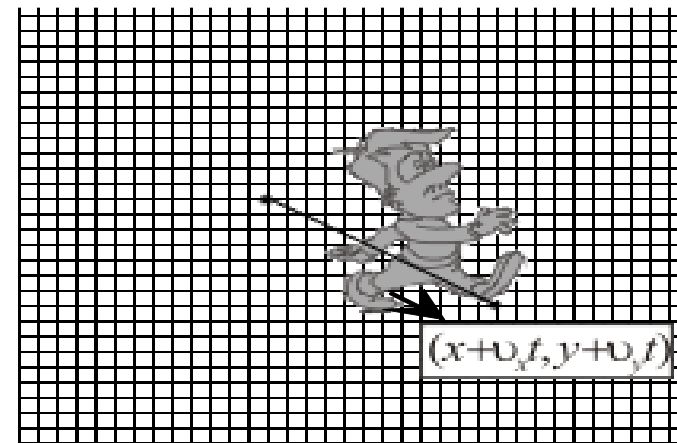
- $f_0(x, y) = f(x, y, 0)$: object image at time $t = 0$.
- $\mathbf{v} = [v_x, v_y]^T$: object motion vector.
- v_x, v_y : horizontal/vertical speed.
- Object image at time t (for homogeneous image background):

$$f(x, y, t) = f(x - v_x t, y - v_y t, 0) = f_0(x - v_x t, y - v_y t).$$

Video Frequency Content



$t=0$



$t>0$

Linear constant 2D object motion.

Video Frequency Content

- **Spatiotemporal Fourier Transform:**

$$F(\Omega_x, \Omega_y, \Omega_t) = \iiint f(x, y, t) e^{-i(\Omega_x x + \Omega_y y + \Omega_t t)} dx dy dt = F_0(\Omega_x, \Omega_y) \delta(\Omega_t + \Omega_x v_x + \Omega_y v_y).$$

δ : delta function.

- Non-zero spectrum $F(\Omega_x, \Omega_y, \Omega_t)$ only on the plane:

$$\Omega_t + \Omega_x v_x + \Omega_y v_y = 0.$$

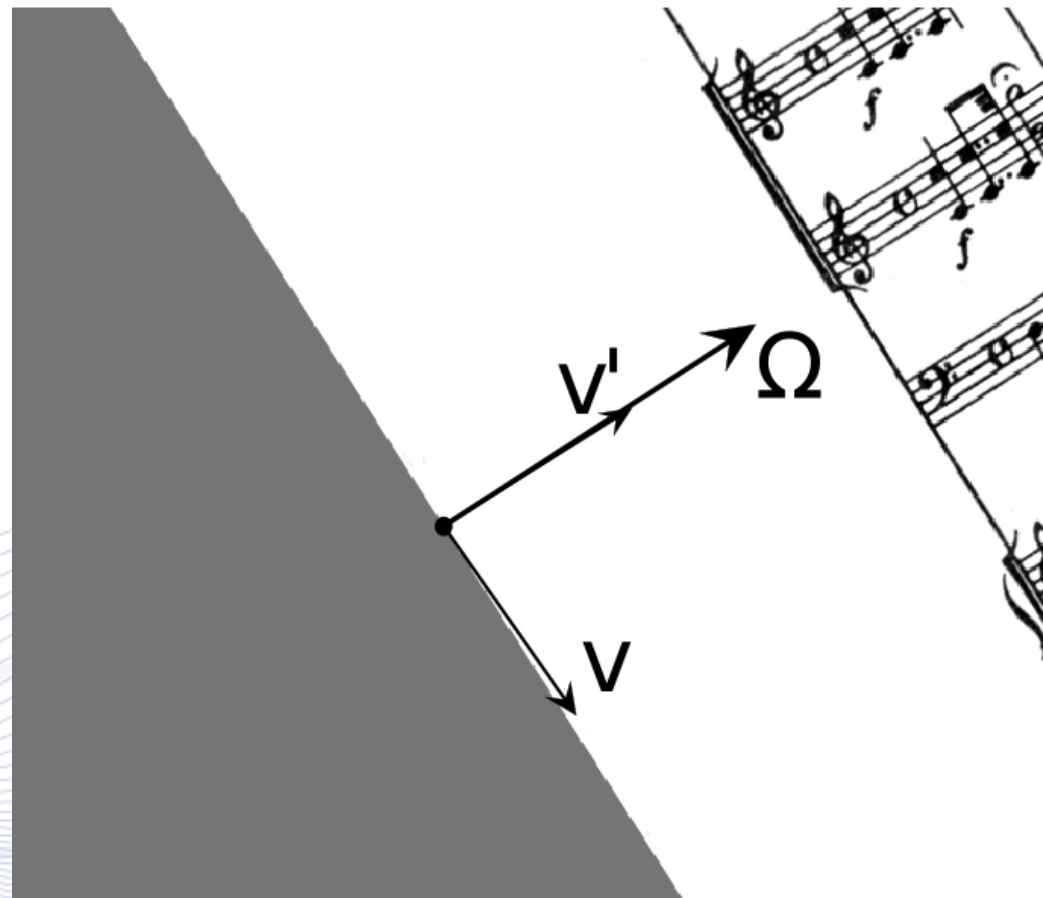
- Temporal frequency due to object motion:

$$\Omega_t = -\Omega_x v_x - \Omega_y v_y = -\mathbf{\Omega}^T \mathbf{v}.$$

Video Frequency Content

- When $\Omega_x = \Omega_y = 0$ (image is blank, has no content), then $\Omega_t = 0$, regardless of image plane motion.
- Temporal frequency $\Omega_t = 0$, if:
 - Motion vector \mathbf{v} is orthogonal to the spatial frequency vector $\mathbf{\Omega} = [\Omega_x, \Omega_y]^T$ or
 - $\mathbf{\Omega} = \mathbf{0}$ (no motion).

Video Frequency Content



Motion vector perpendicular to local image frequency vector.

Moving Image Perception

- Human Vision Modeling
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- **Spatiotemporal HVS Models**
- Video Quality Assessment

Spatiotemporal HVS models

- For the *spatiotemporal modelling* of human vision, dynamic models of neurons must be used:
 - It is particularly difficult.
- The eye is a dynamic system:
 - pupil diameter changes with light intensity,
 - the human eye can rotate and perform smooth pursuit movements.
- Spatiotemporal image perception experiments.

Spatiotemporal HVS models

Temporal sensitivity of the human vision:

- HVS temporal frequency response refers to HVS sensitivity to temporal video content variations.
- ***Display flicker.***

Spatiotemporal HVS models

Kelly experiments: determination of the necessary video frame rate (fps).

- Observers were presented a flat screen whose luminance changed sinusoidally:

$$f(x, y, t) = C(1 + s \cos 2\pi F_t t).$$

- C : constant luminance,
- F_t : temporal frequency,
- s intensity modulation level.

Spatiotemporal HVS models

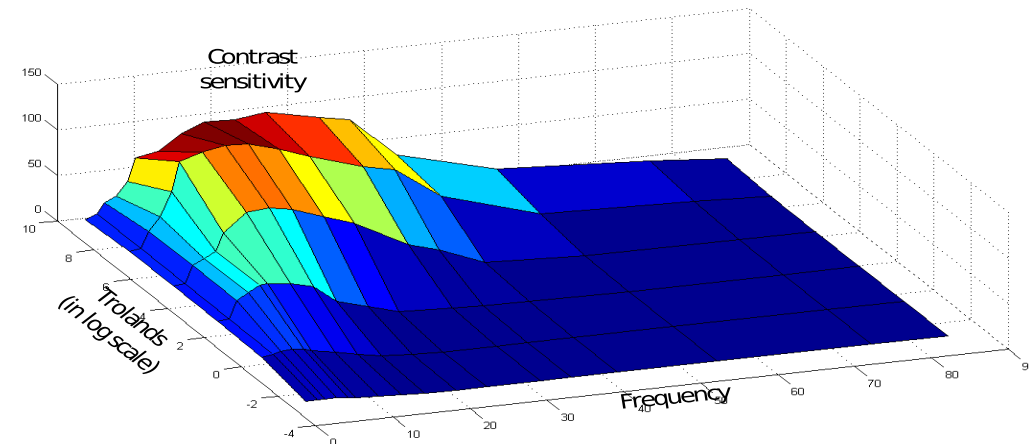
- Intensity modulation levels s were changed for constant luminance C .
- Observers were asked to identify lowest observable modulation level s_{min} .
- Observer **sensitivity** to temporal luminance variations at frequency F_t :

$$s_e = 1/s_{min}$$

Spatiotemporal HVS models

Contrast Sensitivity Function (**CSF**) $s_e(C, F_t)$.

- **Troland**: a unit to describe the light intensity entering the eye retina.
- CSF $s_e(C, F_t)$ is a band-pass function of both C, F_t .



Spatiotemporal HVS models

- Sharp reduction after a cutoff frequency.
- Peak frequency increases with the average image brightness.
- Human eye has low sensitivity to high temporal frequencies because it retains the sense of an image for a short time (***vision persistence***).
- It is caused by the temporal integration of the incoming light energy.

Spatiotemporal HVS models

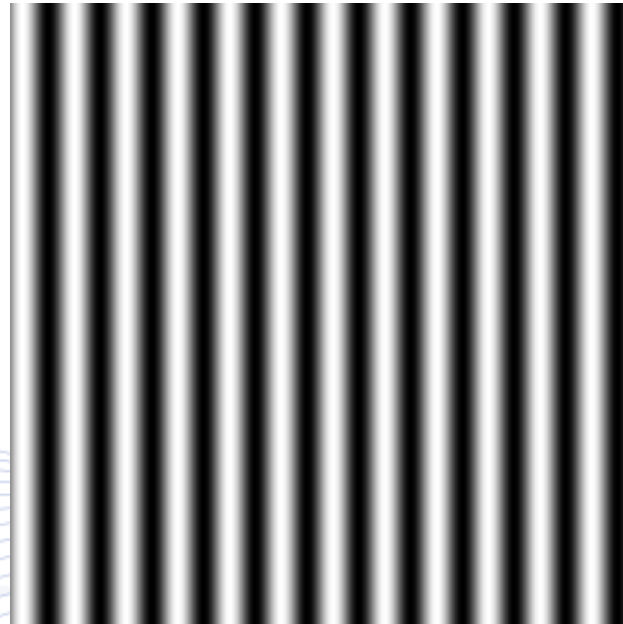
- **Block's law** states that the integration period is inversely proportional to that of the light intensity.
- The brighter the source is, the shorter is the retention period.
- Brighter screens (e.g., computer screens) need higher refresh rate (fps) to avoid flickering.
- For cinema (dark theater) and fps of 24 *Hz* is enough to avoid flickering.

Spatiotemporal HVS models

HVS Spatial Frequency Response (SFR):

- Assumption: spatial HVS sensitivity is isotropic, the spatial frequency response can be measured along any arbitrary spatial axis.
- Spatial sensitivity normalization to the observation distance: SFR expressed as a function of the ***spatial angular frequency***.

Spatiotemporal HVS models



Horizontal 2D sinusoidal signals having $(F_x, F_y) = (6, 0)$.

Spatiotemporal HVS models

Kelly experiments

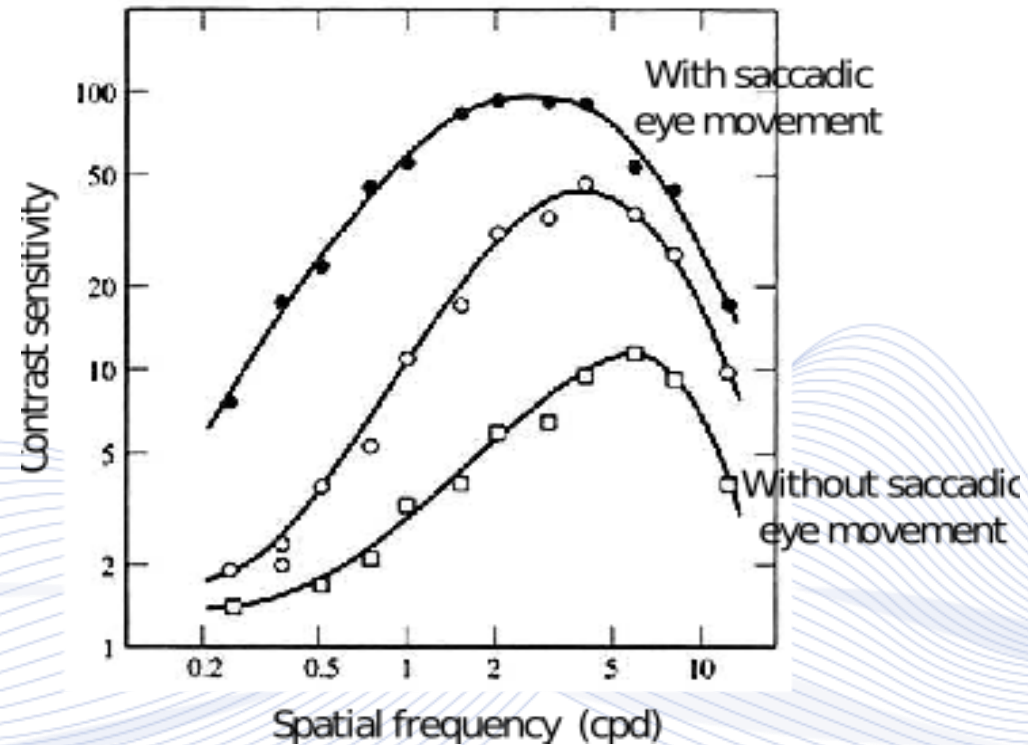
- Human observers observe vertical sinusoidal patterns of amplitude Cs and frequency F_x superimposed on a constant background having illumination C :

$$f(x, y) = C(1 + s \cos 2\pi F_x x).$$

- When they first observe the existence of the pattern having amplitude Cs_{min} , contrast sensitivity is defined by:

$$S_e = \frac{1}{s_{min}}.$$

Spatiotemporal HVS models

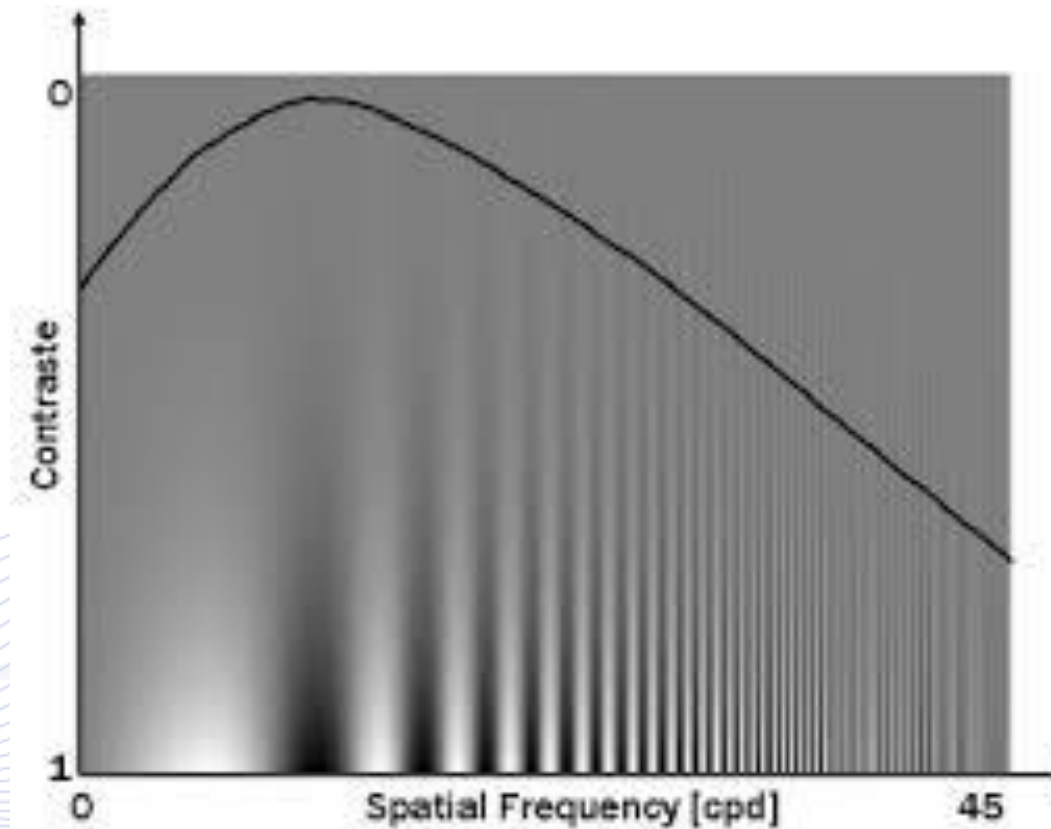


Spatial HVS frequency response.

Spatiotemporal HVS models

- Peak SFR is observed at about 2-5 cycles/degree.
- Cutoff spatial frequency is 30 cycles/degree.
- No details can be observed on a fast moving object.
- SFR varies with eye motion configuration:
 - **Saccadic motion** allows much higher spatial HVS sensitivity.

Spatiotemporal HVS models



Contrast sensitivity.

Spatiotemporal HVS models

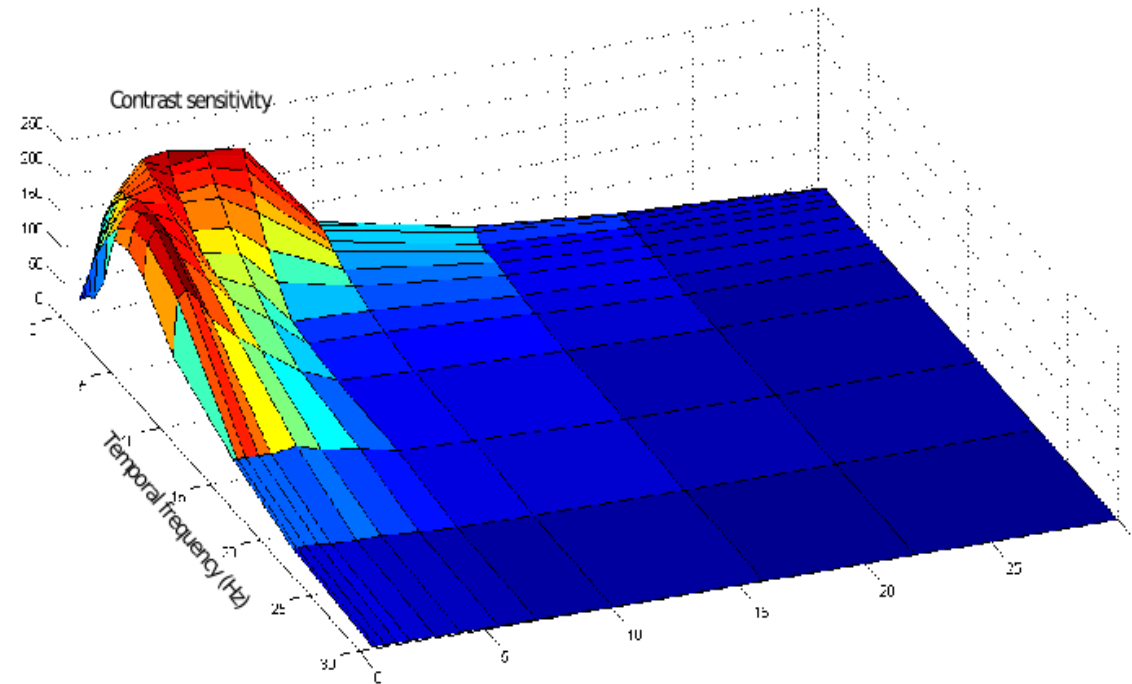
HVS spatiotemporal frequency response experiments:

- Test pattern:

$$f(x, y, t) = C(1 + s \cos(2\pi F_x x) \cos(2\pi F_t t)).$$

- For a fixed pair of F_x and F_y the modulation level s was changed.
- The observer was requested to determine the minimal observable modulation level s_{min} .

Spatiotemporal HVS models



Spatiotemporal HVS contrast sensitivity as a function of F_x , F_t for unconstrained eye motion.

Spatiotemporal HVS models

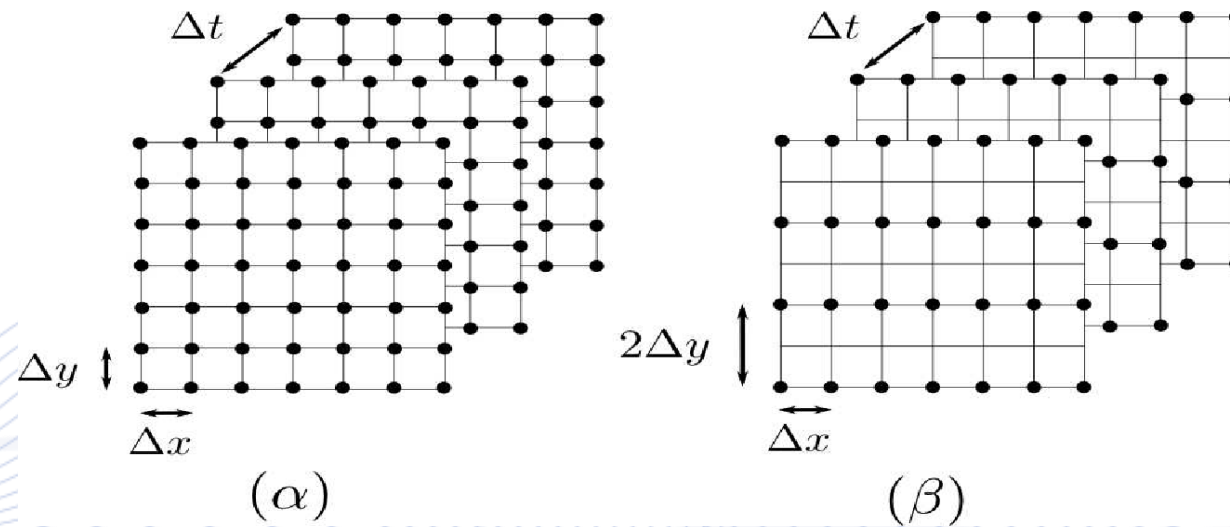
- When the temporal(spatial) frequencies are close to zero, the spatial (temporal) HVS frequency response has band-pass characteristics.
- At high temporal (spatial) frequencies, the spatial (temporal) frequency HVS response is low pass:
- Its peak response frequency reduces, when the temporal (spatial) frequency increases.
- When an object moves too quickly, the eye cannot distinguish object details (high spatial frequencies).

Spatiotemporal HVS models

- HVS can discern object details (higher spatial frequencies) much better, when the object image is stationary.
- Similarly, at higher spatial frequencies the temporal response becomes low pass.
- Implication of this inverse relationship between spatial and temporal HVS responses:
 - ***spatial video resolution can be swapped with temporal one and vice versa.***

Spatiotemporal HVS models

- 2:1 Interlaced video takes advantage of HVS properties.



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

Spatiotemporal HVS models

2:1 interlaced video:

- fast moving scenes:
 - They can be visualized at a limited frame rate, by dividing a frame into two temporally adjacent fields, each with half the number of frame lines.
- slow moving scenes:
 - the lines in two consecutive fields are perceived as one high spatial resolution video frame.

Spatiotemporal HVS models

Smooth eye pursuit movement:

- Limited spatial sensitivity for fast moving objects, but eye does not.
- Observed image on eye retina, when eye moves:

$$\tilde{f}(x, y, t) = f(x + v_x t, y + v_y t, t),$$

- $f(x, y, t)$: displayed image,
- (v_x, v_y) : eye motion velocities.

Spatiotemporal HVS models

- Eye motion cancels object motion on retina.
- Details of a fast moving object can be seen by eye pursuit movement:
 - Increased spatial HVS sensitivity.

Moving Image Perception

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Video Quality Assessment

- In many cases, humans are the final video consumers.
- Perceived video quality must be quantified.
- **Video Quality (VQ)** is influenced by:
 - Acquisition noise;
 - Compression effects;
 - Transmission errors.
- VQ assessment can help meet video storage and transmission requirements.

Video Quality Assessment

Subjective video quality assessment:

- Ask humans to watch the video and assess its quality.
- ***Mean Opinion Score (MoS)***: scale [1, ..., 5].
- 1: worst, 5: best quality.
- Labor intensive and expensive.
- A large number of viewers is needed to lower score variability and provide statistical certainty.
- Impossible to assess all videos before broadcasting.
- Useful in providing a golden standard for automated VQA methods.

Video Quality Assessment

Objective video quality assessment.

- No human observers involved.
- ***Full reference VQA algorithms*** operate on distorted video, while employing the original video reference for comparison.
- VQA measures:
 - Mean Square Error,
 - Peak Signal to Noise Ratio,
 - SSIM.

Video Quality Assessment

- **Reduced reference VQA algorithms** use distorted video and some original video information, e.g., edge locations.
- **Blind VQA algorithms:** no knowledge about distortion nor reference video.

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Q & A

Thank you very much for your attention!

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

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