

# **EECS 4422/5323**

## Unit 7: Spatiotemporal analysis

# Outline

- **Introduction**
- **Orientation in visual space-time**
- **A representation for spatiotemporal patterns**
- **Spatiotemporal boundaries**
- **A framework for spatiotemporal analysis**
- **Applications**
- **Summary**

# Introduction

- We have considered the analysis of spatial structure.
  - Oriented, bandpass representations.
- We have considered analysis of the temporal dimension
  - Motion
- Now we consider the integrated analysis and interpretation of the spatial and temporal dimensions.
  - Spatiotemporal analysis

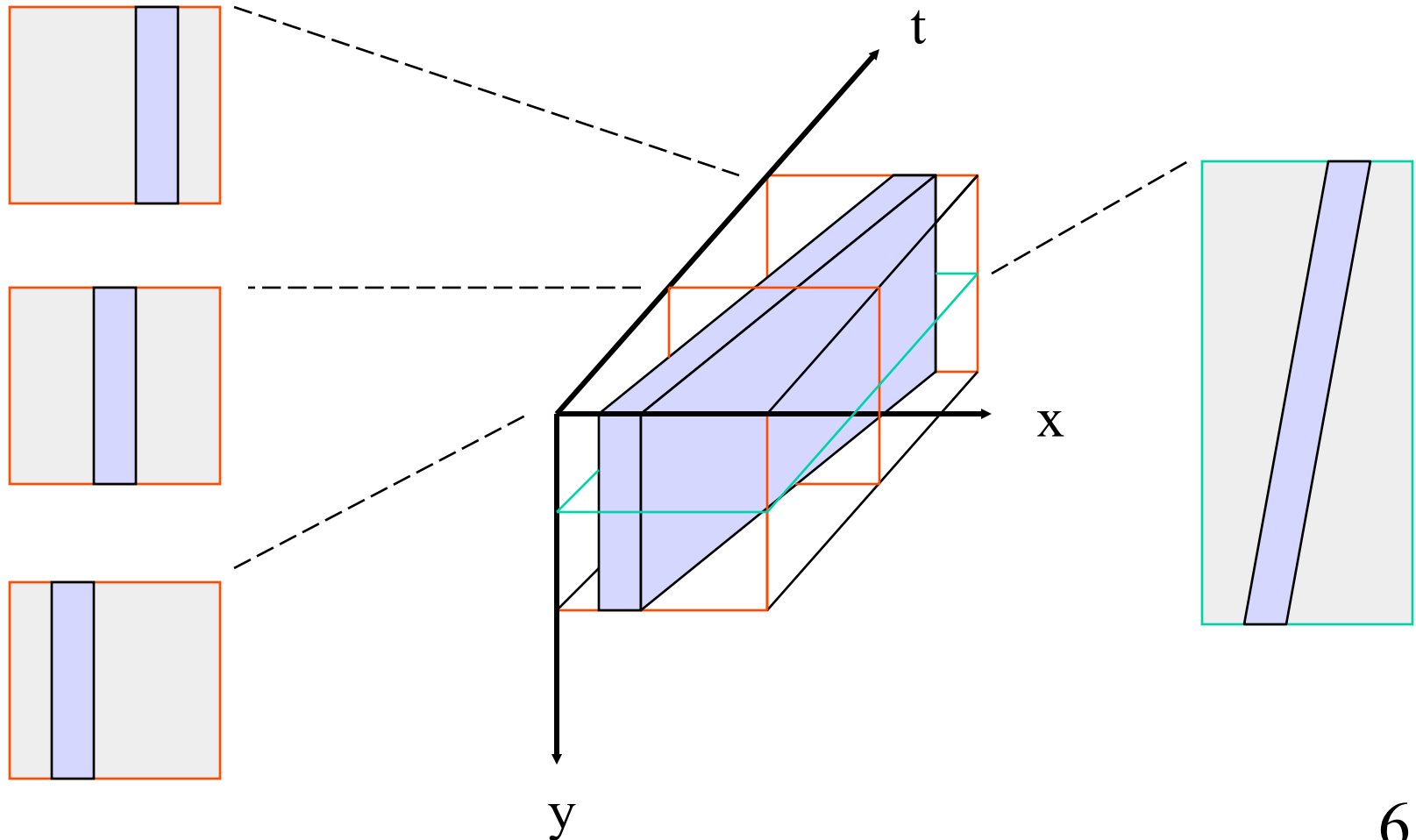
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# Orientation in visual space-time: Basics

- The local orientation (or lack thereof) of a pattern is one of its most salient characteristics.
- Geometrically, orientation captures the local first-order structure of a pattern.
- For vision, local spatiotemporal orientation can have additional interpretations.
  - Image velocity is manifest as spatiotemporal orientation.
  - And more...

# Orientation in visual space-time: Graphic

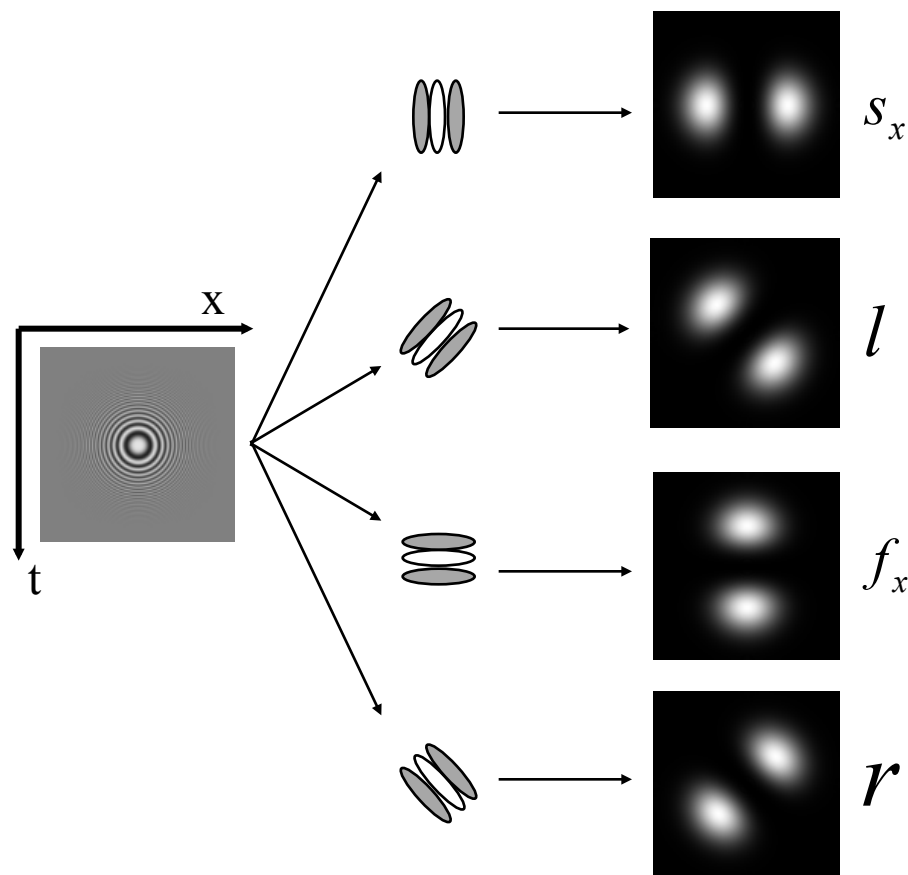


# Orientation in visual space-time: Representation

- Goal is to analyze spatiotemporal data according to its local orientation structure.
  - Consider orientation in x-t and y-t planes, with local weighted averaging in orthogonal spatial dimension.
  - Filter for multiple bands each tuned for certain orientations in a spatiotemporal plane.
  - For example, select 4 orientations/plane: horizontal, vertical, 2 diagonals
- Consider single spatiotemporal scale (for now).

# Orientation in visual space-time: Filtering

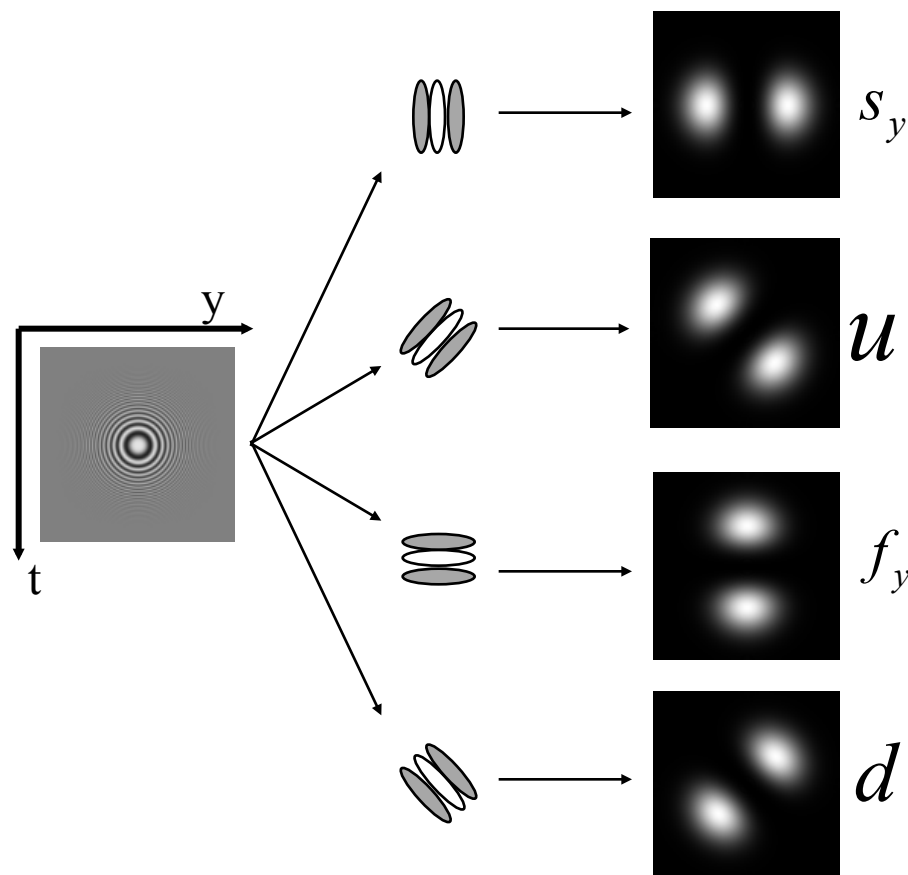
- Apply filters tuned to 4 different orientations in both x-t and y-t domains.
- In general, might consider additional directions.
- Filter specifics:
  - Oriented bandpass filters in spatiotemporal slice.
  - Lowpass filter in orthogonal spatial dimension.
  - Pointwise squared to yield local “oriented energy”.





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# Orientation in visual space-time: Normalization

- For any given orientation, the filter response is a joint function of
  - orientation
  - contrast
- Normalization yields purer measure of orientation

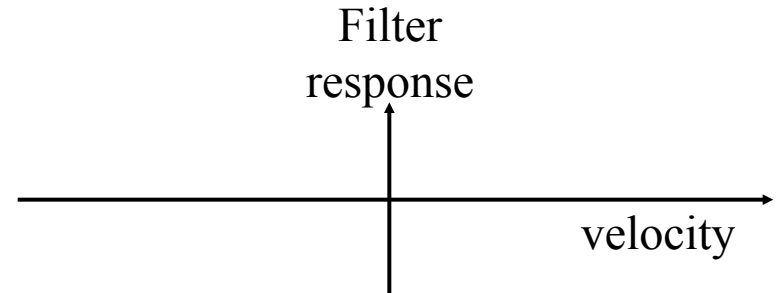
$$R(x, y, t) = \frac{r(x, y, t)}{r(x, y, t) + l(x, y, t) + s_x(x, y, t) + f_x(x, y, t) + \varepsilon}$$

with  $\varepsilon$  a small bias added for stability.

- Similarly for  $l$ ,  $s_x$ ,  $f_x$  and their y-t counterparts.

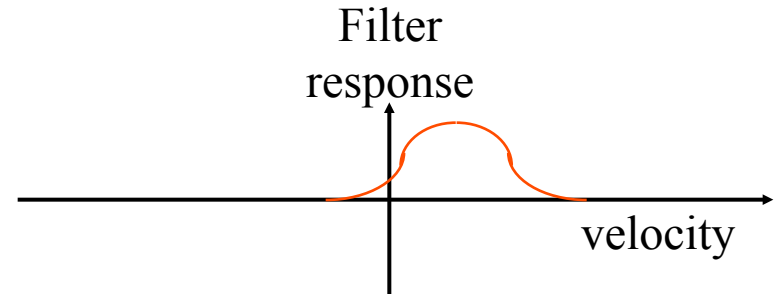
# Orientation in visual space-time: Velocity

- Consider the response to a horizontally moving pattern of  $r$ ,  $l$  and  $s$  filters.



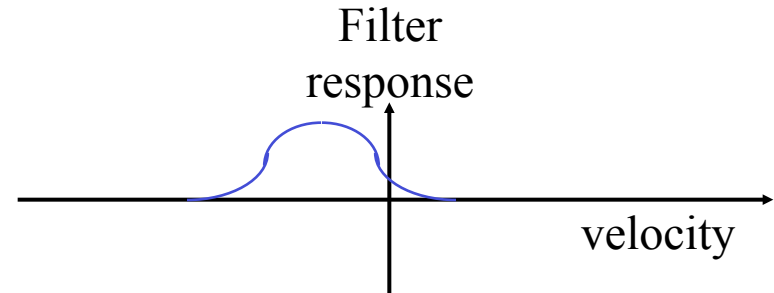
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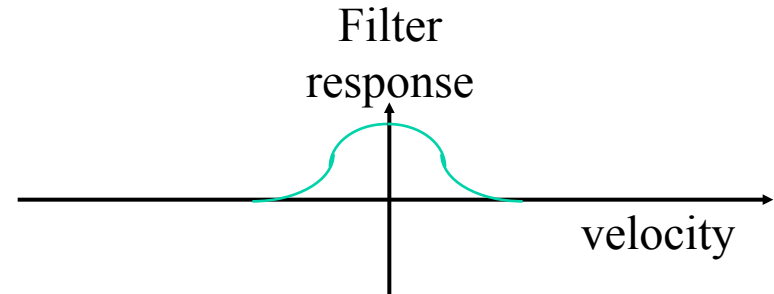
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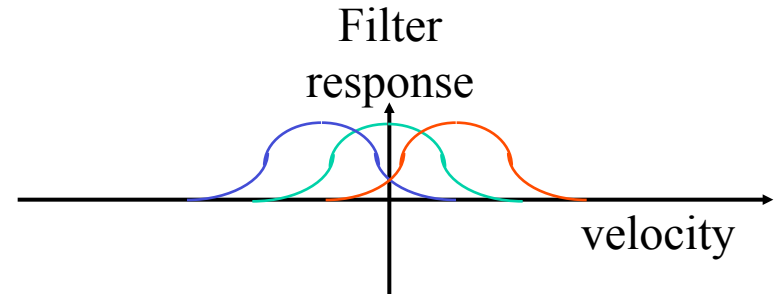
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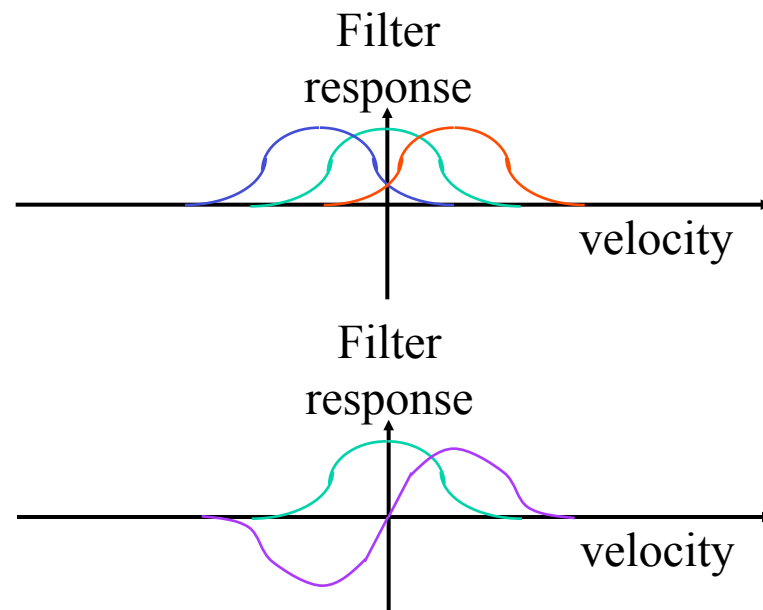
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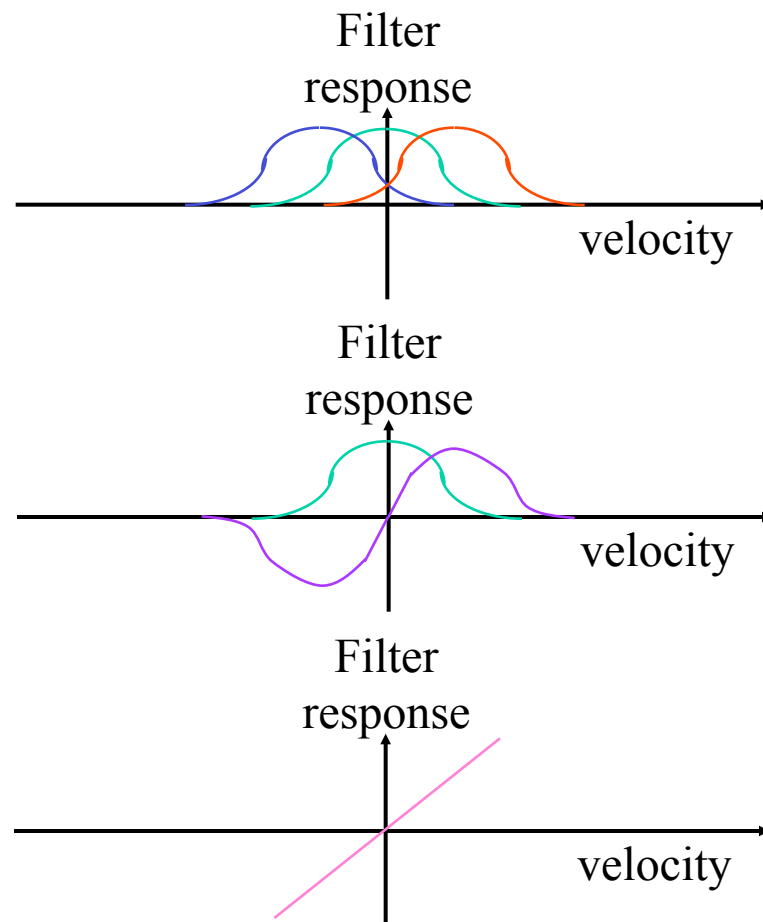
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# Orientation in visual space-time: Velocity

- Consider the response to a horizontally moving pattern of  $r$ ,  $l$  and  $s$  filters.
- By taking the difference  $r-l$  we get a single response that is properly signed WRT velocity.
- Dividing to get  $(r-l)/s$  yields a response that is (approximately) linear with velocity.



# Orientation in visual space-time: Velocity

## Comparison with optical flow constraint equation (OFCE)

- Recall the OFCE as derived from constant brightness assumption

$$E_x u + E_y v + E_t = 0$$

# Orientation in visual space-time: Velocity

## Comparison with optical flow constraint equation (OFCE)

- Recall the OFCE as derived from constant brightness assumption

$$E_x u + E_y v + E_t = 0$$

- Let us restrict consideration to one spatial dimension + time

$$E_x u + E_t = 0$$

- Now, we can directly solve for (1D) velocity

$$u = -E_t / E_x$$

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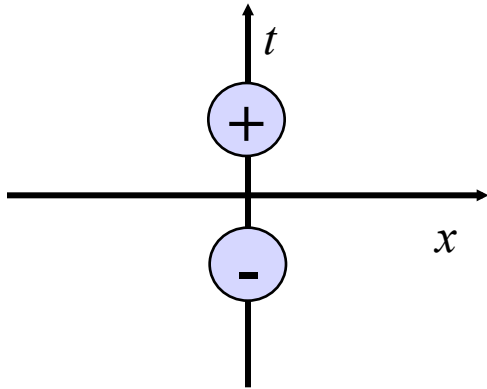
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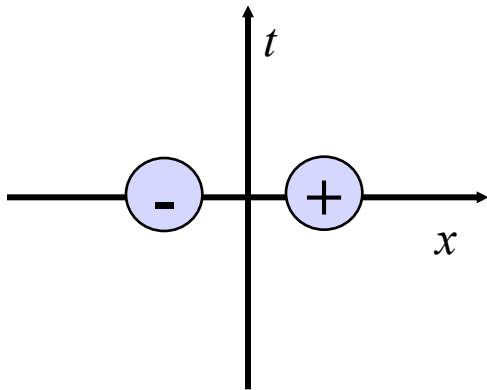


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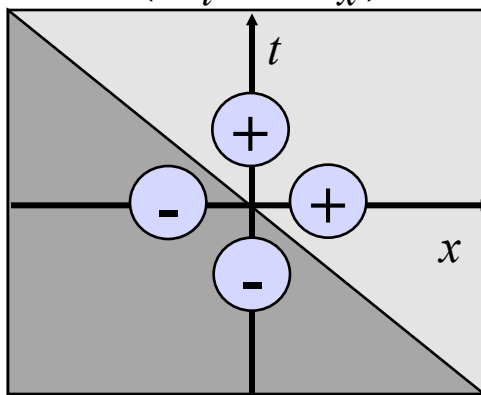
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The diagram shows a square grid with a vertical axis labeled  $t$  and a horizontal axis labeled  $x$ . A diagonal line runs from the top-left corner to the bottom-right corner. The region above and to the right of this diagonal is shaded light gray, while the region below and to the left is shaded dark gray. Four blue circles are arranged in a cross pattern around the center. The top circle contains a plus sign (+), the bottom circle contains a minus sign (-), the left circle contains a minus sign (-), and the right circle contains a plus sign (+).



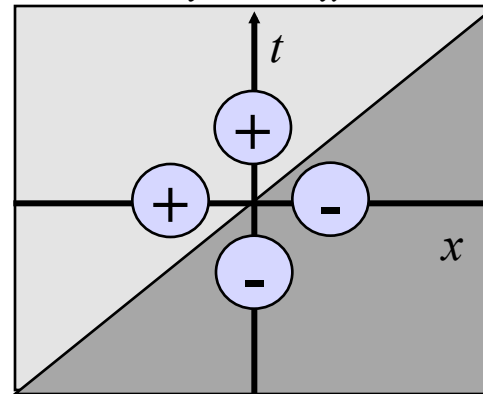
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$$\tilde{r} = (E_t - E_x) / 2$$

- Indeed, as we are concerned not with sign for a given direction, but rather magnitude, it suffices to consider the squared filter responses

$$l = (E_t^2 + 2E_x E_t + E_x^2) / 4$$

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- Finally, to avoid being biased by locally large values of image contrast, we divide through by the square of a first-order measure of local contrast  $s = E_x^2$

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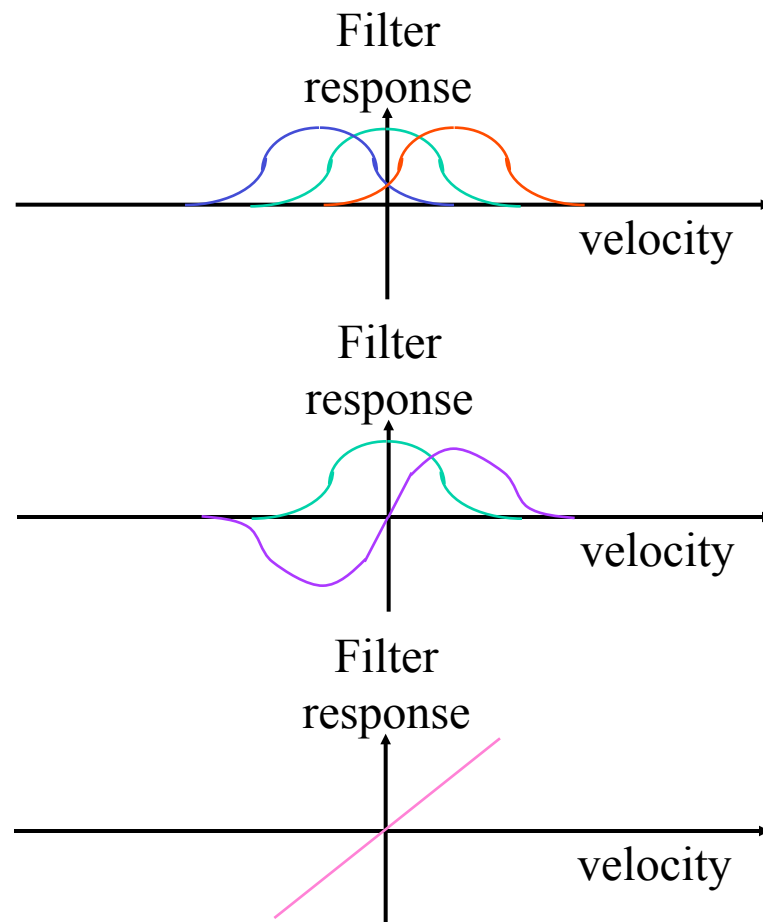
- We now recognize that this computation of velocity is equivalent to that based on the OFCE

$$E_x u + E_t = 0 \Rightarrow u = -E_t / E_x$$

# Orientation in visual space-time: Velocity

## Conclusion

- Oriented filters in visual space-time support the recovery of image velocity, along a particular direction.
- A “bank” of such filters can be used to span direction and provide an approach to optical flow estimation.



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# **A representation for spatiotemporal patterns: Motivation**

- When confronted with spatiotemporal data, an intelligent system can be overwhelmed by sheer quantity.
- An initial organization would be a key enabler for dealing effectively with data of this nature.
- The organization should afford distinctions that can guide subsequent processing.
- Distinctions that go beyond that of simple velocity.



# A representation for spatiotemporal patterns: General approach

- Parse stream of spatiotemporal data into primitive, yet semantically meaningful categories at the earliest stages of processing.
- Make distinctions along the following lines
  - What is moving and what is stationary?
  - Are the moving objects behaving coherently?
  - How much of the variance in the data is due to temporal brightness change?
  - Which portions of the data are simply too unstructured to support further analysis?

# **A representation for spatiotemporal patterns: General approach**

- Integrate information across both the spatial and temporal dimensions.
- Build on analysis of local orientation.
  - The simplest non-trivial characterization of local geometric structure.

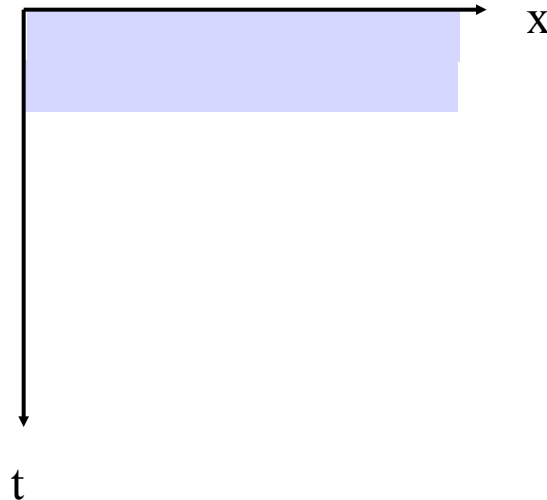
# A representation for spatiotemporal patterns: Primitive patterns

- Consider a spatiotemporal slice
  - As an observer views a **uniform pattern**



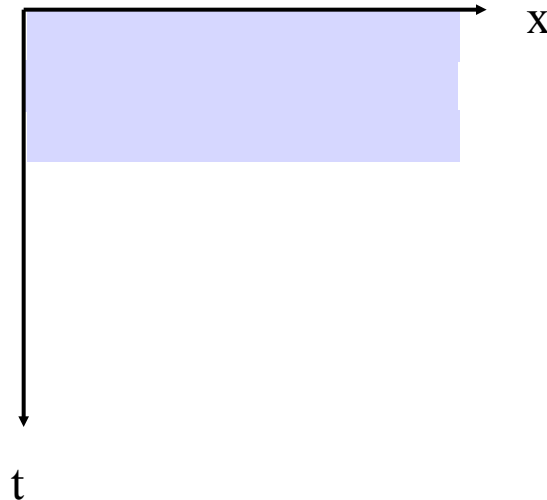
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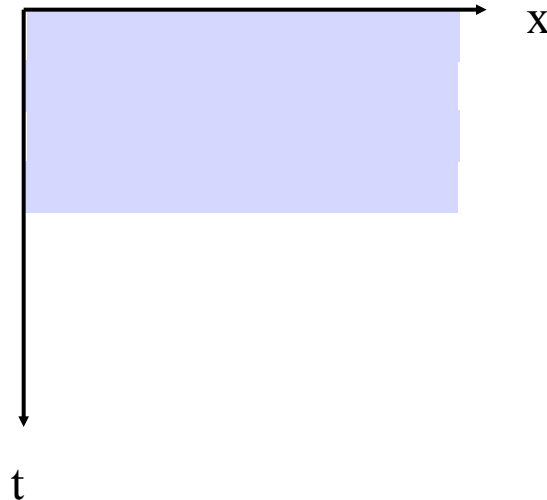
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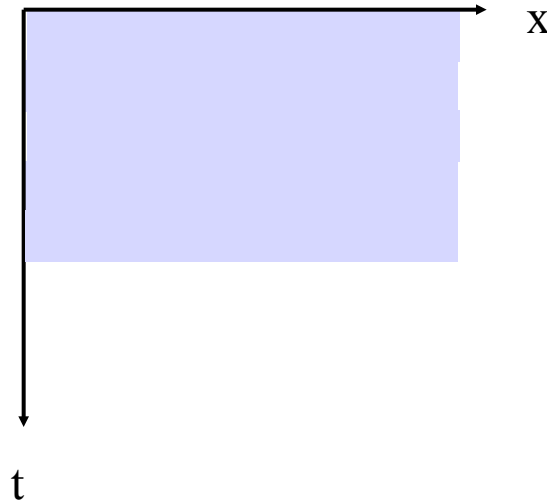
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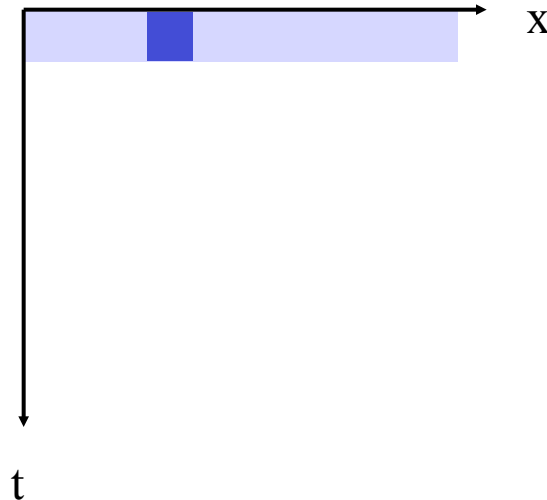
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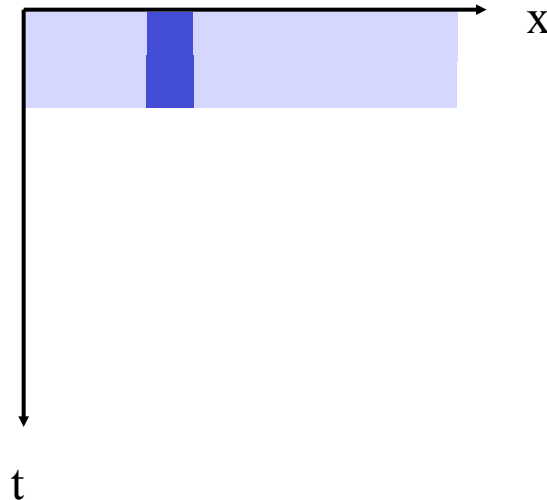
- Consider a spatiotemporal slice
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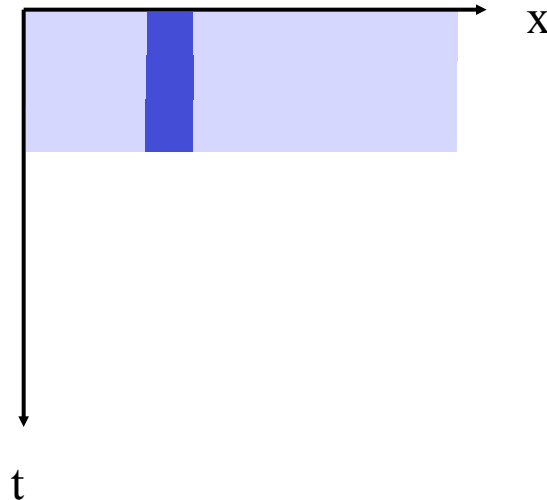
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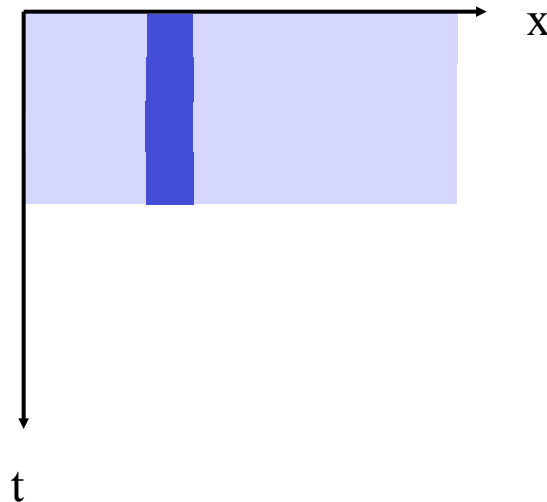
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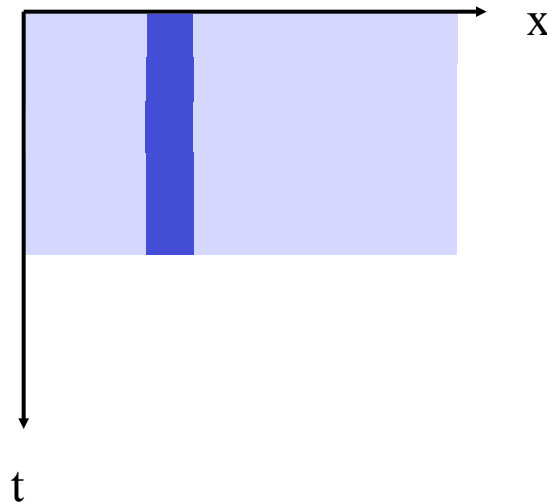
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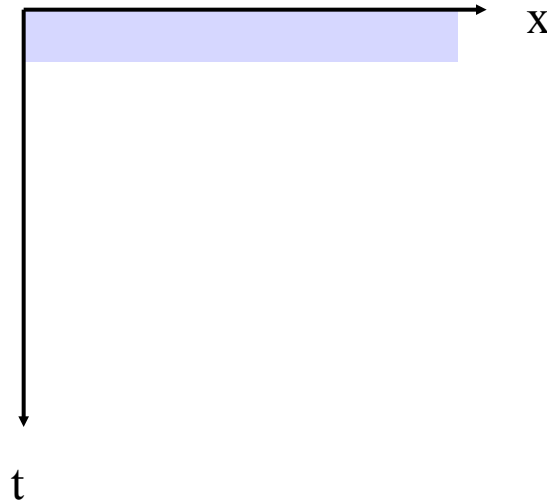
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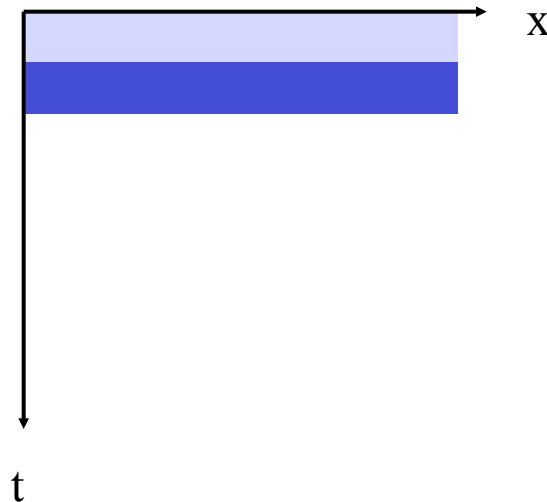
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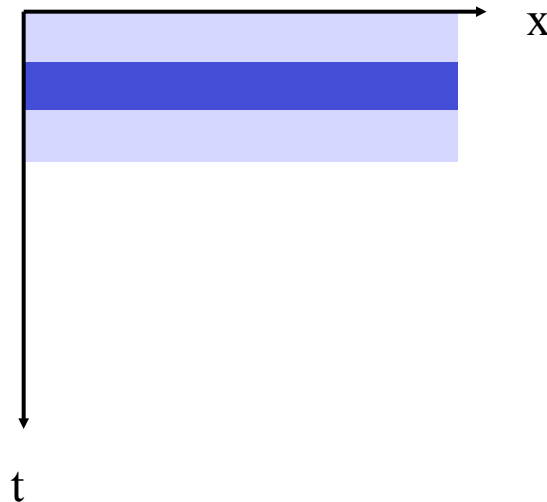
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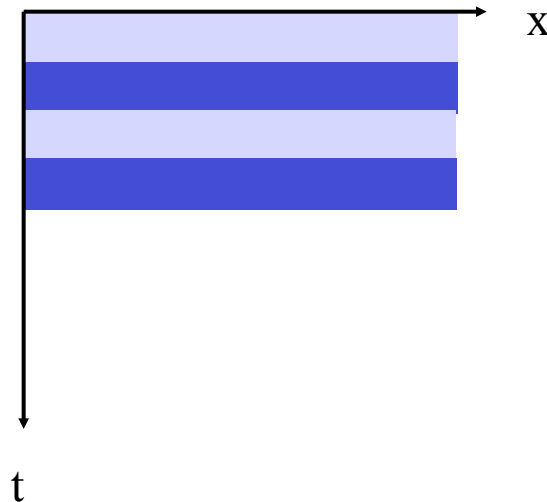
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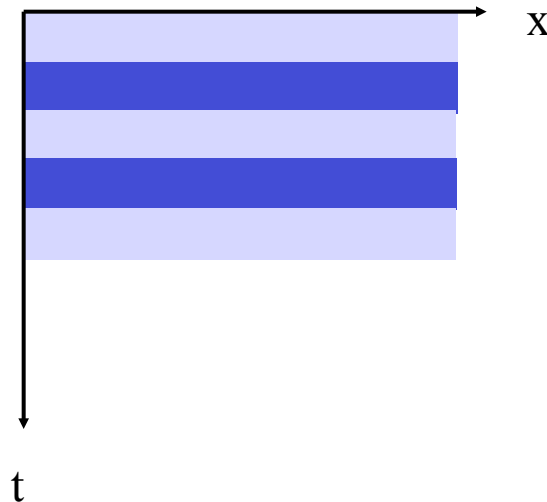
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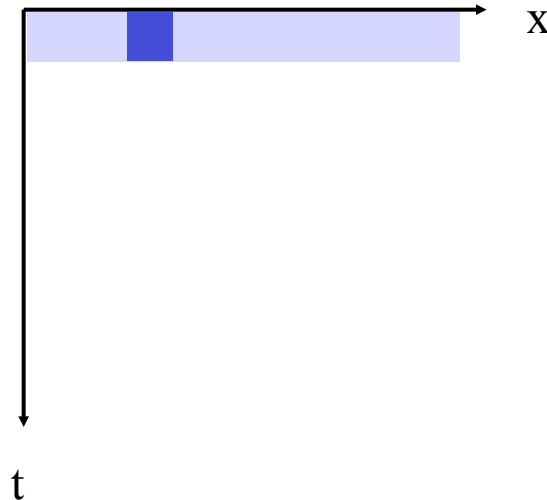
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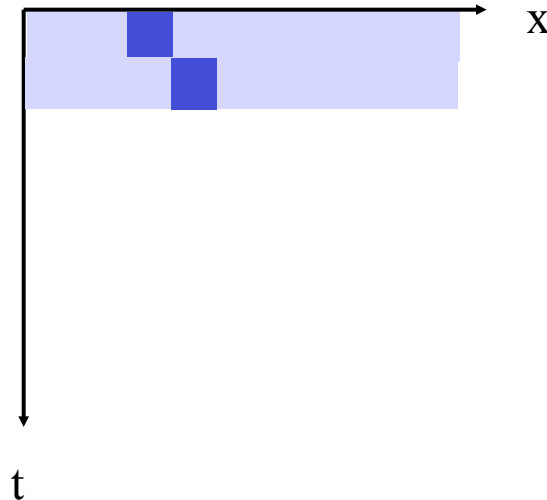
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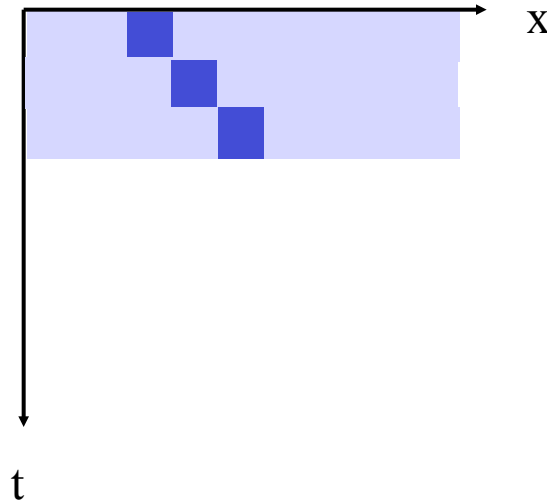
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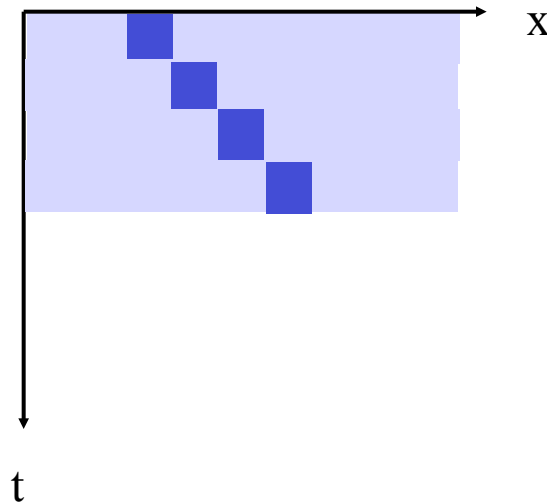
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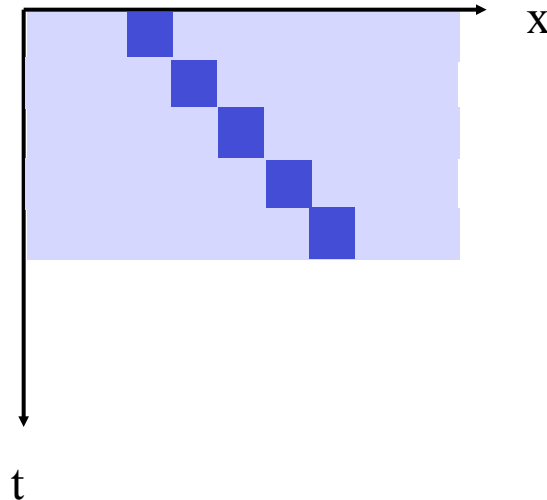
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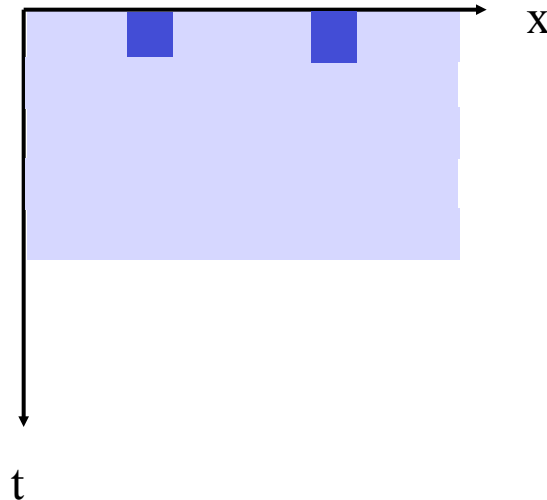
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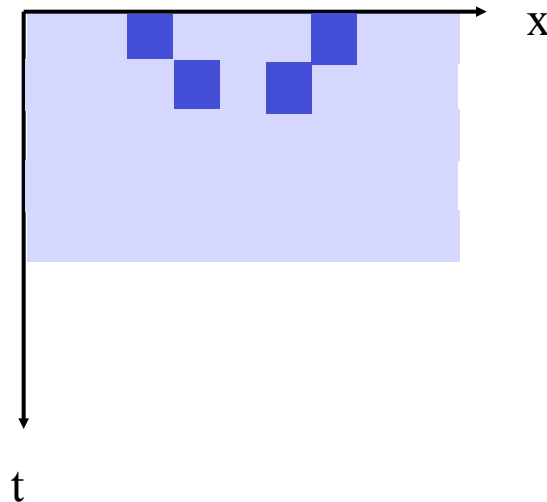
# A representation for spatiotemporal patterns: Primitive patterns

- Consider a spatiotemporal slice
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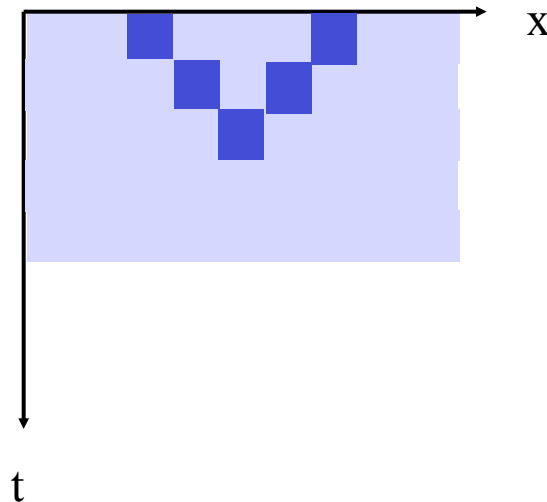
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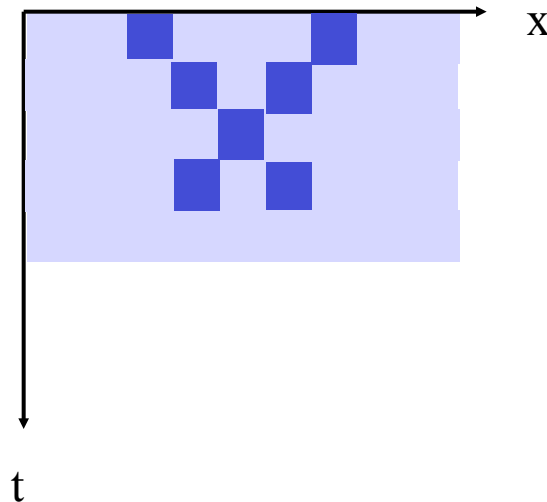
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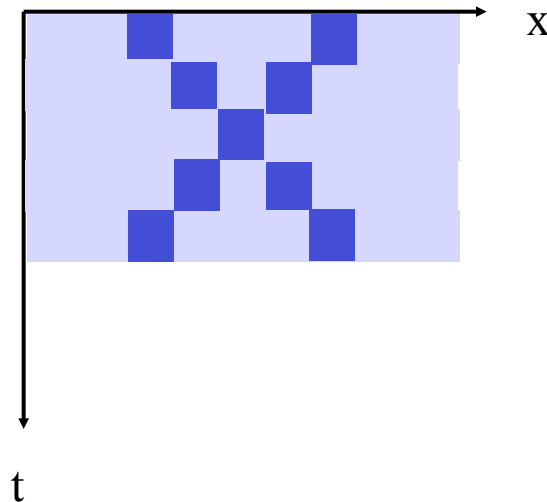
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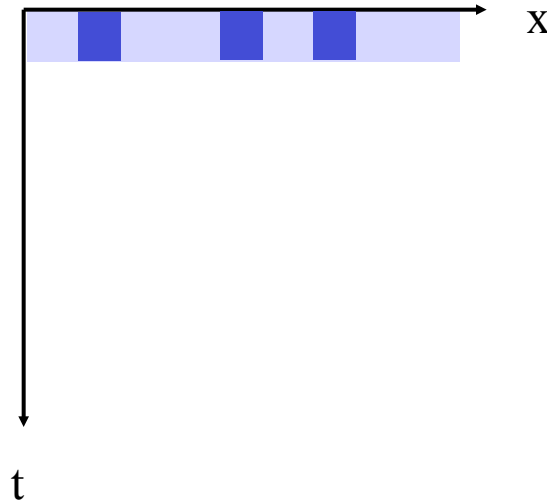
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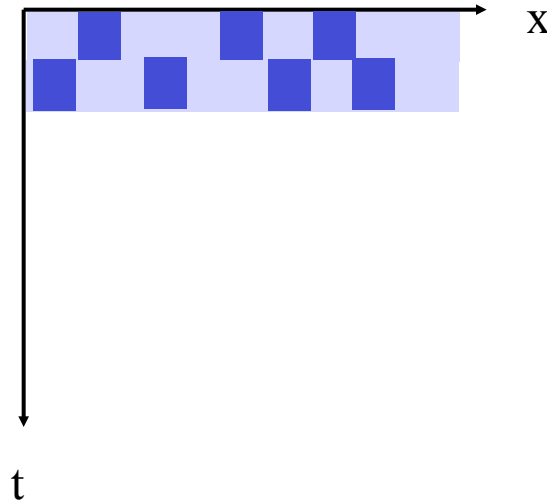
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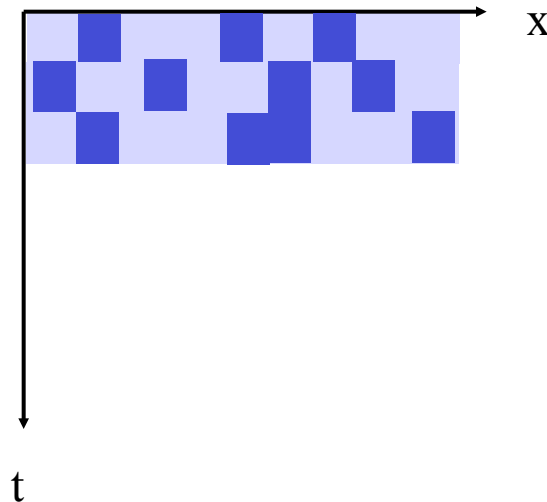
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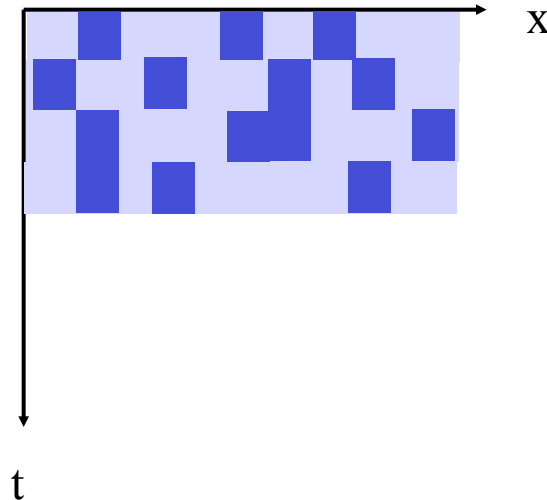
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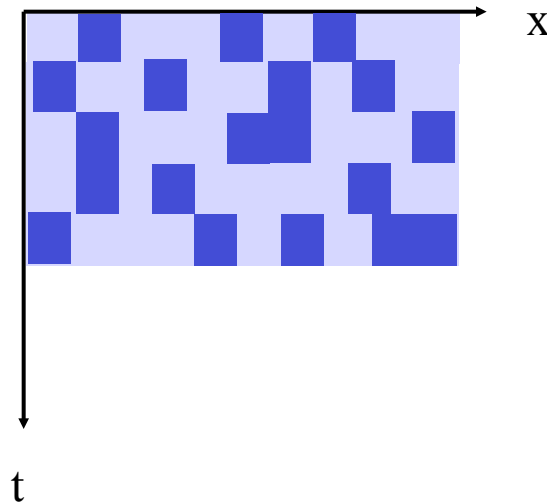
# A representation for spatiotemporal patterns: Primitive patterns

- Consider a spatiotemporal slice
  - As an observer views a pattern, “randomly” structured in space and time.









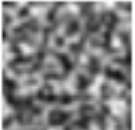
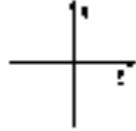



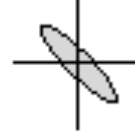
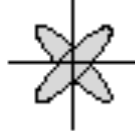
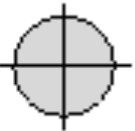
# A representation for spatiotemporal patterns: Primitive patterns

- Consider a spatiotemporal slice
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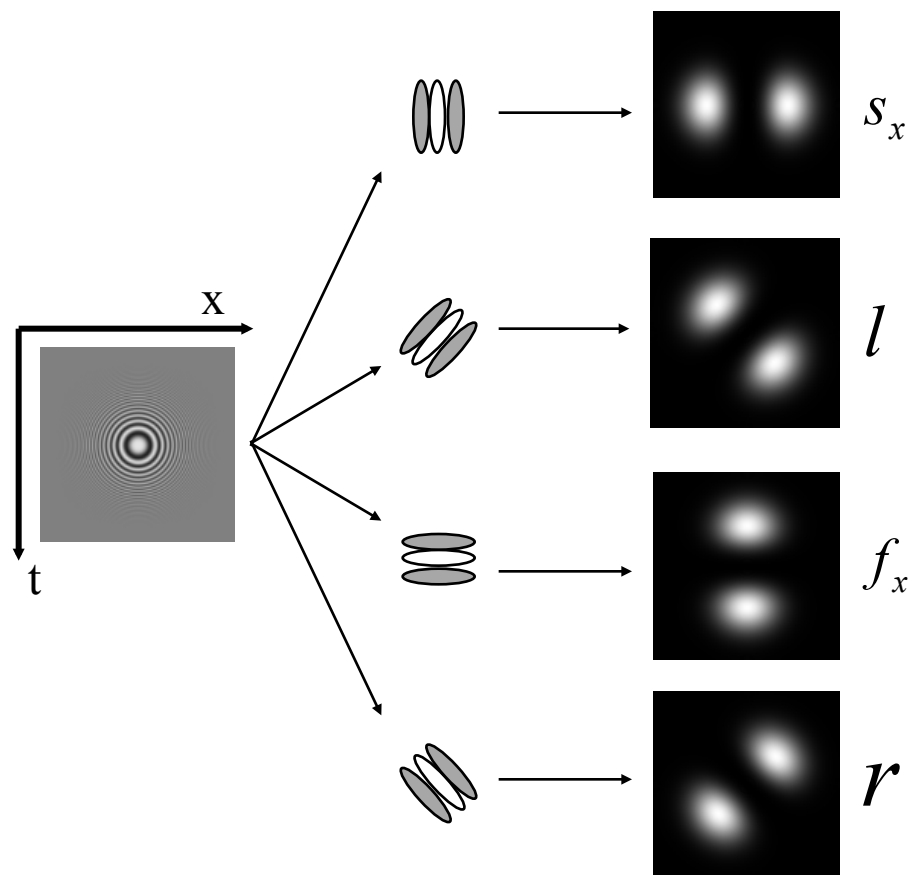


# A representation for spatiotemporal patterns: Primitives

	Unstructured	Static	Flicker	Coherent Motion	Incoherent Motion	Scintillation
						
						

# Orientation in visual space-time: Filtering

- Apply filters tuned to 4 different orientations in both x-t and y-t domains.
- In general, might consider additional directions.
- Filter specifics:
  - Oriented bandpass filters in spatiotemporal slice.
  - Lowpass filter in orthogonal spatial dimension.
  - Pointwise squared to yield local “oriented energy”.



# Orientation in visual space-time: Normalization

- For any given orientation, the filter response is a joint function of
  - orientation
  - contrast
- Normalization yields purer measure of orientation

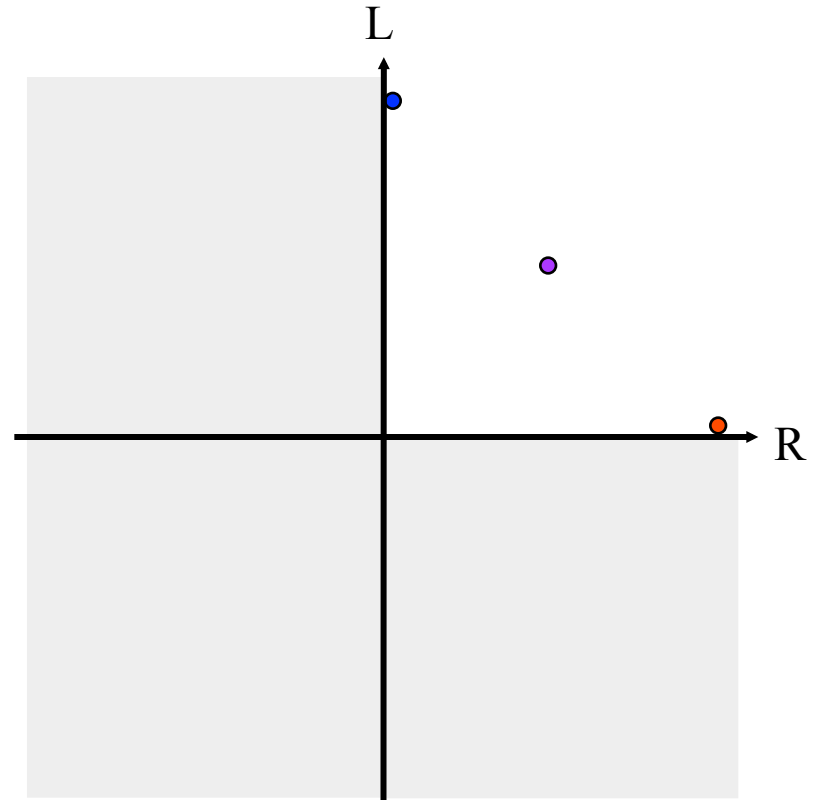
$$R(x, y, t) = \frac{r(x, y, t)}{r(x, y, t) + l(x, y, t) + s_x(x, y, t) + f_x(x, y, t) + \varepsilon}$$

with  $\varepsilon$  a small bias added for stability.

- Similarly for  $l$ ,  $s_x$ ,  $f_x$  and their y-t counterparts.

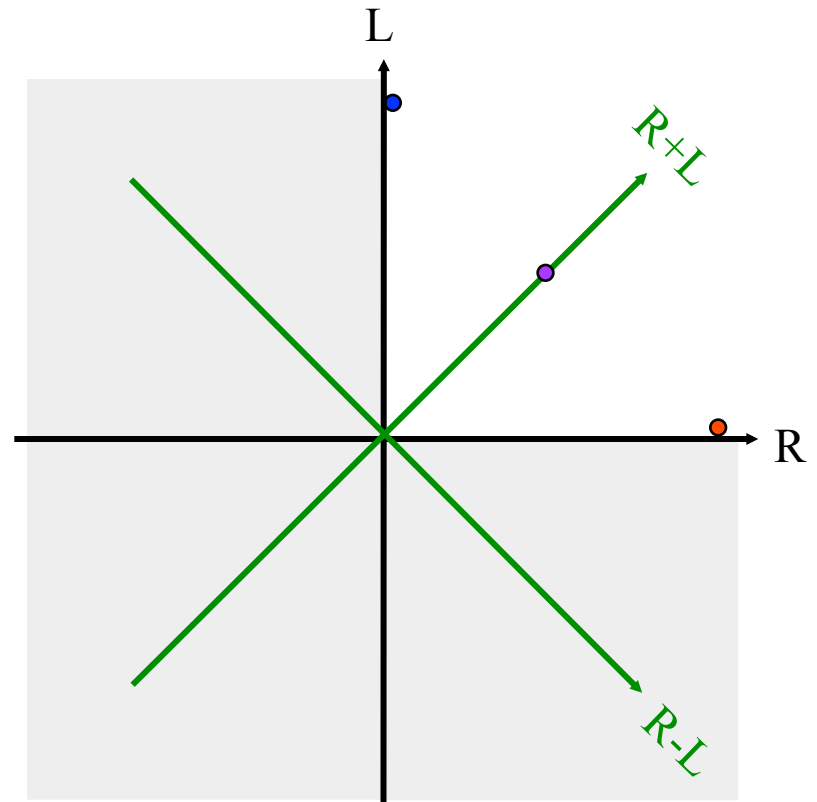
# A representation for spatiotemporal patterns: Opponency and summation

- The R and L (U and D) components are ambiguous WRT coherent and incoherent motion.



# A representation for spatiotemporal patterns: Opponency and summation

- The R and L (U and D) components are ambiguous WRT coherent and incoherent motion.
- Solution: Combine via
  - opponency  $R-L$  (U-D)
  - summation  $R+L$  (U+D)
- Geometrically a rotation of coordinate axes.

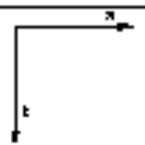




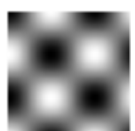

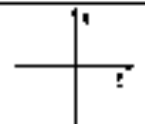

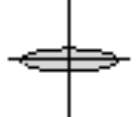

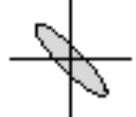
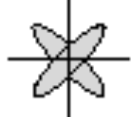
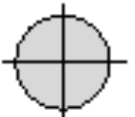


# A representation for spatiotemporal patterns: Spatiotemporal representation

- Proposal: A four band representation for both the x-t and y-t dimensions.

<u>x-t</u>	<u>y-t</u>
$ R - L $	$ U - D $
$R + L$	$U + D$
$S_x$	$S_y$
$F_x$	$F_y$

# A representation for spatiotemporal patterns: Primitives projected on representation

	Unstructured	Static	Flicker	Coherent Motion	Incoherent Motion	Scintillation
						
						
$ R - L $	0	0	0	++	0	0
$R + L$	0	++	++	++	+++	++
$S_{2D}$	0	++	0	+	+	+
$F_{2D}$	0	0	++	+	+	+

# A representation for spatiotemporal patterns: Examples I

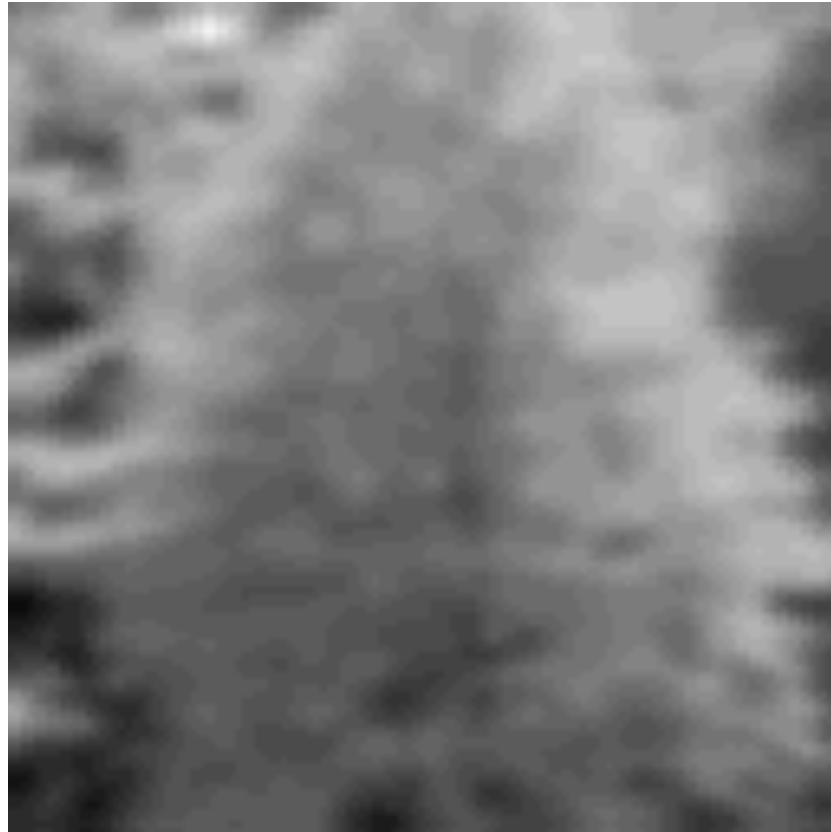
- A natural image sequence of each proposed class was acquired,  $(x,y,t) = (64,64,40)$ .
  - Unstructured: featureless sky
  - static: motionless tree
  - flicker: smooth surface illuminated by lightning flashes
  - coherent motion: field of flowers under camera motion
  - incoherent motion: overlapped legs in complex motion
  - scintillation: rain striking a puddle
- Each sequence brought under proposed representation.



# Unstructured



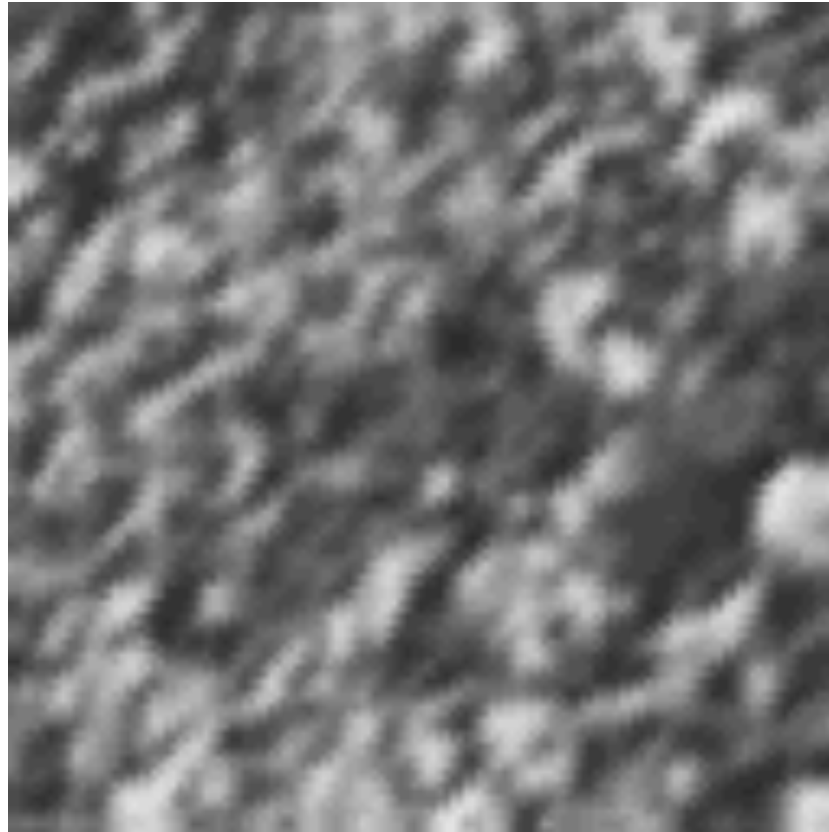
# Static



# Flicker



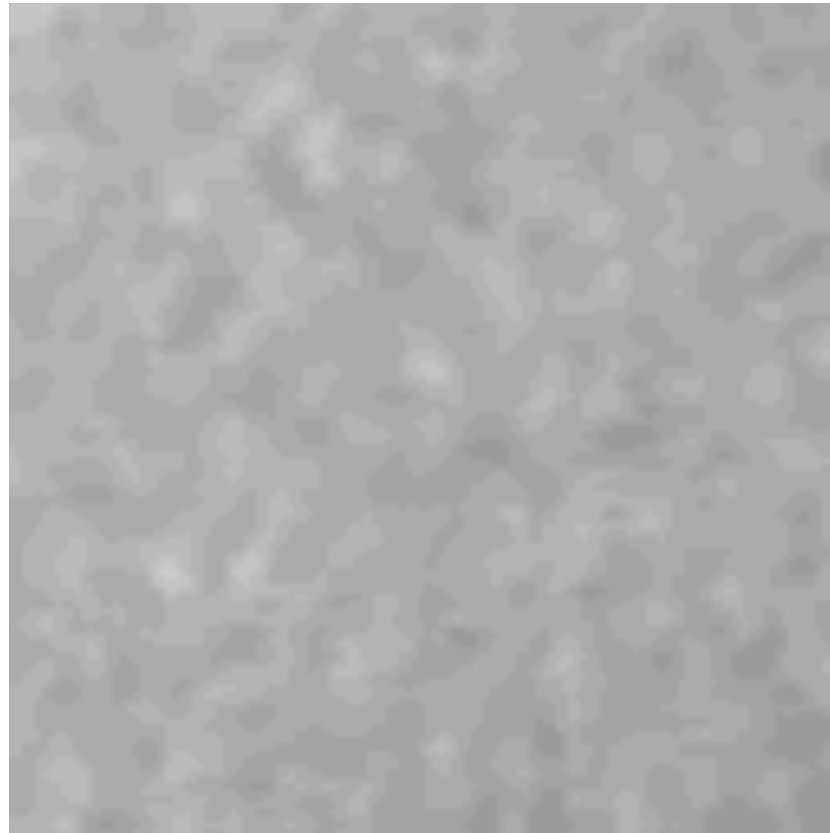
# Coherent motion



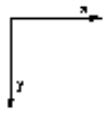



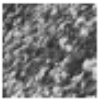

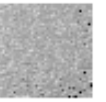
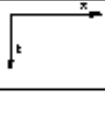

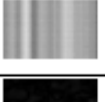
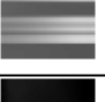
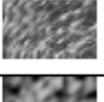



















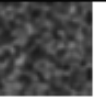

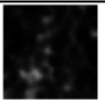

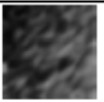

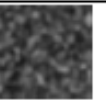
# Incoherent motion



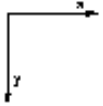



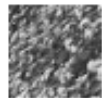


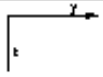











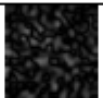











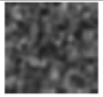

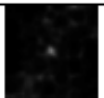

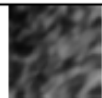

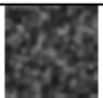
# Scintillation



# A representation for spatiotemporal patterns: Results x-t

	Unstructured	Static	Flicker	Coherent Motion	Incoherent Motion	Scintillation
						
						
$ R - L $	 0.00	 0.00	 0.00	 0.37	 0.05	 0.02
$R + L$	 0.01	 0.40	 0.36	 0.53	 0.58	 0.50
$S_2$	 0.00	 0.55	 0.00	 0.21	 0.17	 0.25
$F_2$	 0.00	 0.04	 0.63	 0.26	 0.25	 0.23

# A representation for spatiotemporal patterns: Results y-t

	Unstructured	Static	Flicker	Coherent Motion	Incoherent Motion	Scintillation
						
						
$ U - D $	 0.00	 0.00	 0.00	 0.34	 0.02	 0.02
$U + D$	 0.01	 0.38	 0.36	 0.52	 0.45	 0.50
$S_y$	 0.00	 0.59	 0.00	 0.19	 0.24	 0.28
$F_y$	 0.00	 0.03	 0.64	 0.29	 0.29	 0.21



# **A representation for spatiotemporal patterns: Remarks**

- Have described a representation for distinguishing primitive spatiotemporal patterns.
- The representation makes use of oriented bandpass image decomposition.
- Initial empirical results support the hypothesis that the proposed representation can afford the desired distinctions.

# Outline

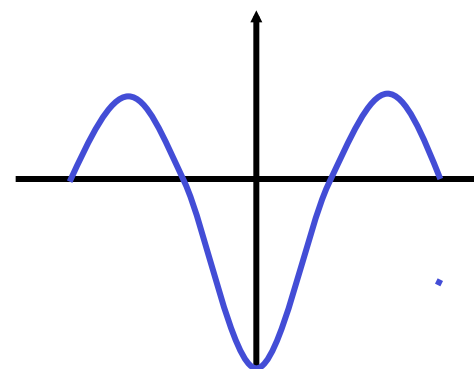
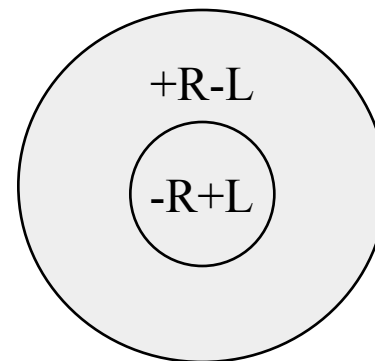
- Introduction
- Orientation in visual space-time
- A representation for spatiotemporal patterns
- **Spatiotemporal boundaries**
- A framework for spatiotemporal analysis
- Applications
- Summary

## Boundaries: Motivation

- Detection and localization of spatiotemporal boundaries is an important aspect of chunking information into meaningful pieces.
- Complimentary to the area-based analysis considered so far.
- Differential operators matched to the juxtaposition of contrasting spatiotemporal structure can be assembled from the primitive filter responses, S, F, R-L, R+L, etc.

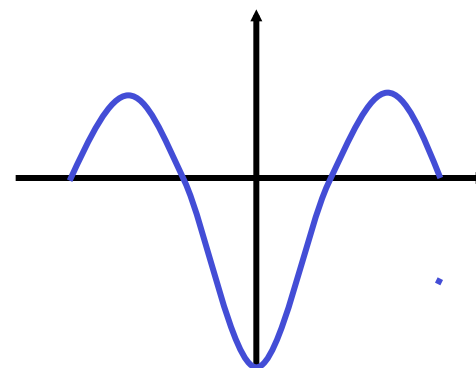
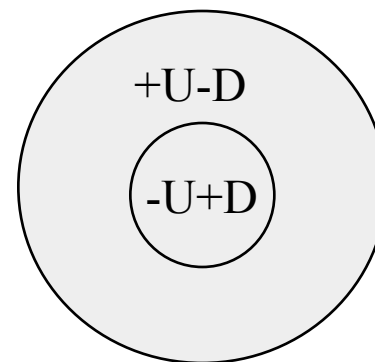
## Boundaries: Coherent motion

- Boundaries in coherent motion discriminate foreground/background.
- Coherent motion related to opponent bands R-L and U-D.
- Combine a spatial Laplacian with opponent filtering to yield double opponent operators.



## Boundaries: Coherent motion

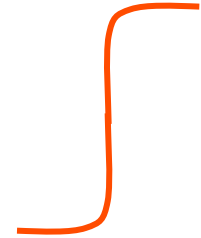
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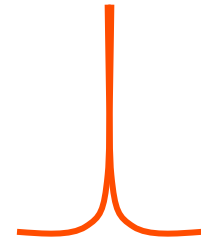
## Boundaries: Signature

- Zero-crossings in the double-opponent motion operator output indicate coherent motion boundaries.
- Slope magnitude taken as strength of boundary signal.
- Sum signals from x-t and y-t dimensions.

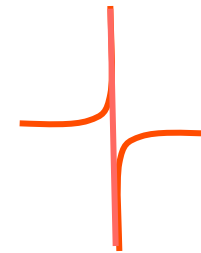
$$R - L$$



$$D_x(R - L)$$



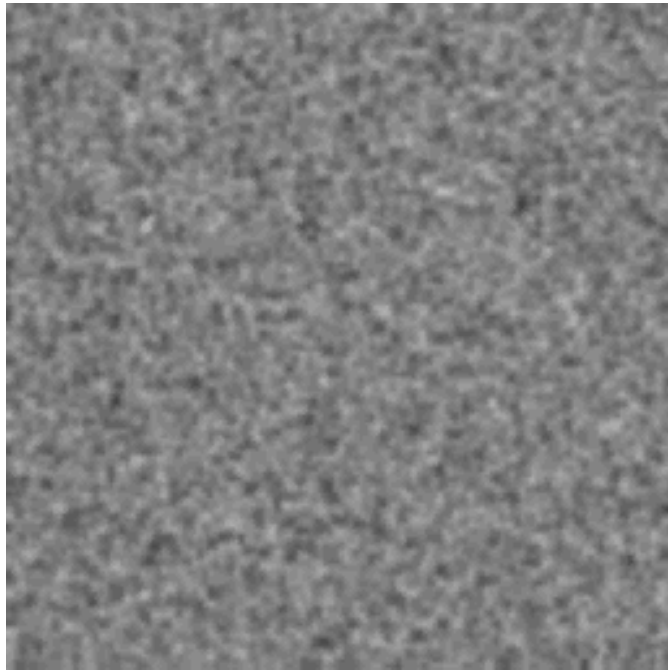
$$D_{xx}(R - L)$$



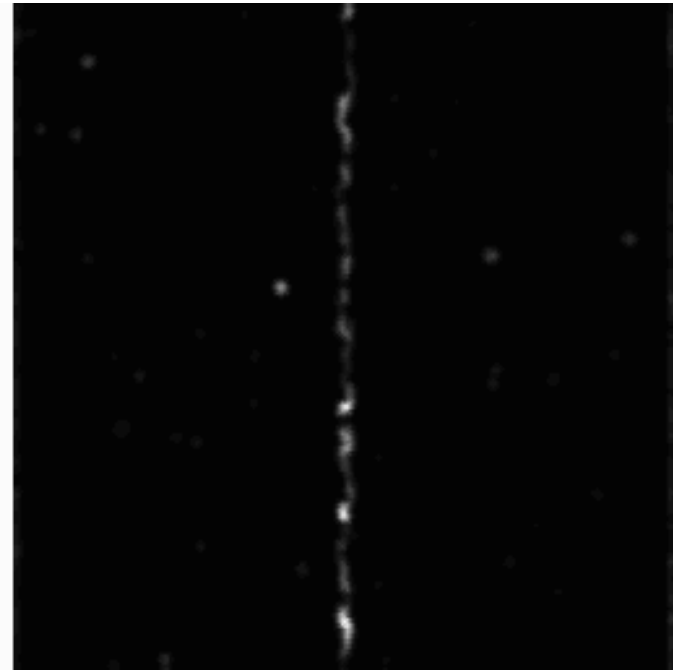
## Boundaries: Examples II

- Two image sequences depicting boundaries of coherent motion,  $(x,y,t) = (256,256,16)$ .
  - random dot cinematogram: left and right sides of display in opposite horizontal motion.
  - natural image sequence: aerial view of tree canopy with movement relative to undergrowth due to camera motion; homogeneous texture of vegetation obscures boundary in any one image.
- Each sequence processed by proposed method for indicating coherent motion boundaries.

## Boundaries: Results



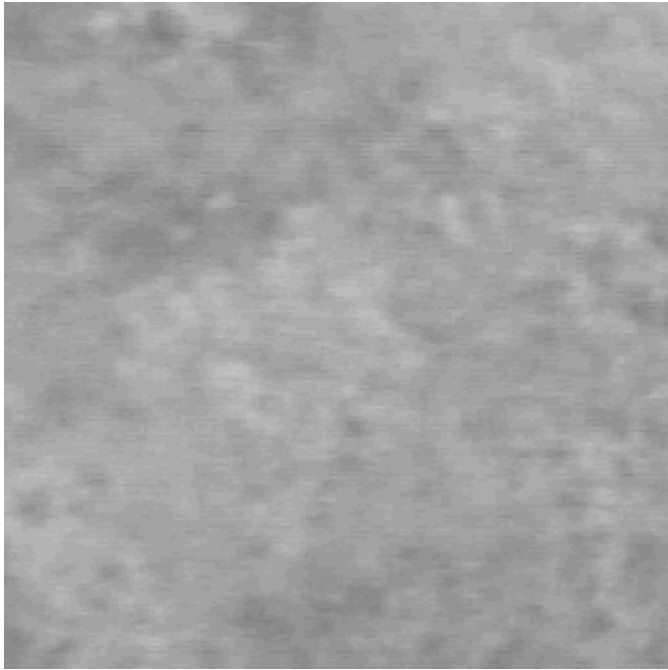
Frame of input sequence.



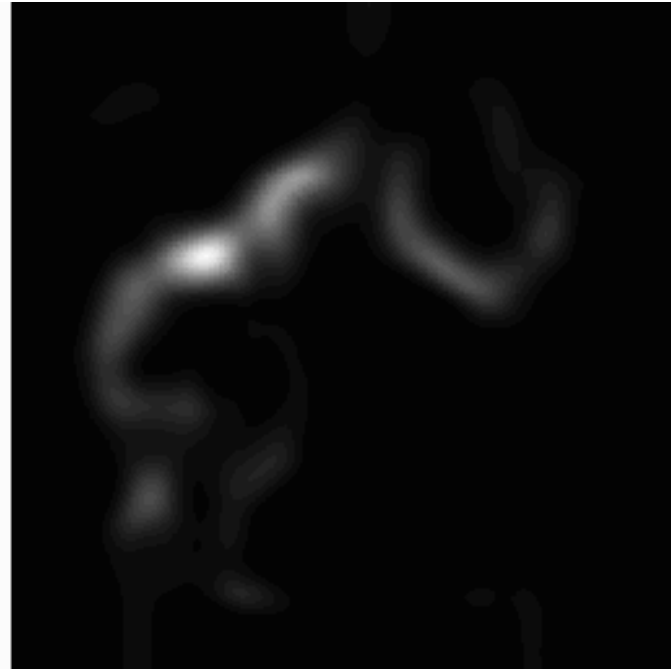
Motion boundary signal intensity.



## Boundaries: Results



Frame of input sequence.

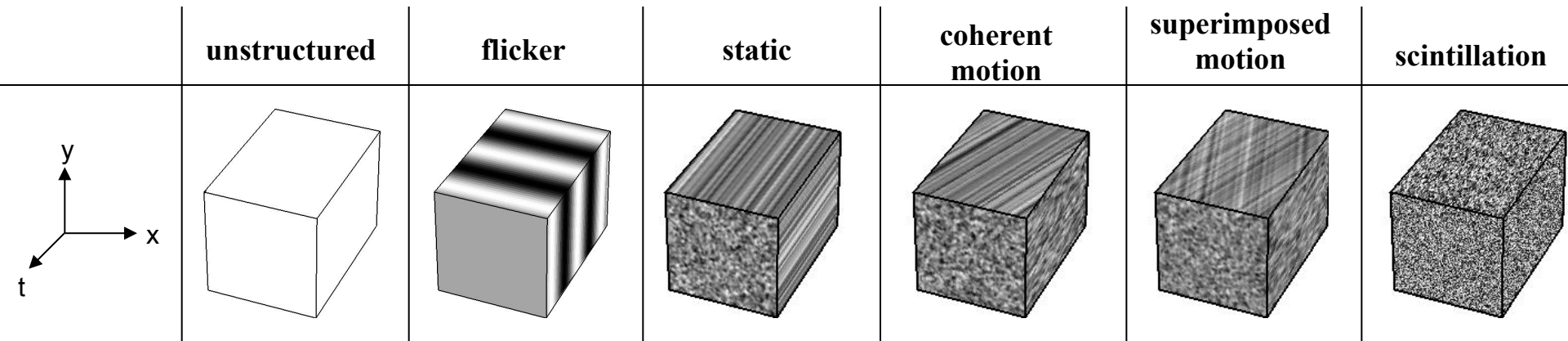


Motion boundary signal intensity.

# Outline

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# Framework: Local orientation



## Appeal to orientation is not arbitrary

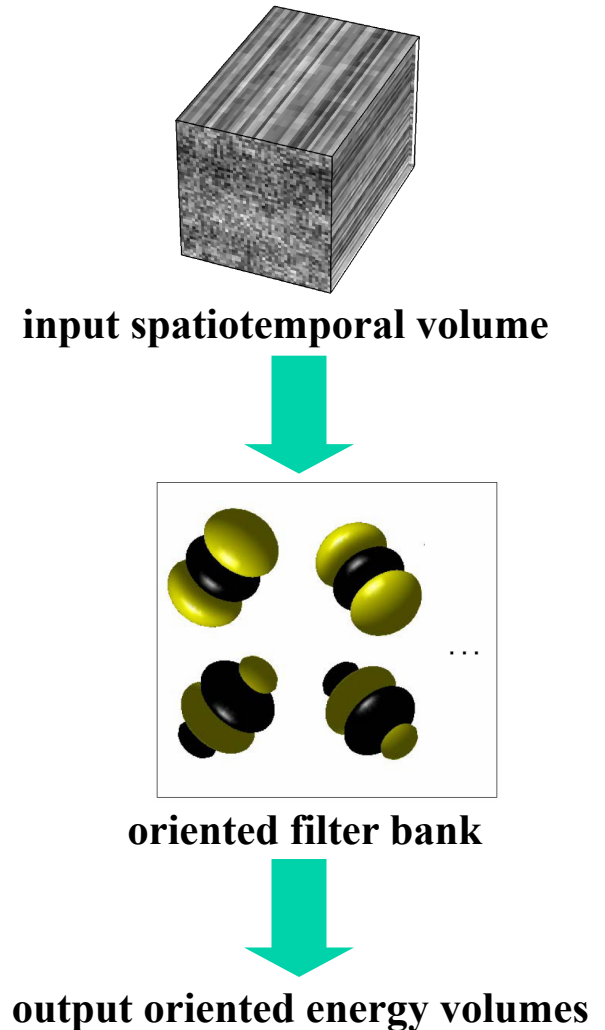
- The local orientation (or lack thereof) of a pattern is one of its most salient characteristics.
- Geometrically, orientation captures the local first-order correlation structure of a pattern, alternatively the local tangent.
- Provides a formal prerequisite to analysis of higher-contact constructions (e.g., curvature)
- For vision, local spatiotemporal orientation can have additional interpretations.
  - Image velocity is manifest as spatiotemporal orientation.
  - And more...

# Framework: Filtering

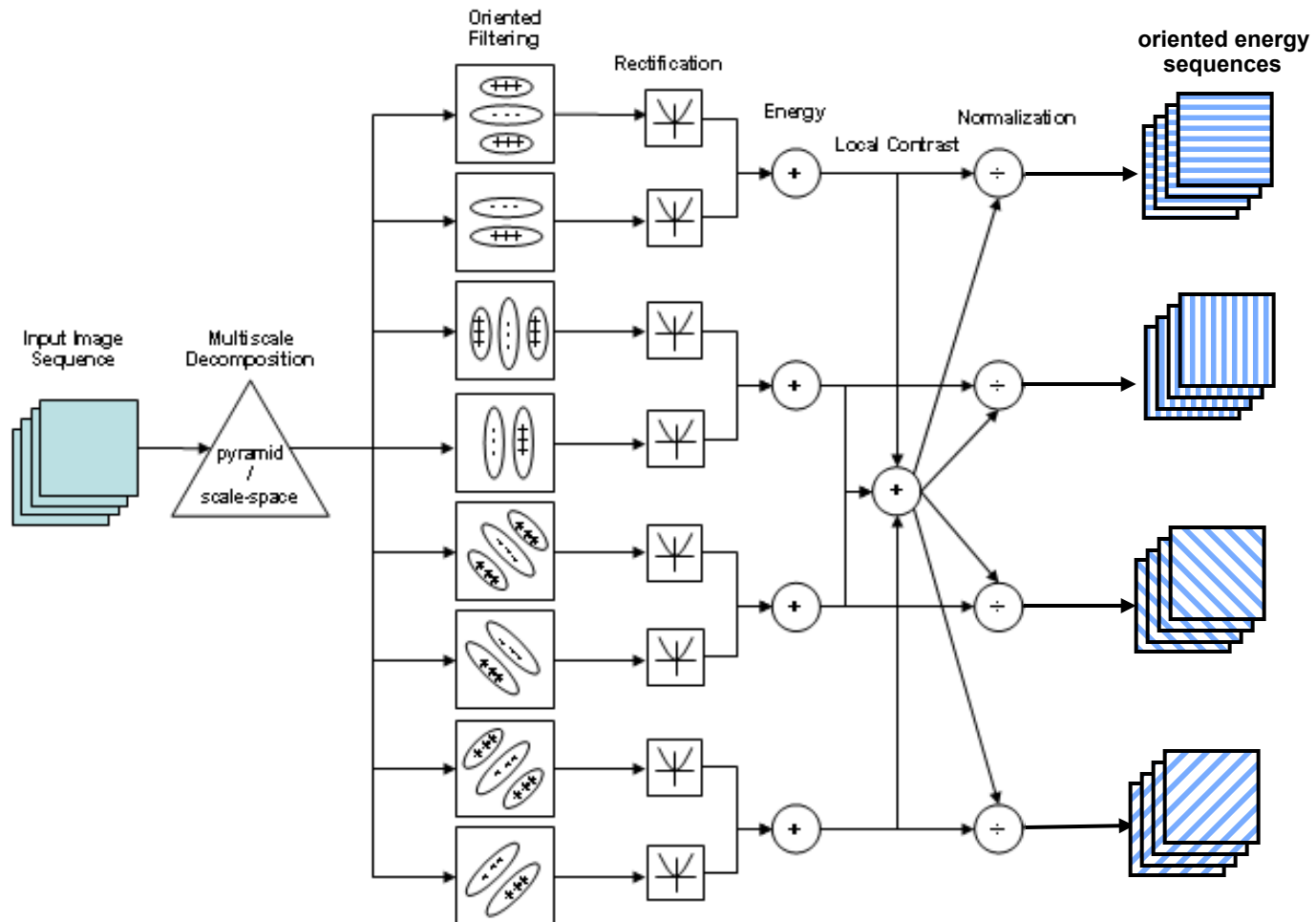
## Orientation selectivity

- Goal is to analyze spatiotemporal data,  $I(x, y, t)$ , according to its local orientation structure.
- Choose a representation with multiple bands each tuned for certain orientations,  $\theta$ , and scales in 3D  $(x, y, t)$ .
- Filter specifics:
  - 3D Gaussian second derivatives,  $G_{2\theta}$ .
  - Corresponding Hilbert transforms,  $H_{2\theta}$ .
  - Rectified and summed in quadrature pairs to yield local “oriented energy”.

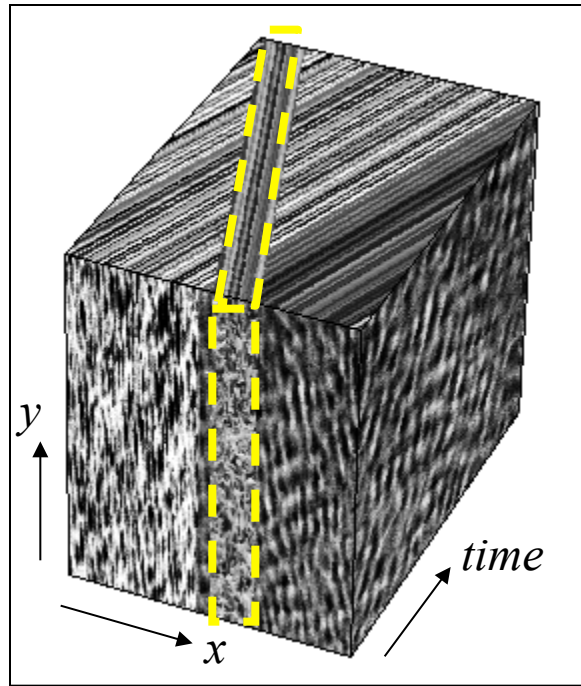
$$E_{\theta}(x, y, t) = \left[ G_{2\theta} * I(x, y, t) \right]^2 + \left[ H_{2\theta} * I(x, y, t) \right]^2$$



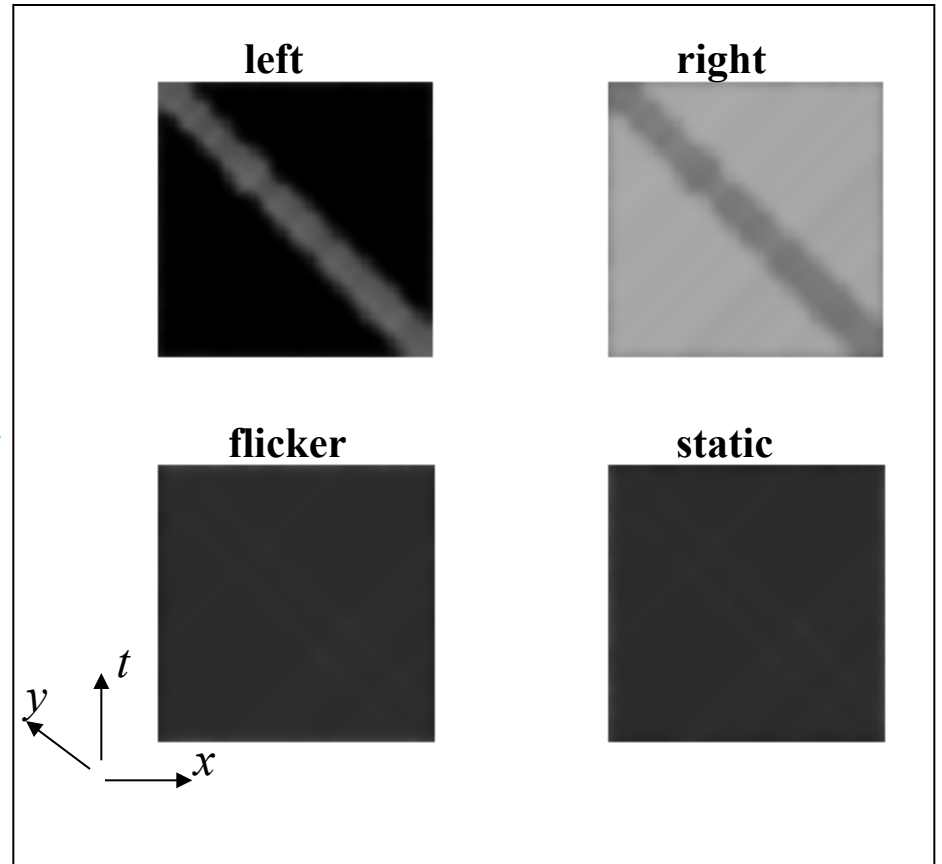
# Framework: Architecture



# Framework: Example



**spatiotemporal volume**

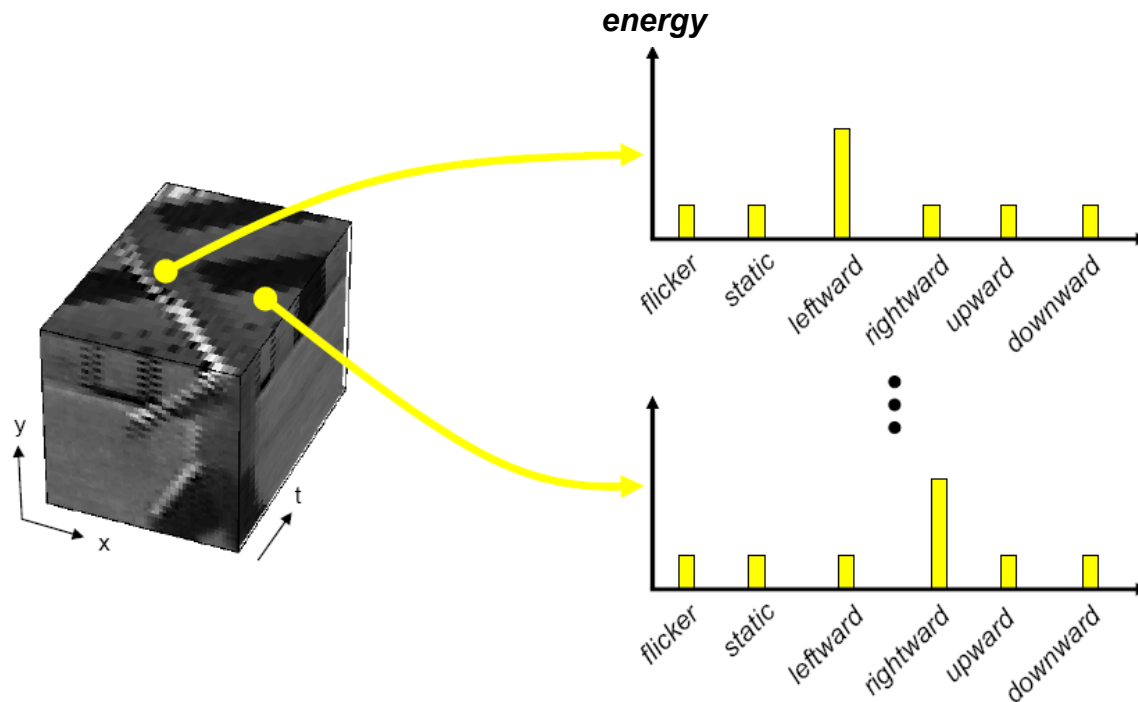


**oriented energy volume**

# Framework: Distributions of oriented energy

## Key ideas

- Build a distributed representation (i.e., histogram) at each point that measures the amount of energy for various kinds of spacetime oriented structures.
- Base subsequent analysis on the distribution of oriented energies across space and time.



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# Applications

Examples provided in lecture.

# Summary

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