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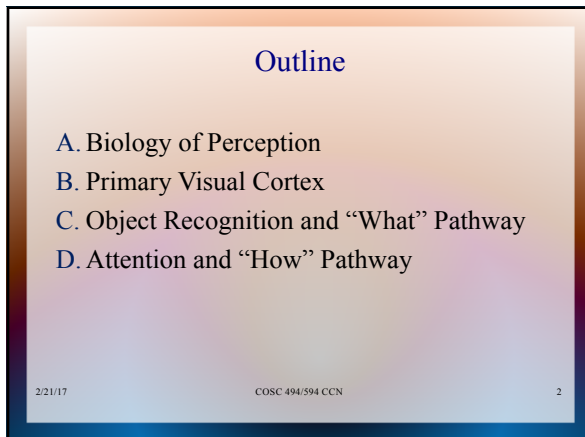
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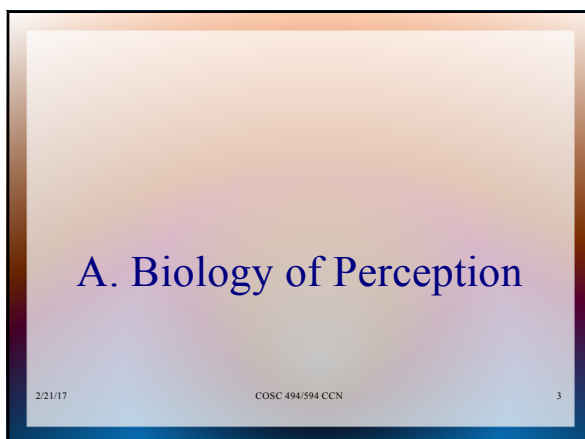
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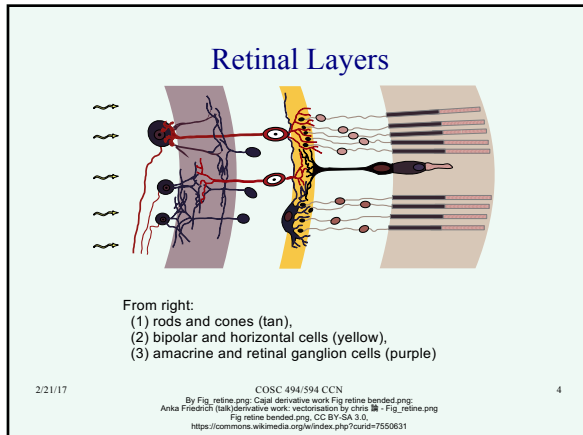
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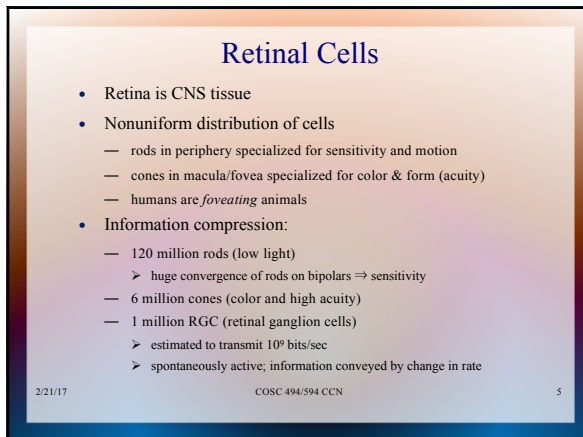
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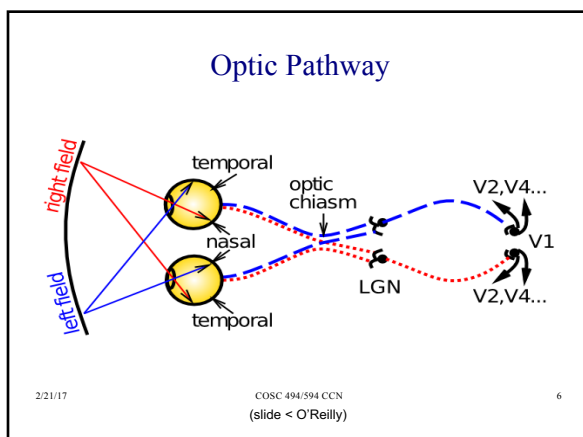
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### Key Organizing Principles

- Transduction of different information
  - wavelength (rods; blue, green, red cones)
  - spatial frequency (resolution)
  - motion
- Topographic organization
  - contrasting similar information
- Filtering to extract relevant information

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### Retinal Contrast Filtering

a) On-center      b) Off-center      c) Contrast sensitive

2/21/17 COSC 494/594 CCN (slide < O'Reilly) 8

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### LGN of the Thalamus

- A “relay station,” but also much more
- Organizes different types of information into different layers with aligned retinotopic maps
- Performs dynamic processing: magnocellular motion processing cells, attentional processing
- On- and off-center information from retina is preserved in LGN

2/21/17 COSC 494/594 CCN (slide based on Frank) 9

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### Structure of LGN

**Parallel pathways**

- Cells have monocular input
- Six layers alternate input from two eyes (RGC)

2/21/17 COSC 494/594 CCN 10  
 (figs < Heeger)

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### V1

**V1 simple cell edge detector**

2/21/17 COSC 494/594 CCN 11  
 (slide < O'Reilly) (Actual Gabor filters for high-res pathway)

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### “What” vs. “Where” Pathways

- “What” ignores differences in location, illumination, size, rotation
- “Where” emphasizes location, size, and ignores object identity

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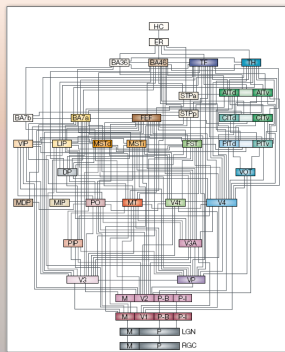
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## Hierarchy of Macaque Visual Areas



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(fig. from Van Essen & al. 1992)

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## Principal Regions in "What" Pathway

- V1: Primary Visual Cortex
  - encodes image in terms of oriented edges
- V2: Secondary Visual Cortex
  - encodes in terms of intersections & junctions
- V4
  - more complex features over wider range of locations
- PIT: Posterior Inferotemporal (IT) Cortex
  - location & size invariant object recognition
  - includes FFA (fusiform face area)
- AIT: Anterior IT Cortex
  - abstract/semantic visual information

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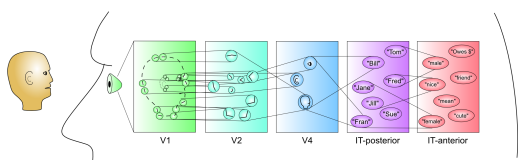
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## Hierarchy of Visual Detectors



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## B. Primary Visual Cortex

What is the origin of detectors for oriented bars of light?

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### Self-Organization of V1 Orientation Selective Neurons

**STIMULUS**

**RESPONSE**

Stimulus off   Stimulus on   Stimulus off

**TUNING CURVE**

FIGURE 4.8 Response of a single cortical cell to bars presented at various orientations.

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(slide < O'Reilly)

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### Topographic Maps

- Map of orientations
- Hypercolumn: Full set of coding for each position
- Pinwheel can arise from learning and lateral connectivity: not hard-wired!

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(slide based on O'Reilly)

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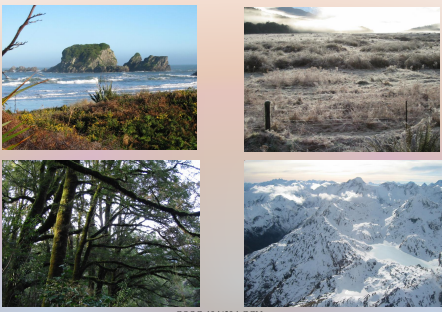
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### What is Common?



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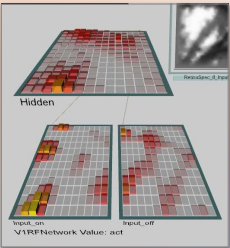
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### V1Rf: Simulating One Hypercolumn

- Natural visual scenes are preprocessed by passing them (separately) through layers of on-center and off-center inputs
- Hidden layer: edge detectors seen in layers 2/3 of V1; Layer 4 (input) just represents unoriented on/off inputs like LGN (but can be modulated by attention)
- Circular neighborhood of lateral excitatory connectivity in Hidden layer



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## emergent demonstration: V1Rf

2/21/17 COSC 494/594 CCN 21

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### Self-Organized Topography

Model shows how documented V1 properties can result from interactions between learning, architecture (connectivity), and structure of environment

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### Faces vs. Natural Scenes

Faces

Nature Scenes

Some differences, but pinwheels still emerge

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### Rewiring Cortex

- Experiments by Mriganka Sur & colleagues (MIT)
- What happens if retinal axons are redirected into auditory thalamus (MGN) instead of its usual inputs?
- Answer: Auditory cortex (A1) develops orientation columns and retinotopic maps similar to V1
- Animals experience activity in A1 as visual perception

(a) Normal

(b) Rewired

Auditory cortex

Visual cortex

Superior colliculus

Inferior colliculus

LGN

MGN

Visually responsive auditory cortex

Visual cortex

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### Orientation Columns in A1

Orientation columns develop in A1 similar to those in V1

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### Are They Having Visual Experiences?

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### They Are Having Visual Experiences

Group	Condition	Sound	Left light	Centre light	Right light
Rewired ferret R1	Pre-LGN/LP lesion	~90	~90	~90	~90
	Post-LGN/LP lesion	~90	~90	~90	~90
	Post-A1 lesion	~90	~90	~90	~90
	Rewired	~90	~90	~90	~90
Rewired ferret R2	Pre-LGN/LP + SC lesion	~90	~90	~90	~90
	Post-LGN/LP + SC lesion	~90	~90	~90	~90
	Normal	~90	~90	~90	~90
	Pre-LGN/LP + SC lesion	~90	~90	~90	~90

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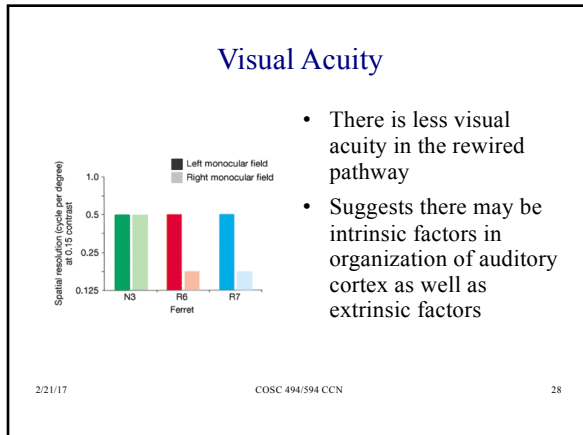
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## C. Object Recognition and “What” Pathway

How do we recognize objects (across locations, sizes, rotations with wildly different retinal images)?

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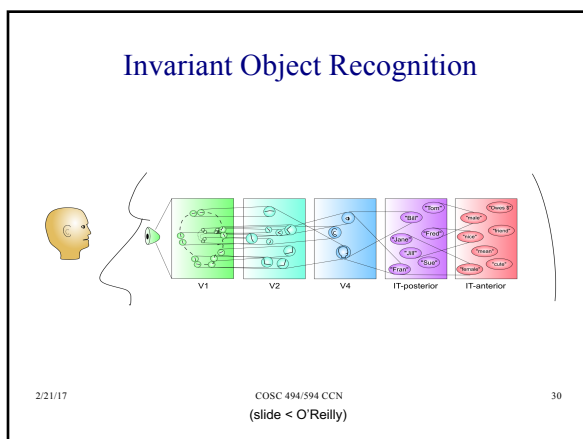
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### It's Hard

No Overlap

Output = "A"

Output = "F"

Visual Inputs

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(slide < O'Reilly) 31

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### Invariant Object Recognition

- Hierarchy of increasing:
  - Feature complexity
  - Spatial invariance
- Increasing RF size:
  - Conjunction of features (to form more complex objects)
  - Collapsing over location information ("spatial invariance")
- Strong match to RF's in corresponding brain areas

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(slide based on O'Reilly & Frank) 32

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### Biological Data: Increasing Complexity and Invariance

V2	V4	posterior IT	anterior IT

(%) Smax/MAX

(%) receptive field size

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(slide < O'Reilly) 33

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The Model: combining Fukushima with convolutional neural nets, bidirectional connectivity and learning!

V1 = oriented line (edge) detectors, hard-coded  
 V4 units encode conjunctions of V1 edges across a subset of space  
 Each IT unit pays attention to all of V4  
 (V2 omitted here, important for figure-ground etc)

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 (slide < Frank)

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### V1 Receptive Fields

- 4x5 hypercolumns
- Two rows of simple cells at 4 orientations and two polarities
- Two rows of end-stop complex cells
- One row of length-sum complex cells
- 50% overlap with adjacent hypercolumns

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### Simple Textbook Test

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### Activation-Based Receptive Fields

- How do we plot receptive fields for V4?
- Receiving weights show which V1 units a V4 unit responds to, but they don't show what thing in the world the unit responds to
- **Solution:** Show the network lots of input patterns.
- Then, display a composite of all the input patterns that activate the unit (weighted by activity).

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### V4 Receptive Fields

V1	V4	IT	Dend	Image

- Some V4 units code for location-specific conjunctions of V1 features
- This will show up as a sharp receptive field for Image input

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(slide < Frank)

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### V4 Receptive Fields

V1	V4	IT	Dend	Image

- Some V4 units code for simple features in a location invariant way
- This will show up as smeary parallel lines in Image input

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(slide < Frank)

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### V4 Receptive Fields

VI	V4	IT	Output	Image
[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]
[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]
[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]
[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]	[Grid of small images]

- Can also look at which Output units tend to get active for any given V4 unit
- Generally a given V4 unit is associated with multiple objects

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(slide < Frank)

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### 3D Object Recognition Test



3D models from Google SketchUp

100 categories

9–10 objects per category

2 objects left out for testing

+/- 20° horiz depth rotation  
+ 180° flip

0–30° vertical depth rotation

14° 2D planar rotations

25% scaling

30% planar translations

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<http://grey.colorado.edu/CompCogNeuro/index.php/CU3D> (slide < O'Reilly)

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### Depth & Lighting Variations for One Object



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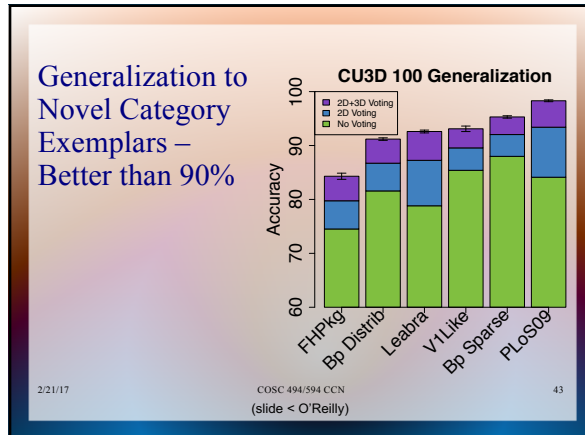
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emergent demonstration:  
Objrec

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D. Attention and “How”  
Pathway

Why is visual system split into what/where pathways?  
Why does parietal damage cause attention problems (neglect)?

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### Some Functions of Dorsal Pathway

- “Where” pathway (spatial relations)
  - visual attention (this chapter)
- But more broadly “how” pathway
  - maps perception to action (next chapter)
- Numerical and mathematical processing
- Representation of abstract relationships
- Modulation of episodic memory
- Aspects of executive control

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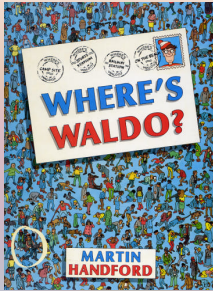
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### Spatial Attention and Neglect



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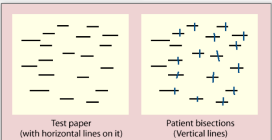
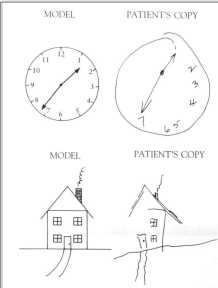
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### Hemispatial Neglect

Mainly from injuries to right parietal cortex



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(slide based on O'Reilly)

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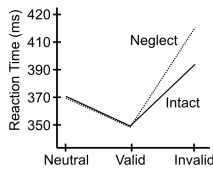
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### Posner Task

- Valid cues speed performance (relative to "no cue" condition)
- Invalid cues slow performance (relative to "no cue" condition)



Cue Type	Intact (ms)	Neglect (ms)
Neutral	~370	~365
Valid	~345	~355
Invalid	~395	~420

Valid Trial

Cue: [ ] + [ ]

Target: [ \* ] + [ ]

Invalid Trial

Cue: [ ] + [ ]

Target: [ ] + [ \* ]

- Patients perform normally in the "neutral" (no cue) condition, regardless of where the target is presented
- Patients benefit just as much as controls from valid cues
- Patients are hurt more than controls by invalid cues

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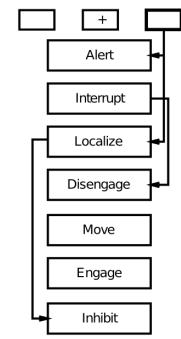
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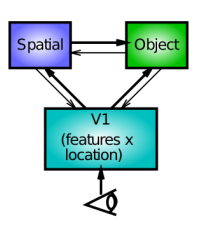
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### Models: Boxology vs. Biology





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### Posner Task Simulation

- Model explains the basic finding that valid cues speed target processing, while invalid cues hurt
- Also explains finding that patients with small unilateral parietal lesions benefit normally from valid cues in ipsilateral field but are disproportionately hurt by invalid cues
- No need to posit "disengage" module
- Also explains finding of **neglect** of contralateral visual field after large, unilateral parietal lesions when some stimulus is present in ipsilateral field ("extinction")

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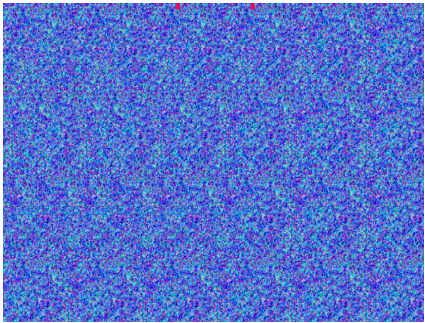
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What's Missing? Lacking Depth..



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Supplementary

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Gabor Uncertainty Principle  
and Gabor Elementary  
Functions

MacLennan, B. J. *Gabor Representations of Spatiotemporal Visual Images*. University of Tennessee, Knoxville, Computer Science Department technical report CS-91-144, September 1991

2/22/17 COSC 494/594 CCN 54

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### Dennis Gabor

- Dennis Gabor (1900–79) is the father of holography (1947, 1971 Nobel Prize in Physics)
- “the future cannot be predicted, but futures can be invented”
- Developed a theory of information (1946) complementary to Shannon’s theory
- Gabor Uncertainty Principle based on same mathematics as derivation of Heisenberg Uncertainty Principle
- Nearly optimal Gabor representations are used in primary visual cortex

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### Time to Detect Difference in Frequency

$\Delta f \Delta t \geq 1$

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### A Little More Formally...

- **Nominal duration** ( $\Delta t$ ) = duration of rectangular pulse with same area as signal and height equal to amplitude at origin
- Hence,  $\Delta t |\varphi(0)| = \int_{-\infty}^{\infty} |\varphi(t)| dt$
- Some details omitted

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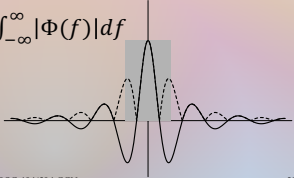
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### A Little More Formally (2)

- **Nominal bandwidth** ( $\Delta f$ ) of spectrum = width of rectangular pulse with height equal to spectrum's amplitude at origin and same area as absolute value of spectrum
- Hence,  $\Delta f |\Phi(0)| = \int_{-\infty}^{\infty} |\Phi(f)| df$



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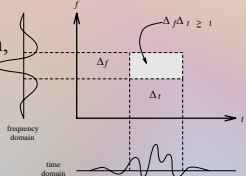
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### A Little More Formally (3)

- Computing the Fourier transform at origin,  $|\Phi(0)| \leq \Delta t |\varphi(0)|$
- So  $\Delta t \geq |\Phi(0)|/|\varphi(0)|$
- Computing the inverse Fourier transform at origin,  $|\varphi(0)| \leq \Delta f |\Phi(0)|$
- So  $\Delta f \geq |\varphi(0)|/|\Phi(0)|$
- Hence,  $\Delta f \Delta t \geq 1$



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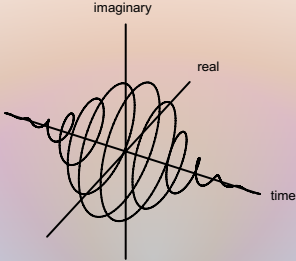
### 1D Gabor Elementary Function

imaginary

real

time

Gaussian modulated complex exponential



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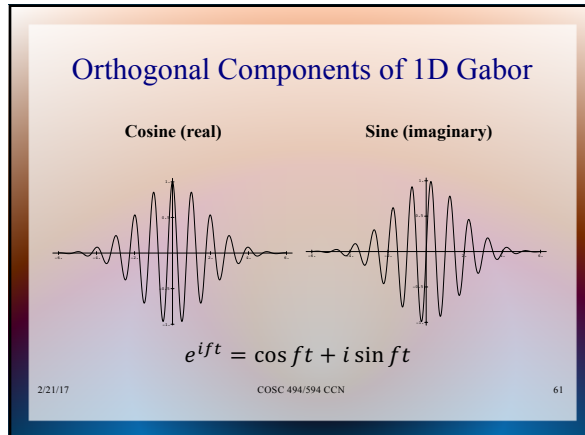
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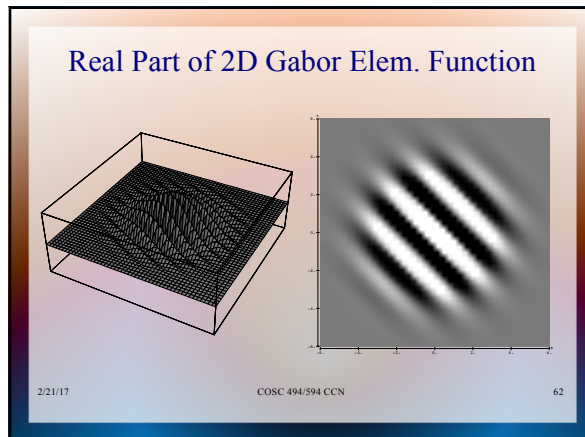
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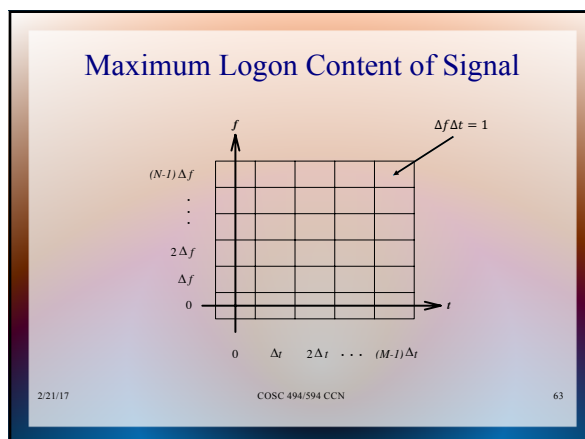
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### Maximum Logon Content

- If  $T = M\Delta t$  is the duration and  $F = N\Delta f$  is the bandwidth
- The maximum number of logons  $MN$  is achieved when  $\Delta t\Delta f = 1$  (i.e., Gabor elementary functions)
- In general, the area doesn't have to be divided into rectangles of the same shape, so long as area is 1
- So the maximum logon content is  $TF$  (duration times bandwidth)
- Any such signal can be represented uniquely as a sum of  $TF$  Gabor elementary functions

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### Gabor Representations

- Any “finite energy” function  $\psi$  of finite duration  $D$  and finite bandwidth  $F$  is equal to a linear superposition of Gabor elementary functions:

$$\psi(t) = \sum_{j=0}^{M-1} \sum_{k=0}^{N-1} a_{jk} C_{jk}(t) + b_{jk} S_{jk}(t)$$

where  $C_{jk}(t) = e^{-\pi(t-j\Delta t)^2/\alpha^2} \cos[2\pi k\Delta f(t-j\Delta t)]$

and  $S_{jk}(t) = e^{-\pi(t-j\Delta t)^2/\alpha^2} \sin[2\pi k\Delta f(t-j\Delta t)]$

- The same applies in higher dimensions.

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### Gabor Filters in Early Vision

- Measurements of receptive fields of simple cells in cat visual cortex have show them to be like Gaussian-modulated sinusoids (Jones & Palmer, 1987)
- Daugman (1984, 1985, 1993) showed 97% of them are statistically indistinguishable from the odd- or even-symmetric parts of a 2D Gabor elementary function
- Adjacent simple cells have grating patches that are  $90^\circ$  out of phase, but matched in preferred orientation and frequency
- And more... (MacLennan, 1991)

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