

Life Patterns

structure from wearable sensors

by

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The Question: Can we computationally structure a sensor record of an individual's day-to-day life?

sensor record - “dumb” measurements via sensors

structure - similarity measure, perplexity, prediction, classification

The Applications:

memory prosthesis

automatic diary

the frame problem

“the past explains”

context-aware agents

# Prologue

## Background

Vannevar Bush - memex (1945)

### Wearable Sensing

Steve Mann - wearable cameras (1997)

Thad Starner - Patrol system (1999)

Farrington et. al. - sensory badges & jackets (1999)

Jennifer Healey - wearable bio-sensing (1998)

### Context-Awareness

Brad Rhodes - Remembrance Agent (2000)

Lamming & Flynn - “Forget-me-not” (1994)

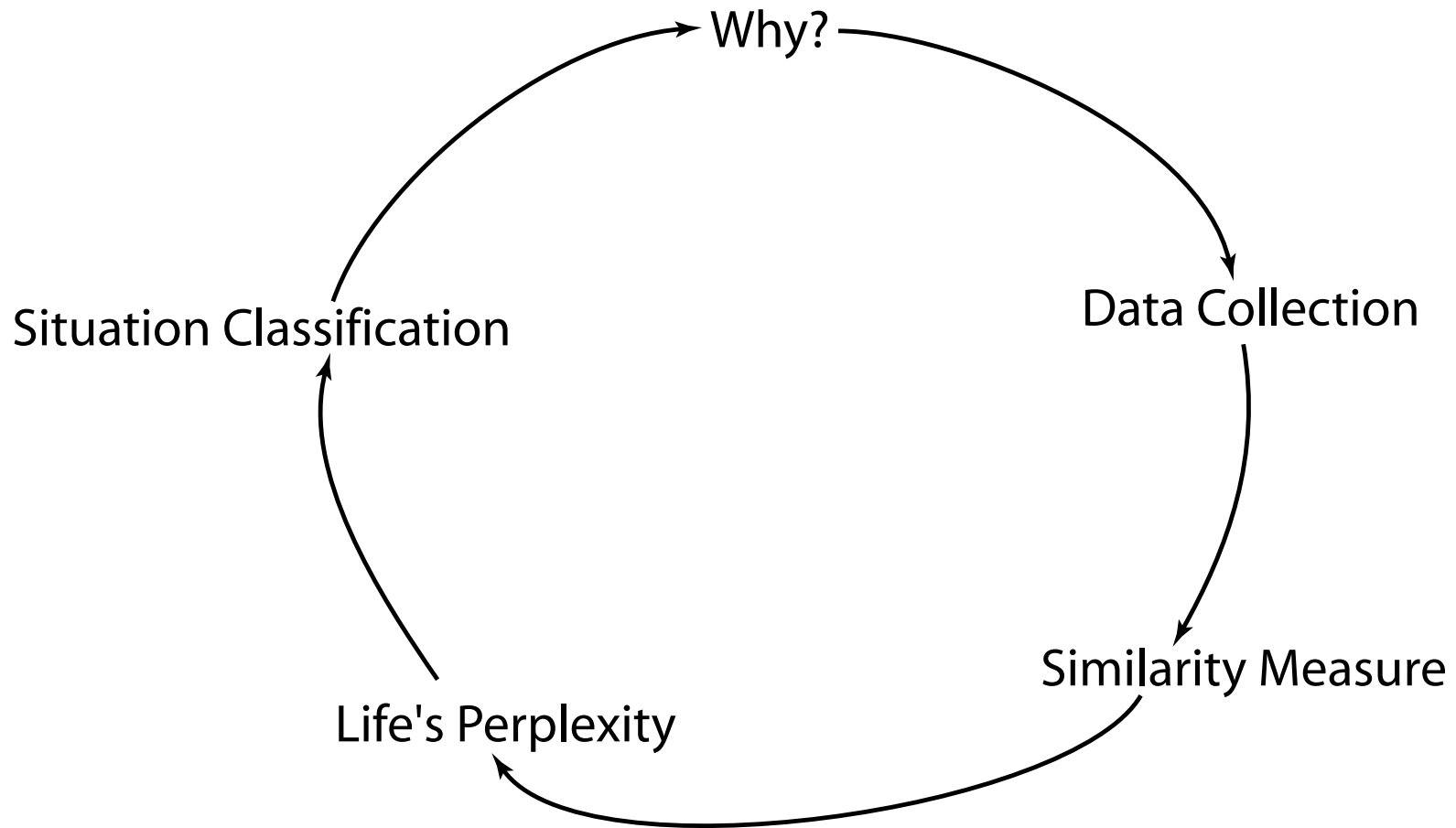
### Robotics & AI

Grimson et. al. - long-time monitoring of a site (1998)

Jogan & Leonardis - localization via panoramic views

Prologue

Talk Outline





# The “I Sensed” Data Set

Mr. Kawara On [1933- ]



Date paintings

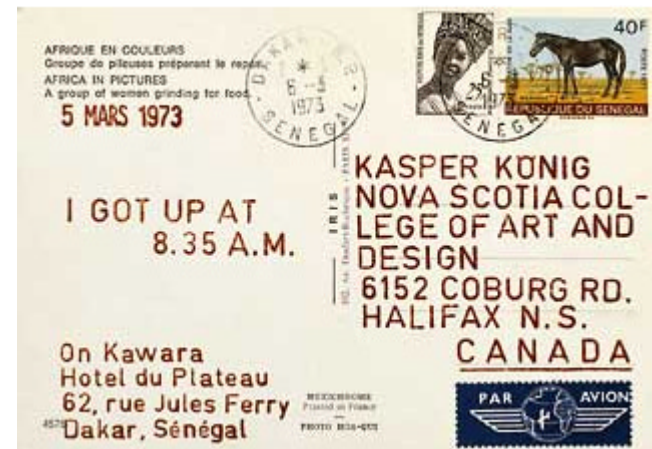
# The “I Sensed” Data Set

Mr. Kawara On [1933- ]

## The “I Got Up” Series



I Went  
I Met  
I Am Alive  
I Read ...



# The “I Sensed” Data Set

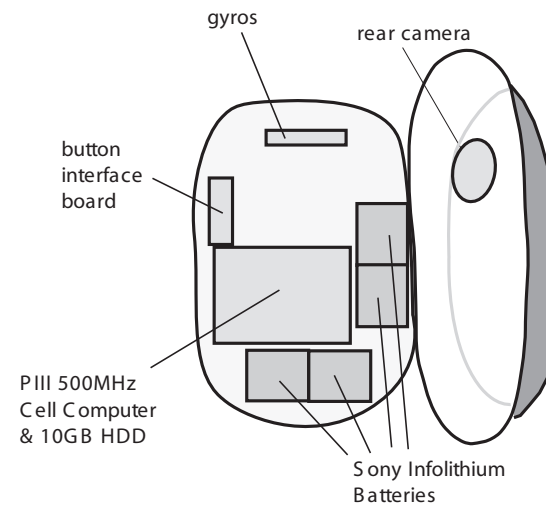
## Data Collection Wearable





# The “I Sensed” Data Set

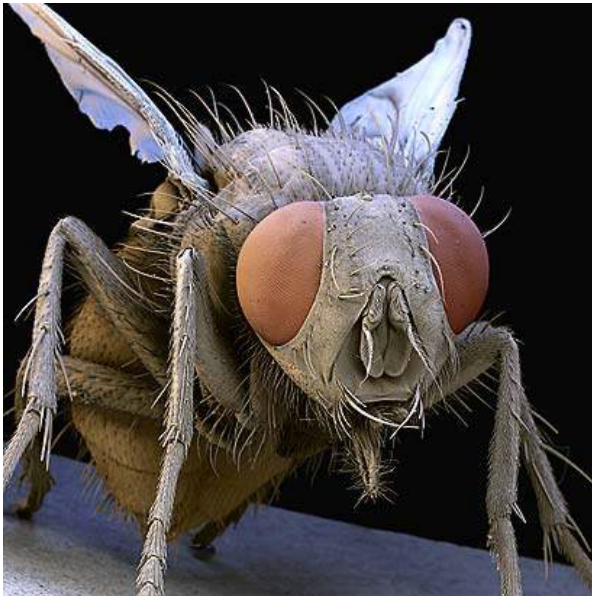
## Data Collection Wearable



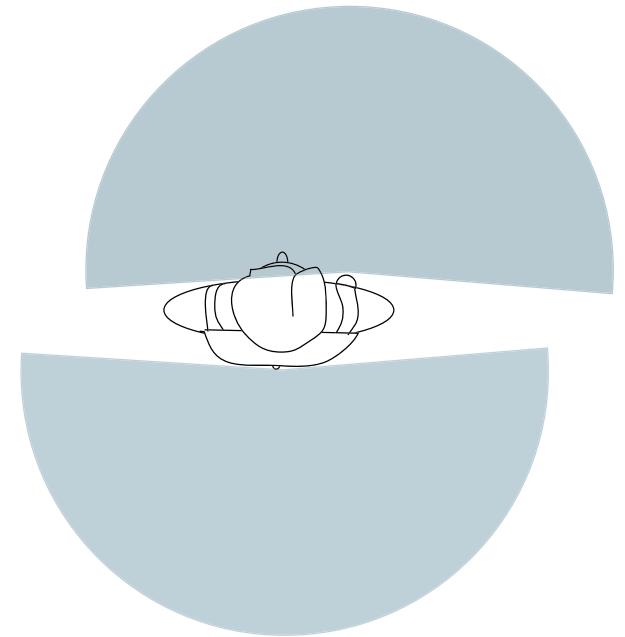
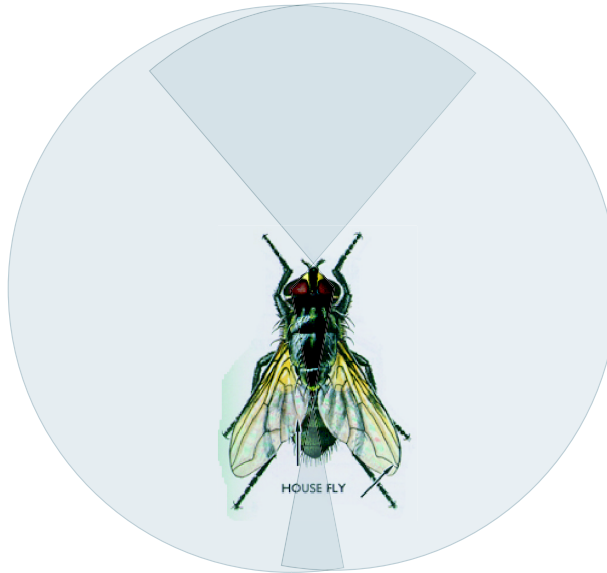
# The “I Sensed” Data Set

## Insect Perception

Key Properties: low resolution, wide field-of-view



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compared to humans:

- 100,000 times fewer photoreceptors
- 360-deg field of view
- 800 eye units, each having 8 photoreceptors

# The Similarity Measure

## Peripheral Perception

### Peripheral vs. Attentive



Walking over a bridge



Shopping at BestBuy

Without target of attention



Renting a video



Working at the desk

With target of attention

Peripheral sensing is robust to small changes in the environment.



Walking over a bridge



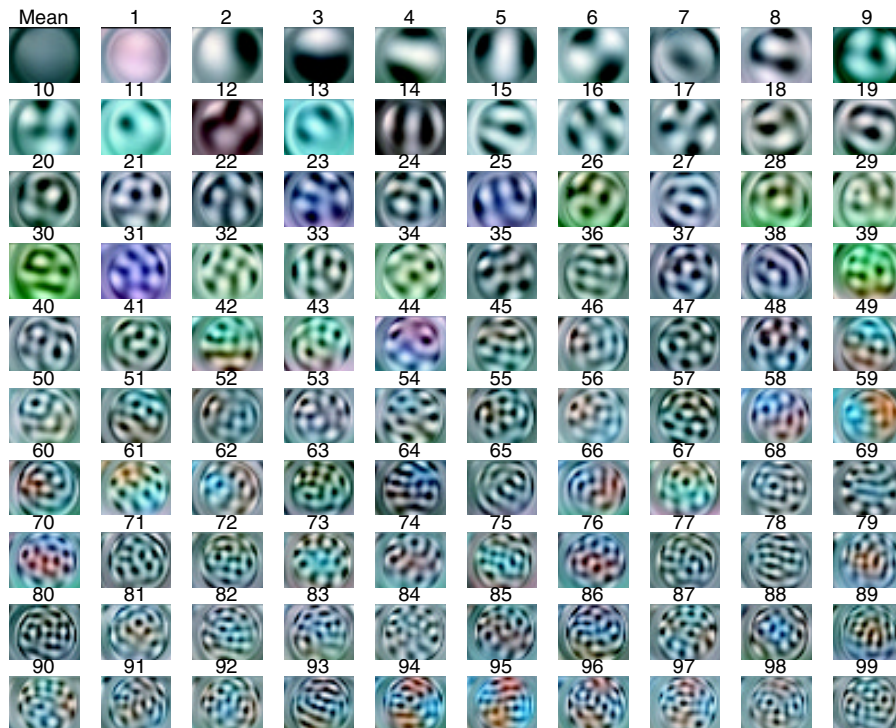
Walking over a bridge

Direct image matching without correspondence is potentially useful!

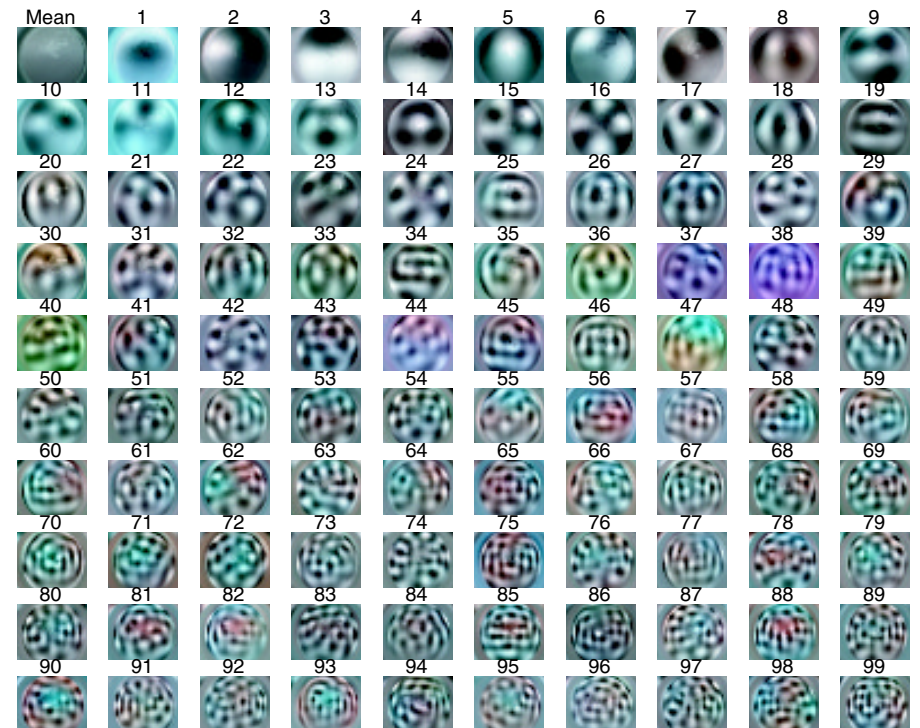
# The Similarity Measure

## Principle Components

Front View Eigenvectors



Rear View Eigenvectors



Trained on 32x24 pixel images over 30 days of video.

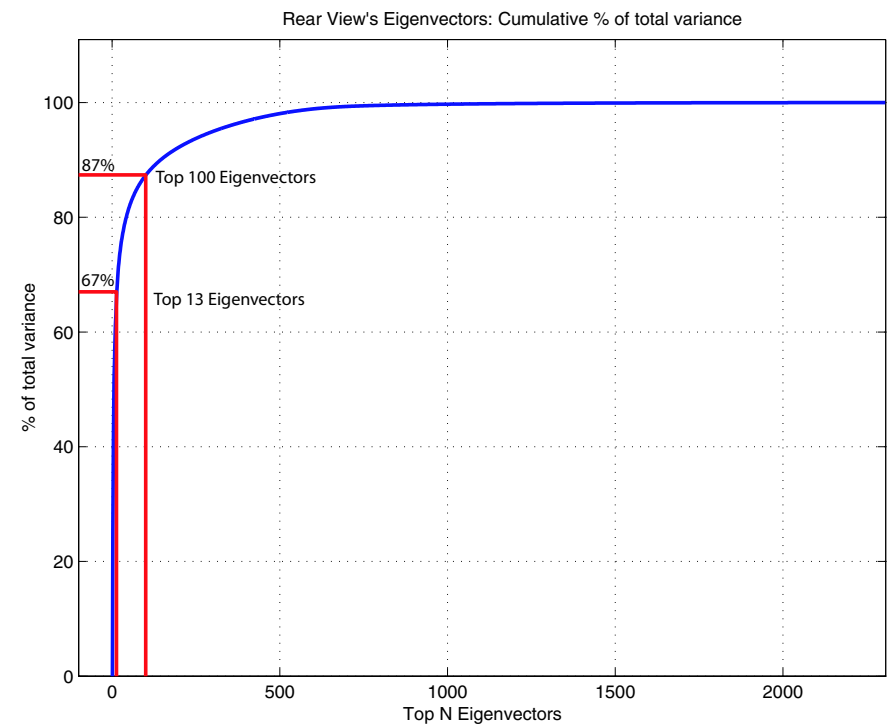
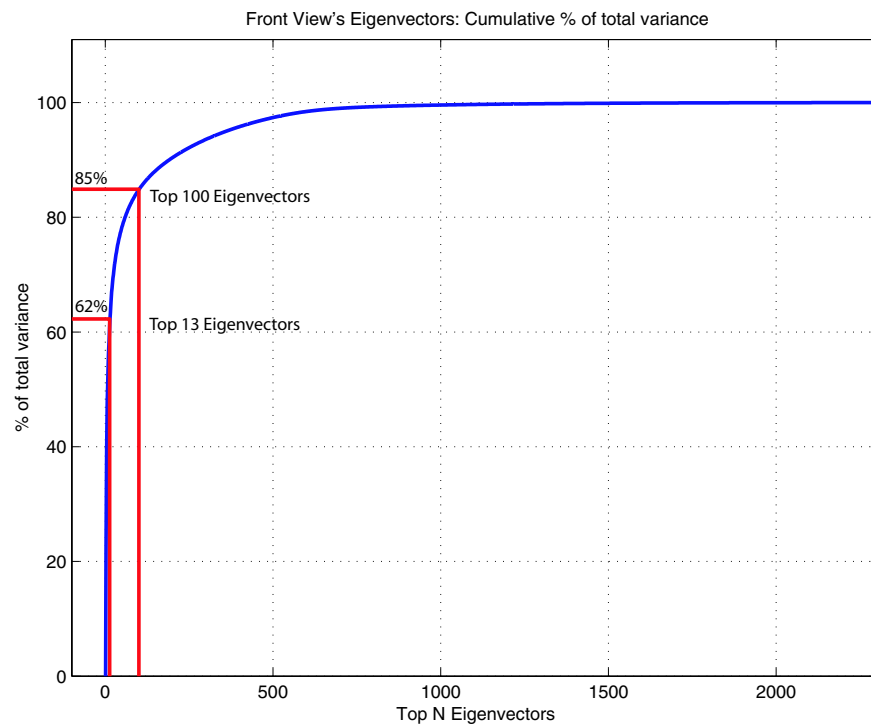
Bell & Sejnowski - PCA of natural scenes yields Fourier basis

ICA of natural scenes yields localized edge filters



# The Similarity Measure

## Variance Accounted For

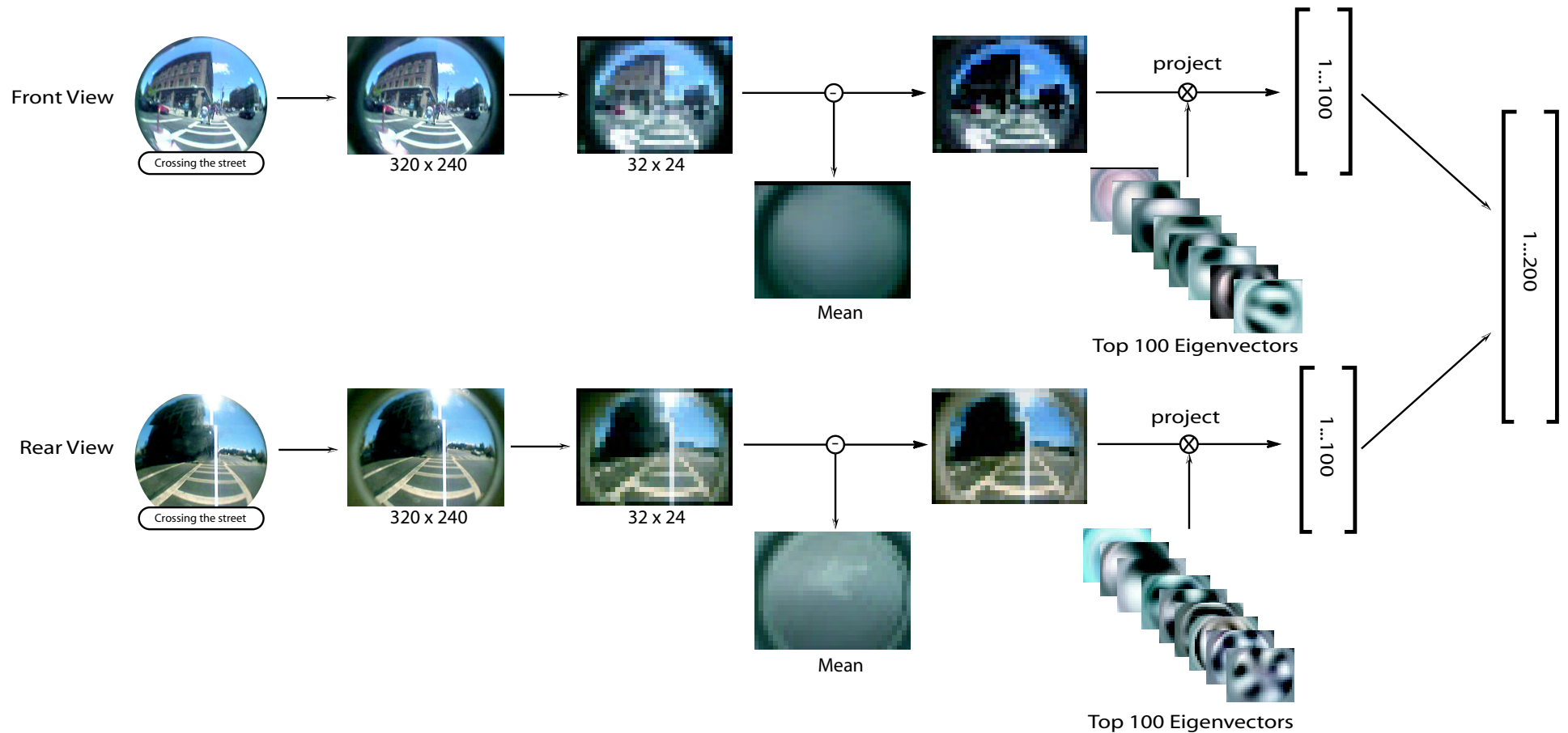


- Compression without loss of detail is difficult.
- We use the top 100 eigenimages for the remaining experiments.



# The Similarity Measure

## Feature Computation Pipeline

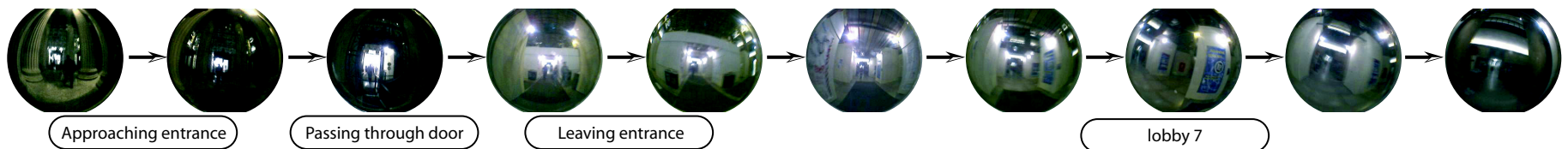
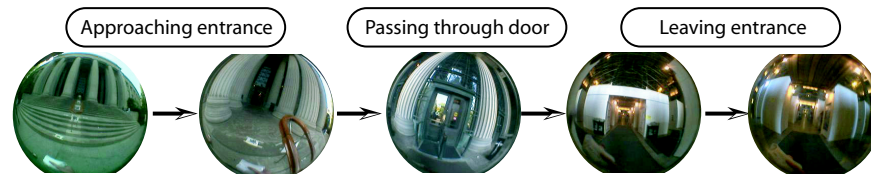


- The result is a 200-dimensional feature vector per frame
- To compare frames, we use the L-1 norm.

# The Similarity Measure

## Sequence Similarity

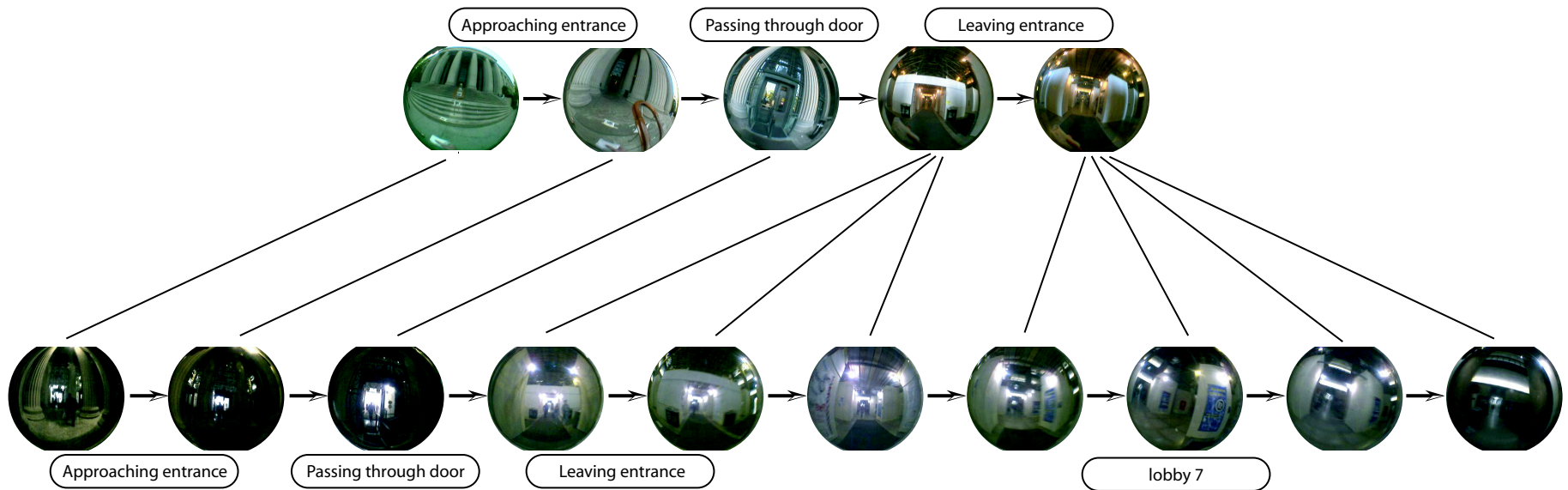
How do we compare a pair of frames sequences?



# The Similarity Measure

## Sequence Similarity

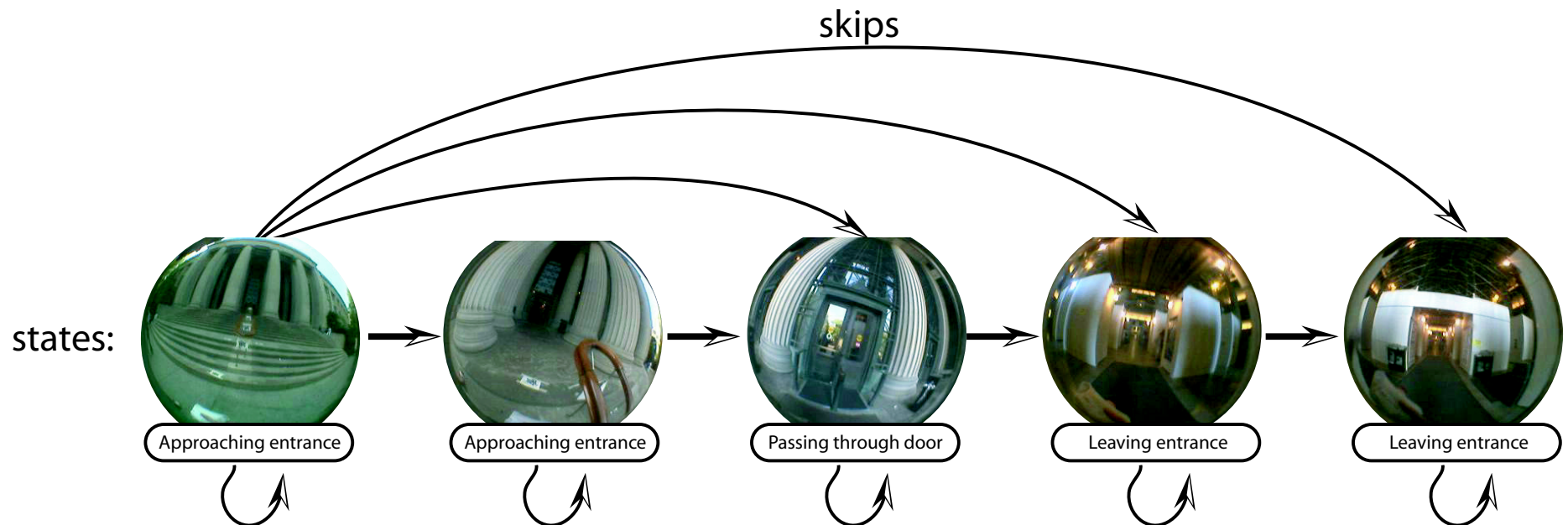
Match one sequence to the other and accumulate frame-by-frame similarities.



# The Similarity Measure

## Alignment Model

A sequence of frames becomes a Hidden Markov Model.

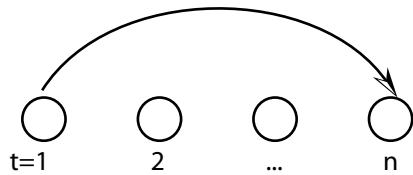


the transitions...

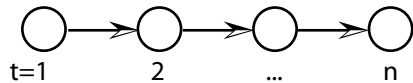
# The Similarity Measure

## Alignment Model

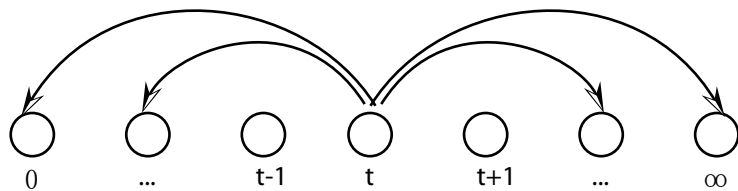
### $\alpha$ -transitions



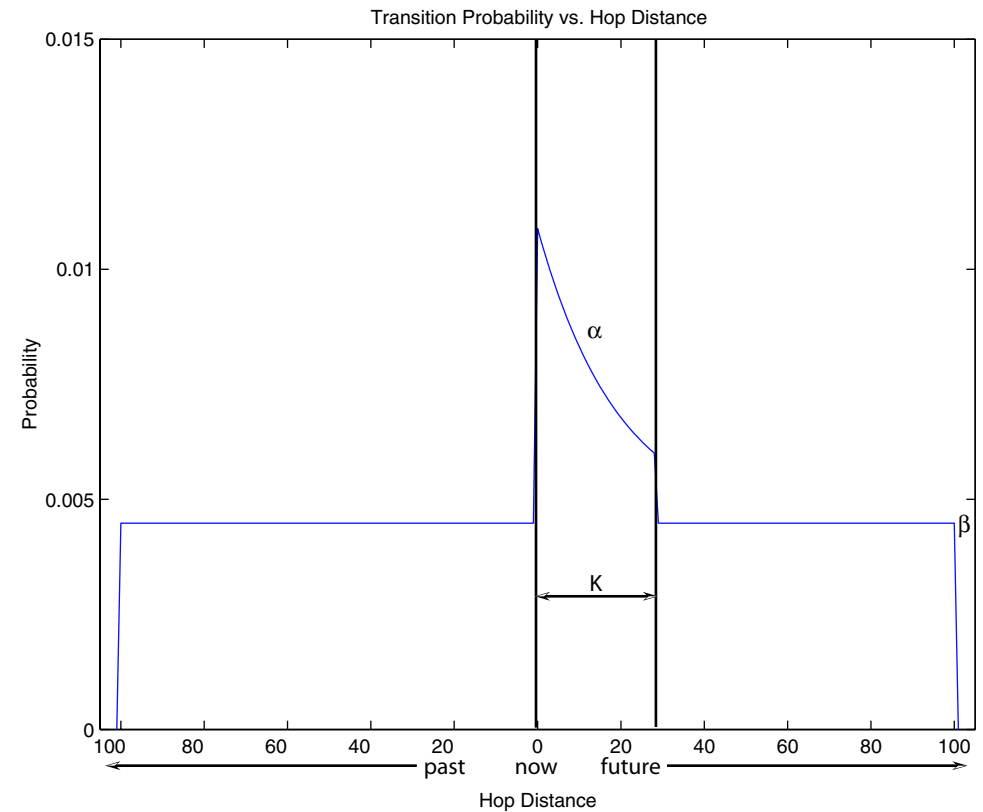
equal probability



### $\beta$ -transitions



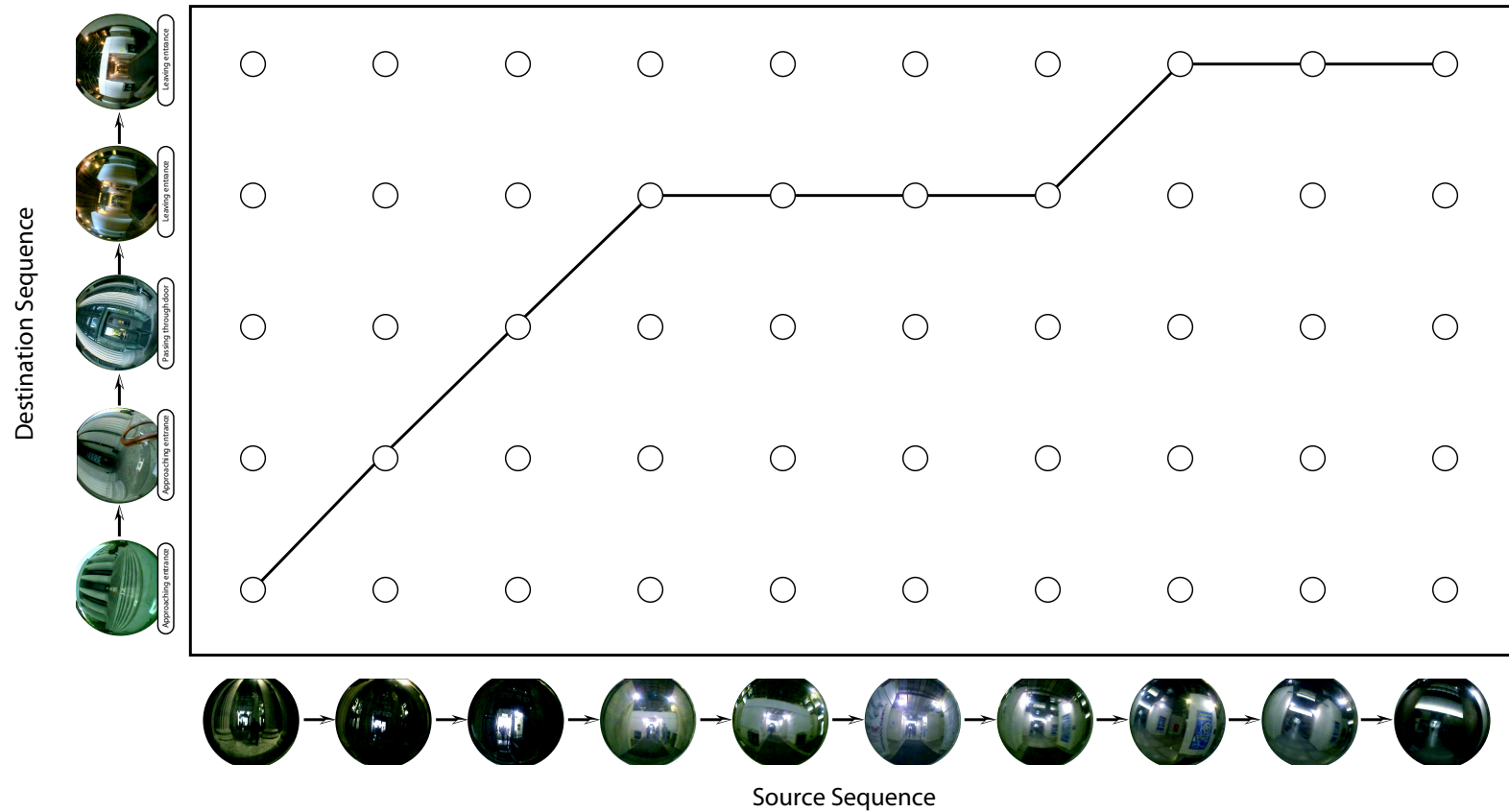
low constant probability



# The Similarity Measure

## Alignment Path

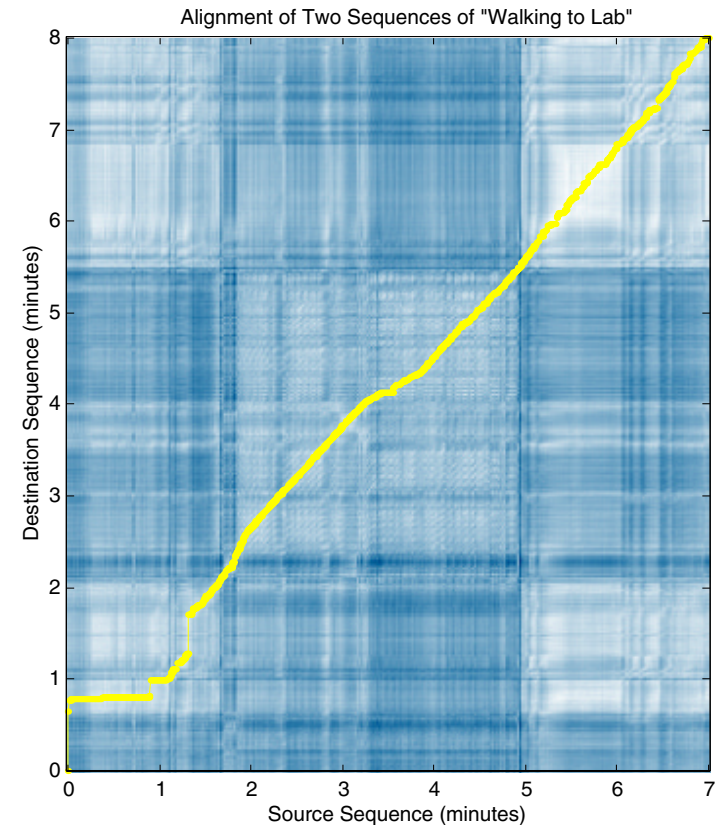
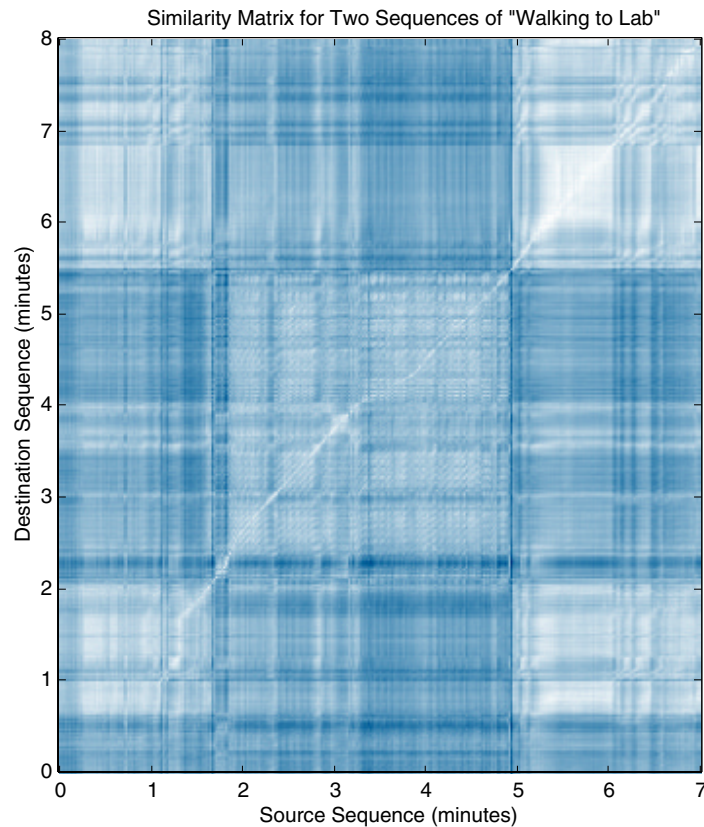
The Viterbi Algorithm produces the best possible alignment.



# The Similarity Measure

## An Example: walking to lab

these are very similar events...



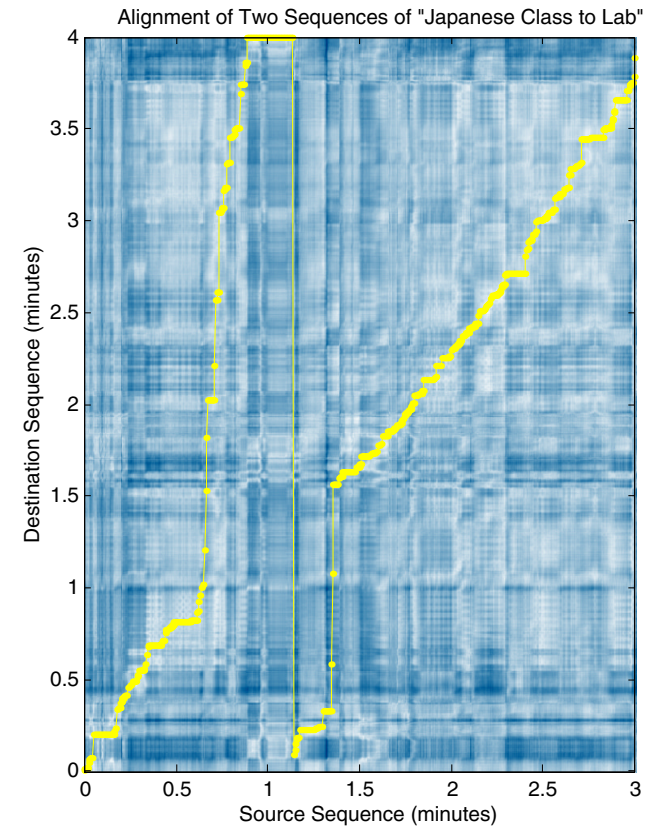
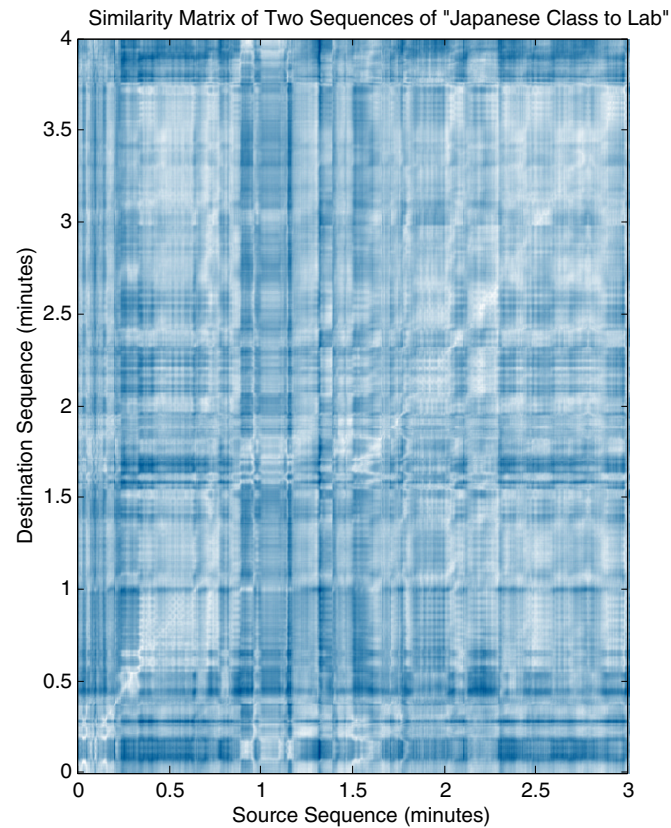
3000 frames thus 3000 states (computationally heavy!)



# The Similarity Measure

## Another Example: leaving class

similar at times, dissimilar at others





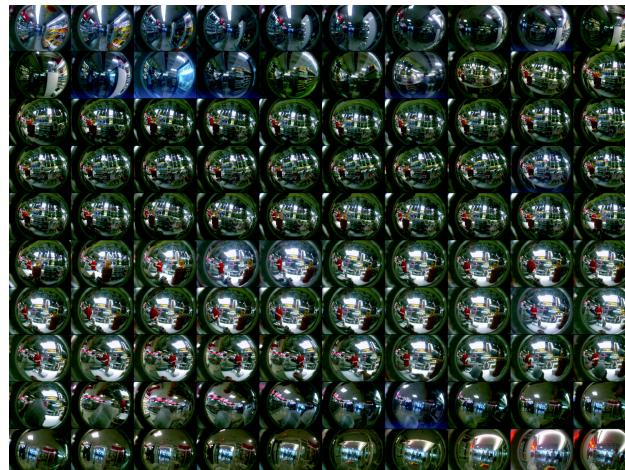
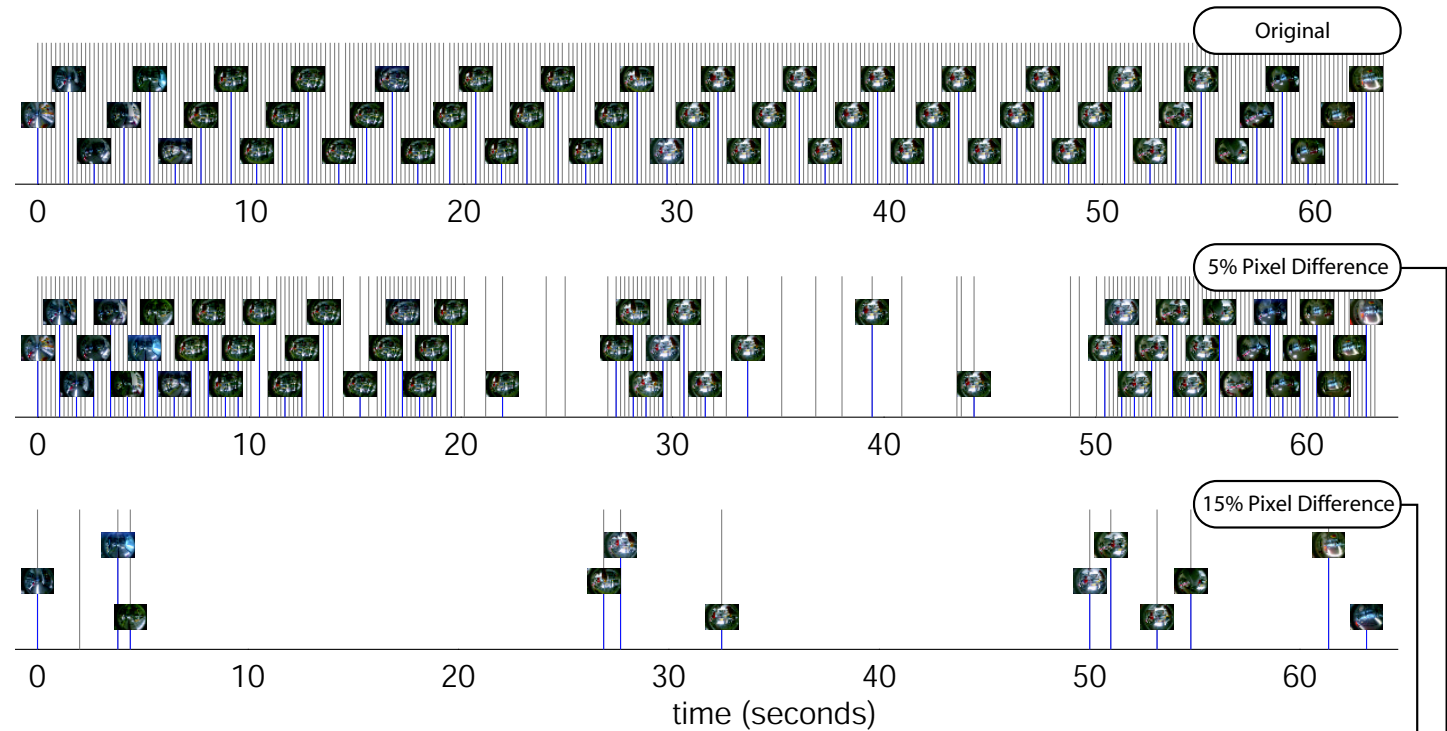
# The Similarity Measure

## Run Length Encoding

Remove Redundancy  
via Image Similarity

Before  
1 day = 200,000 images

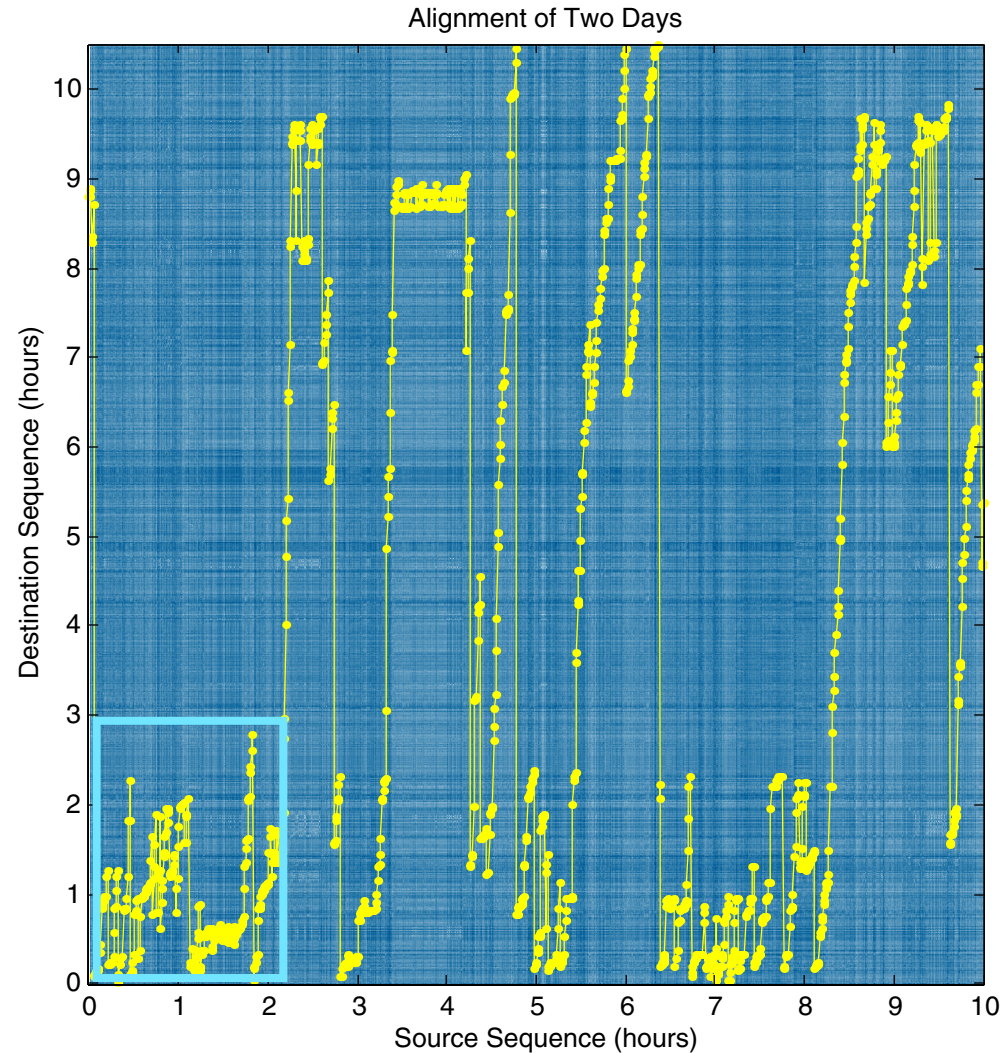
After  
1 day = 3,000 images



# The Similarity Measure

## Alignment of a day

RLE at 15% allows alignment of a pair of days!

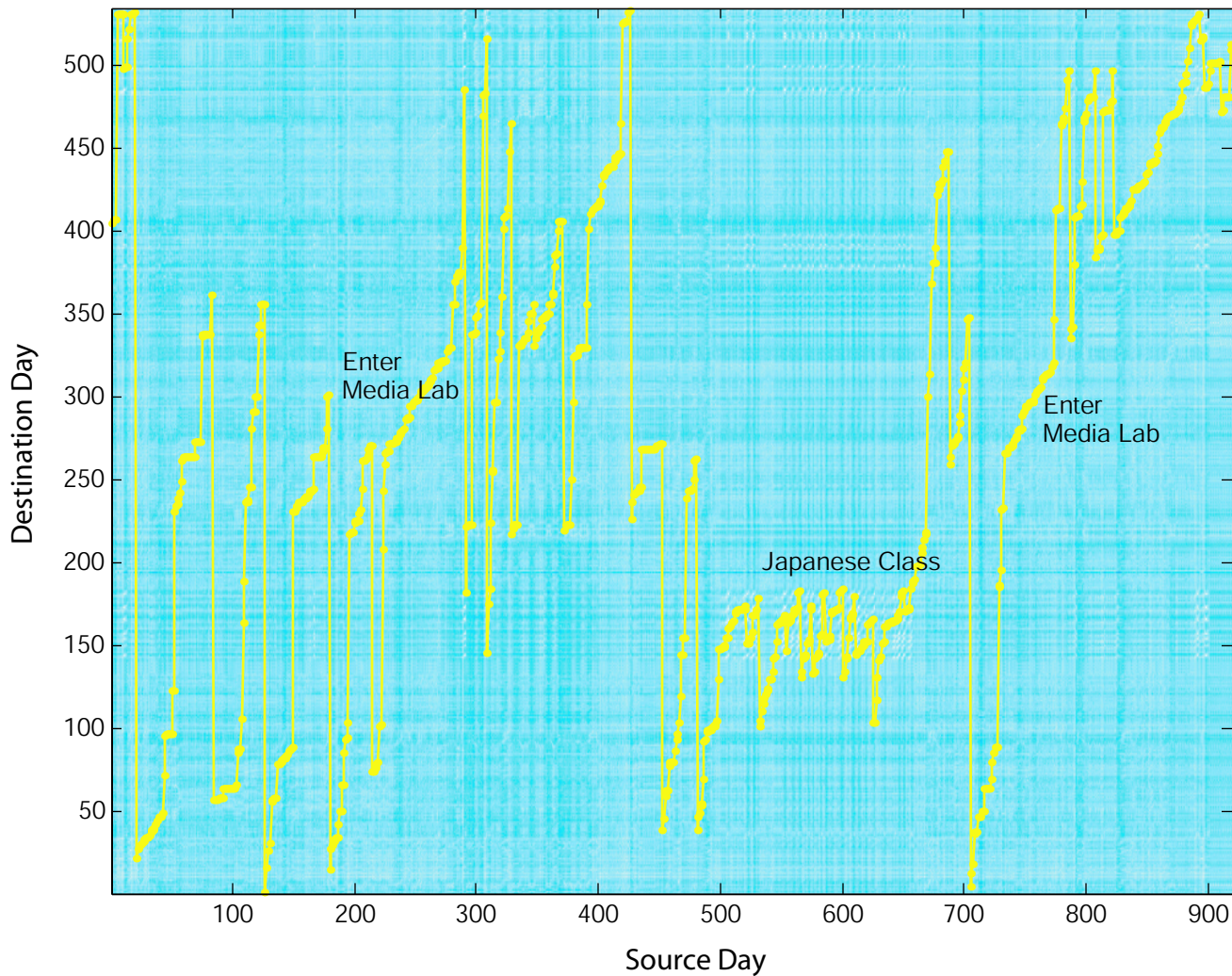




# The Similarity Measure

## Alignment of a day

The finer detail in the morning...



# The Similarity Measure

## Alignment of a month

Problem:

- A pair of days typically only align sensibly for a few situations.

Solution:

- Keep adding more sequences to the alignment HMM.

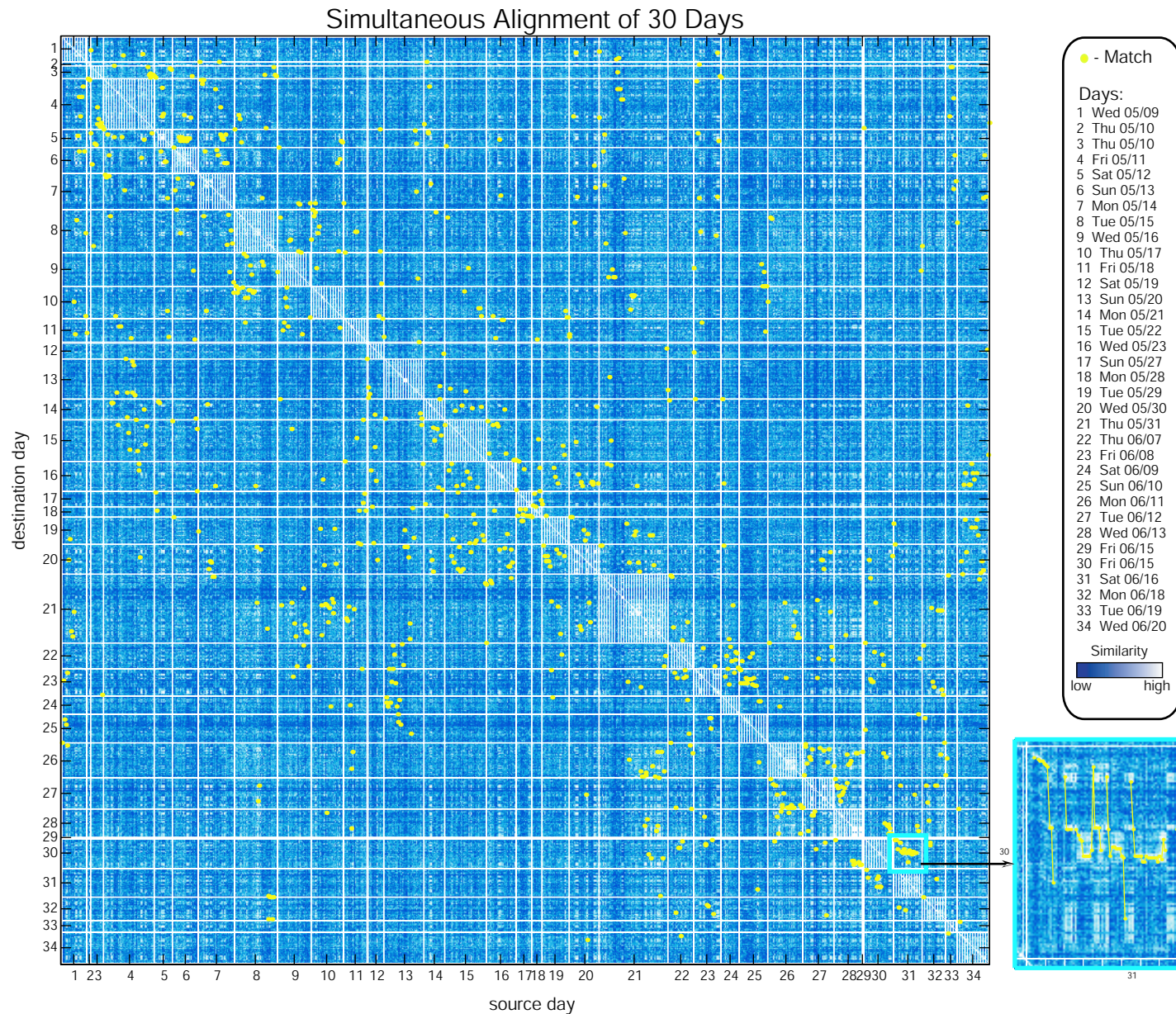
Aligning 1 day to 30 days:

- Build similarity matrix of 30 days at 5 minute resolution.
- Build alignment HMM from each 5 minute chunk in 29 days.
- Align remaining day to the 29 days using the HMM.
- repeat 30 times...



# The Similarity Measure

## Alignment of a month

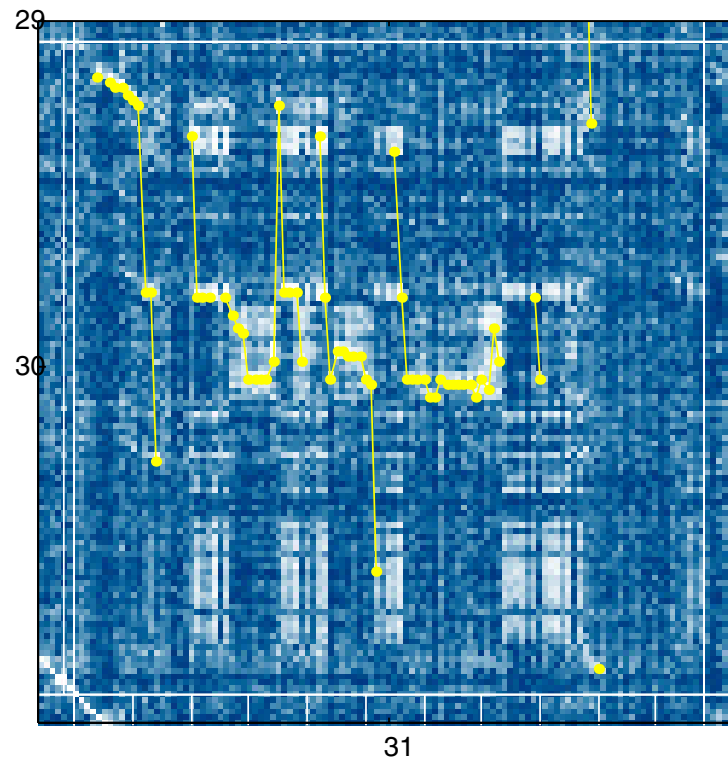


# The Similarity Measure

## Alignment of a month

Each moment can be aligned to any moment in 29 days.

we will use  
this later...

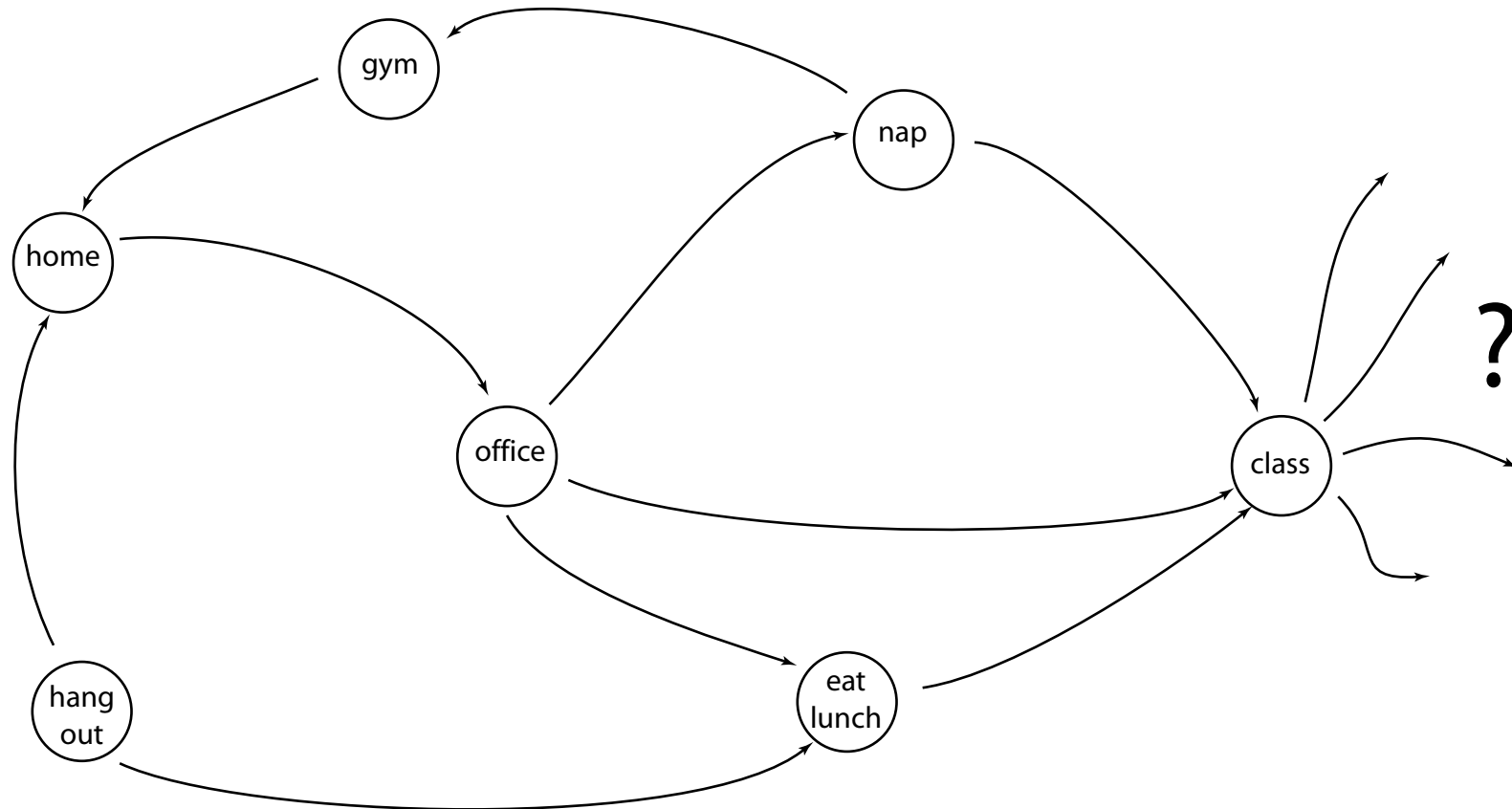


# Life's Perplexity

## Motivation

“When you come to a fork in the road, take it.”

- Yogi Berra



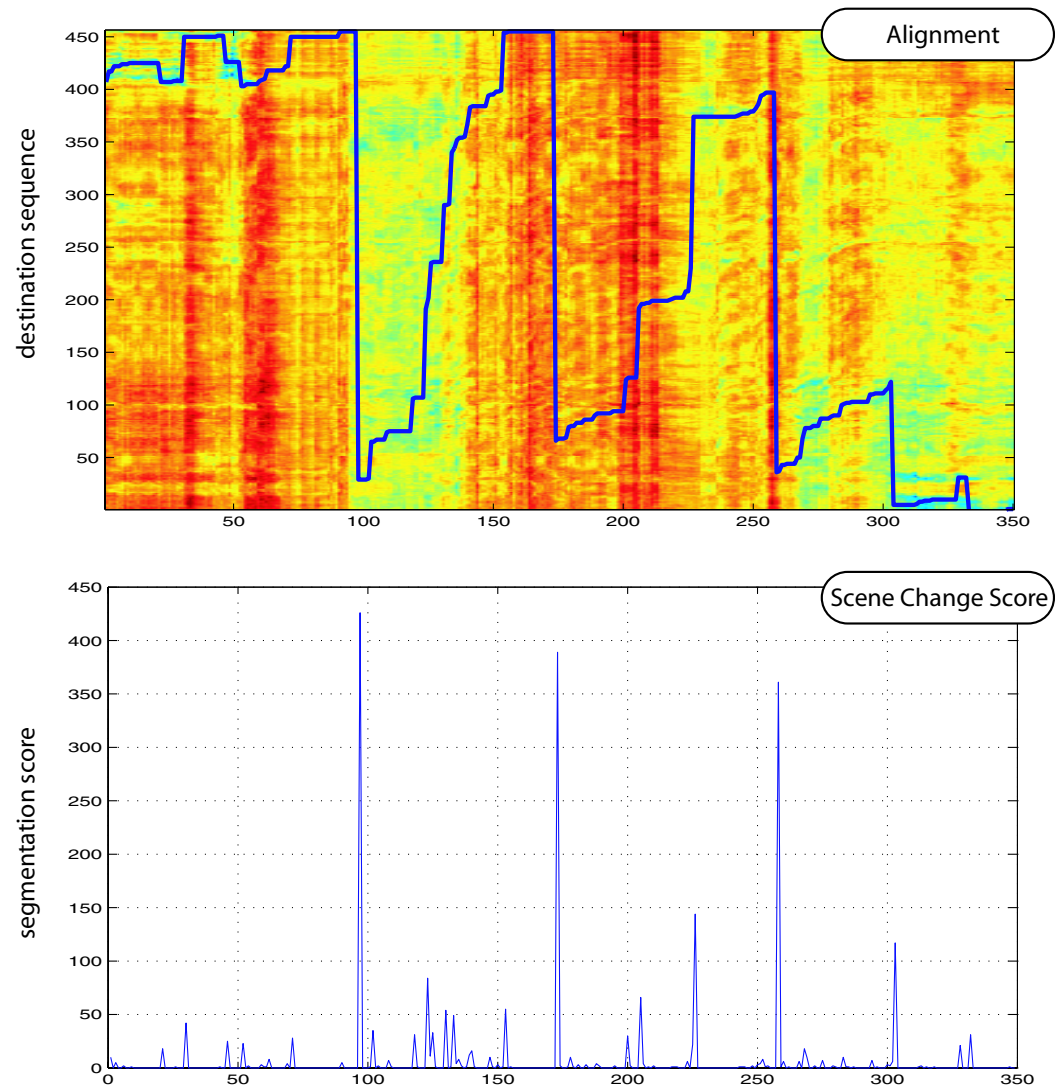
- Where are the decision points? (i.e. what are the nodes)
- What is the perplexity each time a decision is made?
- How consistent is the decision?



# Life's Perplexity

## Scene Segmentation

$\beta$ -transitions denote moments of divergence from past behavior.



Score each  $\beta$ -transition by its size in time.

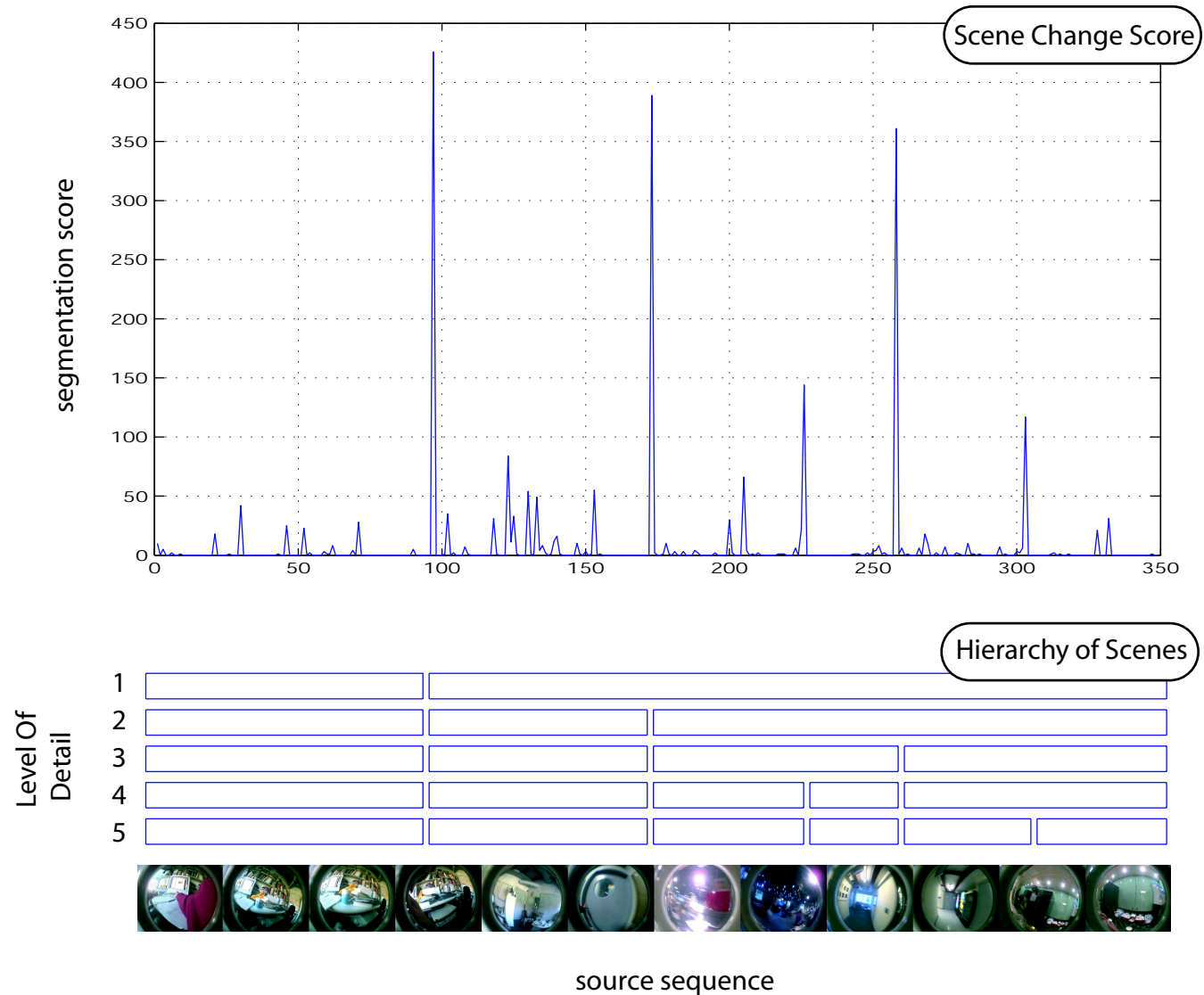


# Life's Perplexity

## Scene Segmentation

Sweeping a threshold  
yields a hierarchy of  
scene segmentations.  
30 days =  $\sim 1000$  scenes

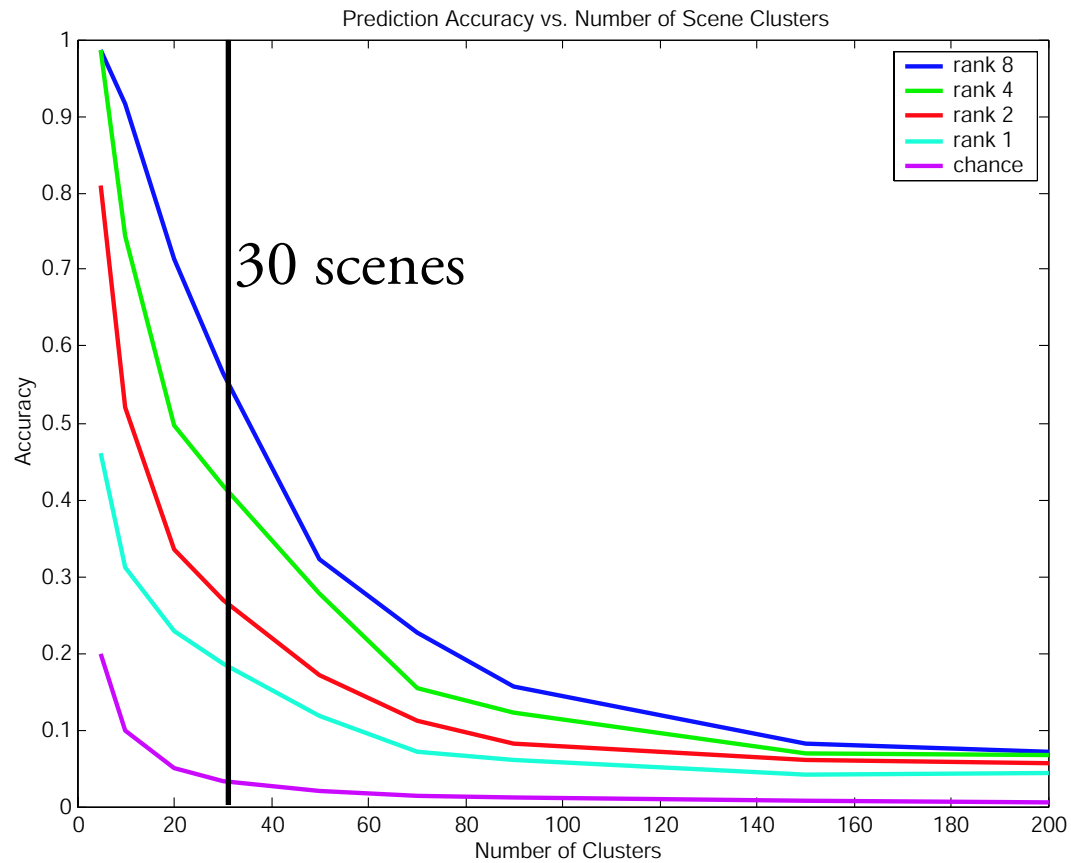
To get scenes cluster  
using alignment as the  
similarity measure.



# Life's Perplexity

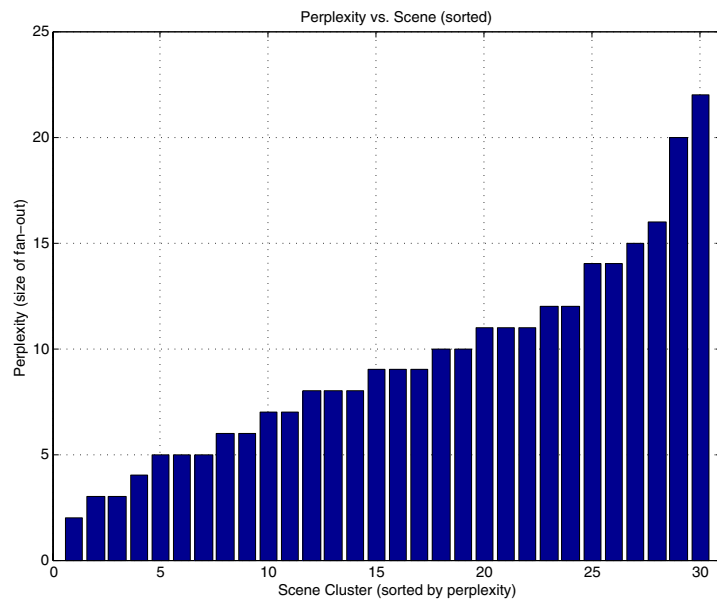
How many nodes?

Degree of redundancy is independent of the # of nodes.

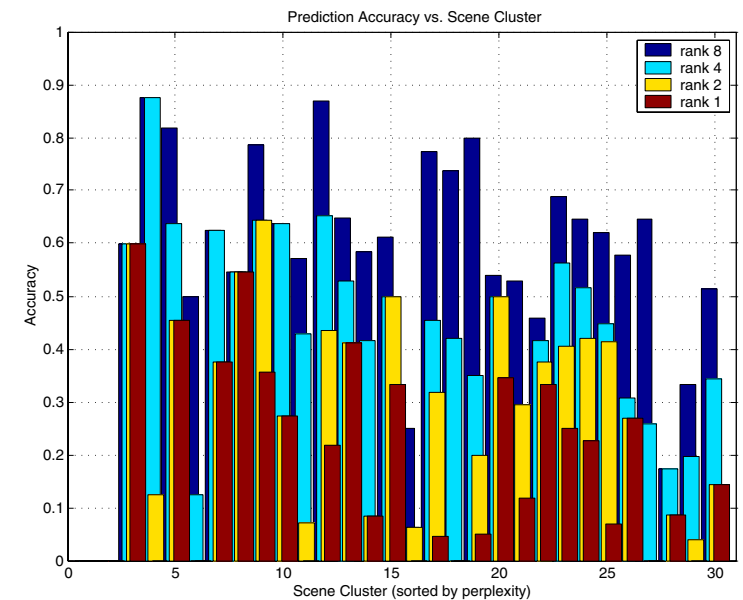
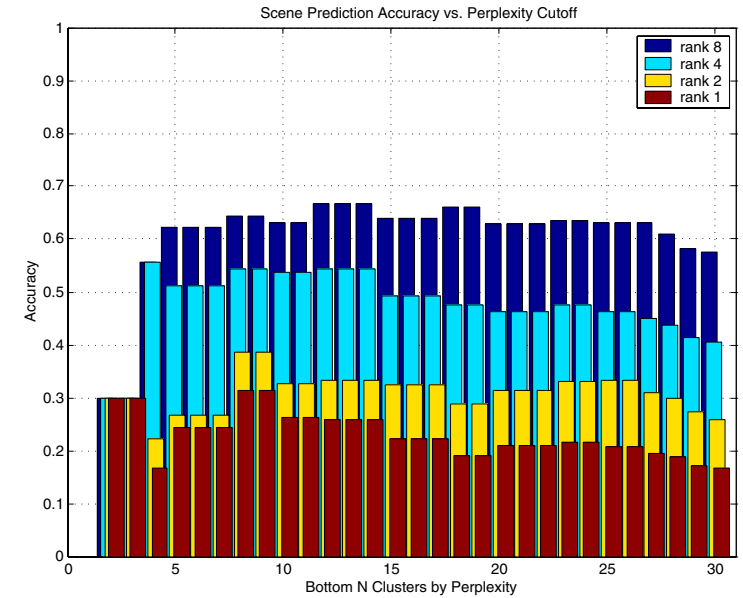


# Life's Perplexity

## 30 Scene Clusters

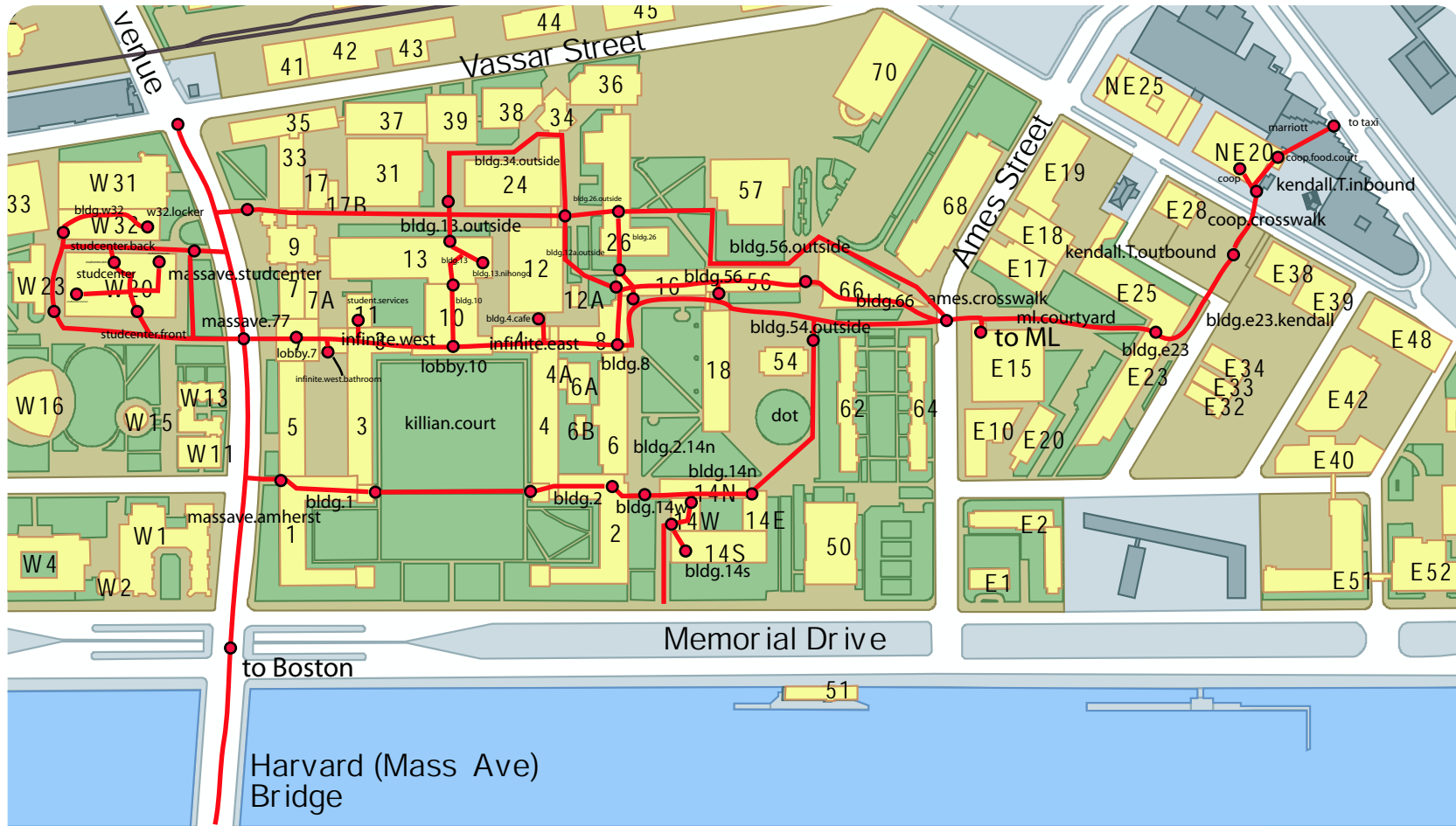


Accuracy is independent of perplexity!



# Situation Classification

What is a situation?



situation = location + activity

# Situation Classification

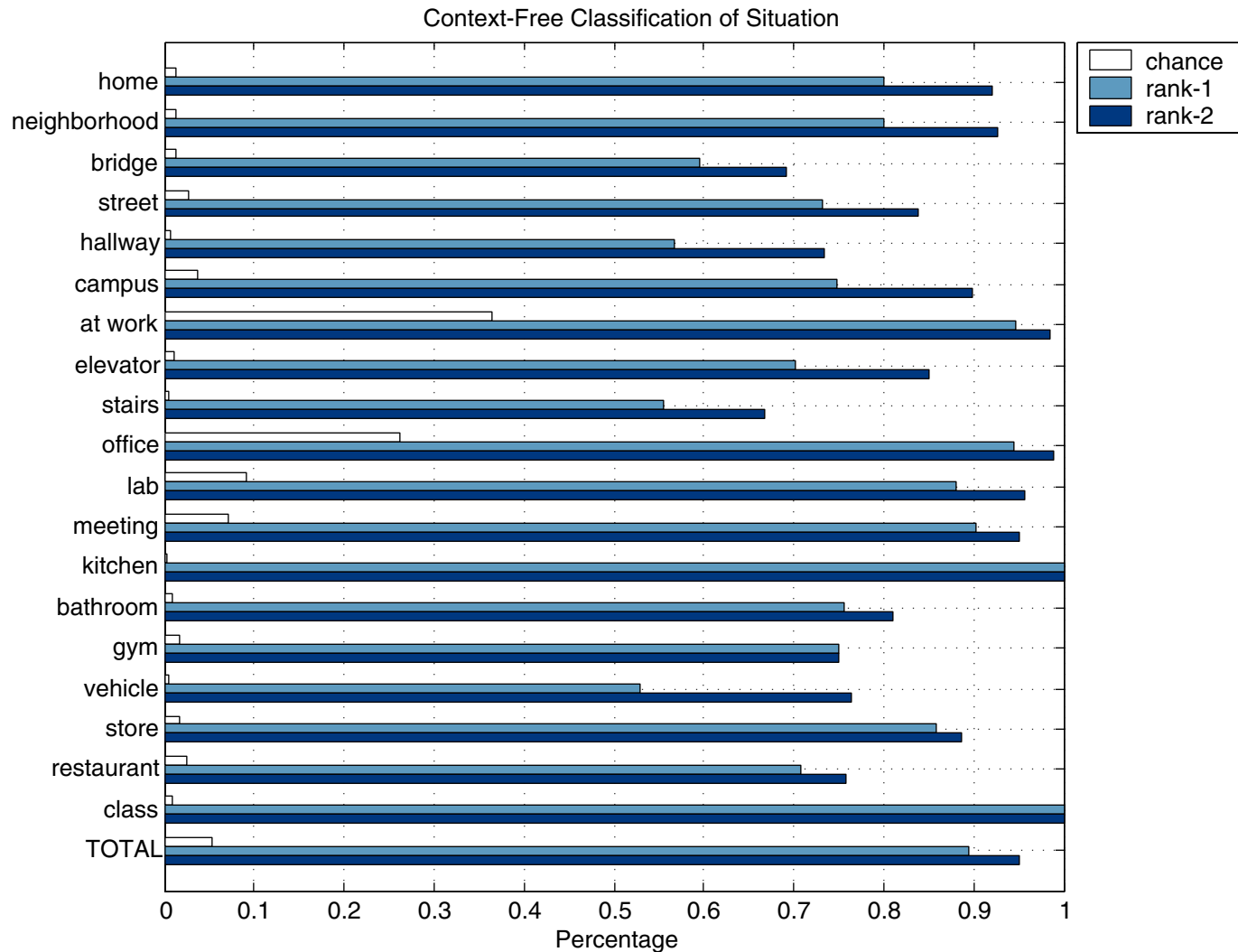
## 19 Situations

home	apartment
neighborhood	Beacon St., Mass. Ave. (Boston-side)
bridge	Harvard Bridge, Longfellow Bridge
street	Kendall Sq., Boston Downtown, Main St., Memorial Dr., and more
hallway	Infinite Corridor and more
campus	inside & outside of bldg. 56, 66, 7, 10, and more
at work	anything in the Media Lab
elevator	any elevator
stairs	any stairs
office	my office at lab
lab	the area outside of my office
meeting	any meeting
kitchen	kitchen (at home and lab)
bathroom	any bathroom
gym	Dupont
vehicle	taxi, subway, bus
store	any store
restaurant	any restaurant
class	any class

\* Every 5 minute interval over 20 days was labeled with its situation(s).

# Situation Classification

Context-free



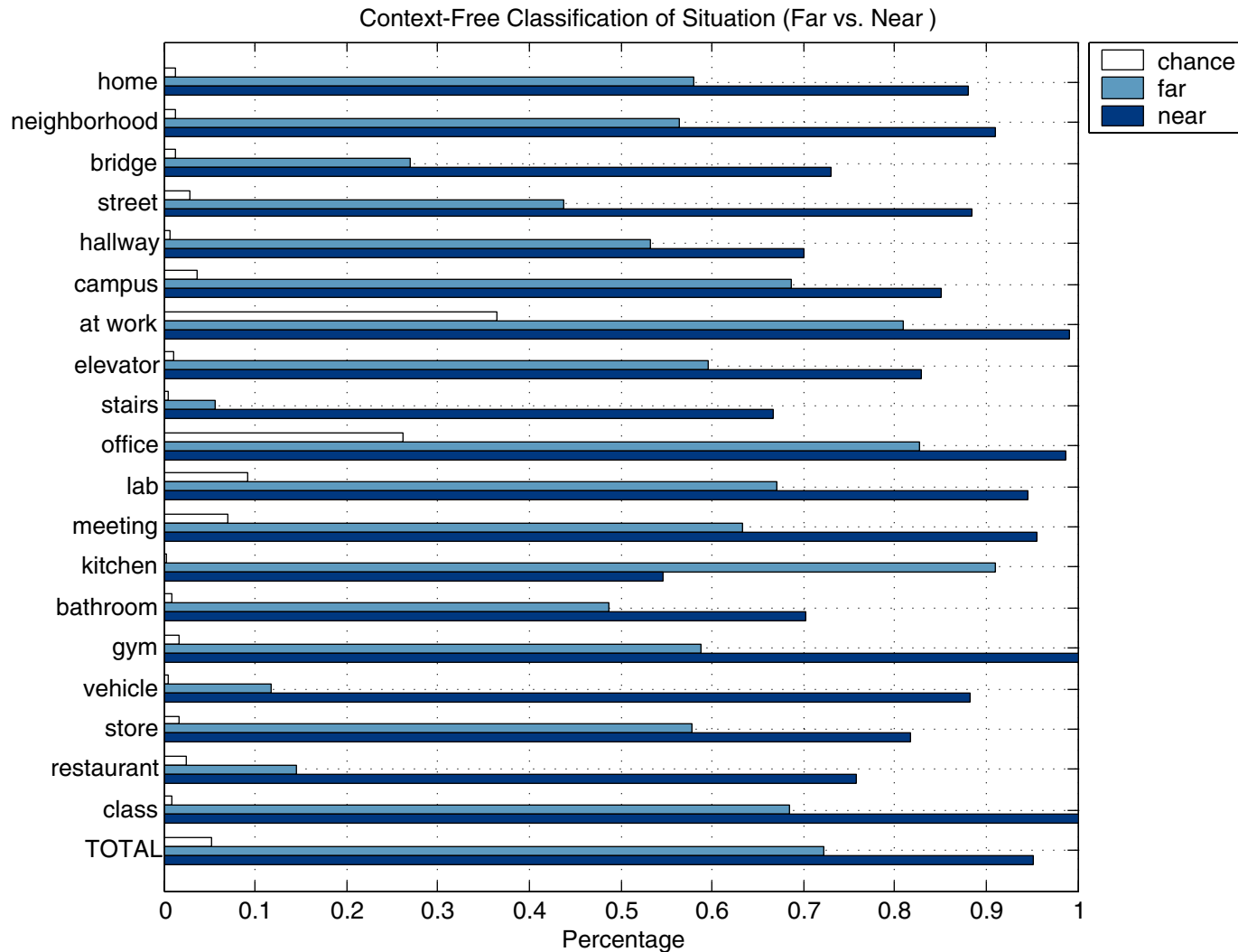
Total:

rank-1 = 89%

rank-2 = 95%

# Situation Classification

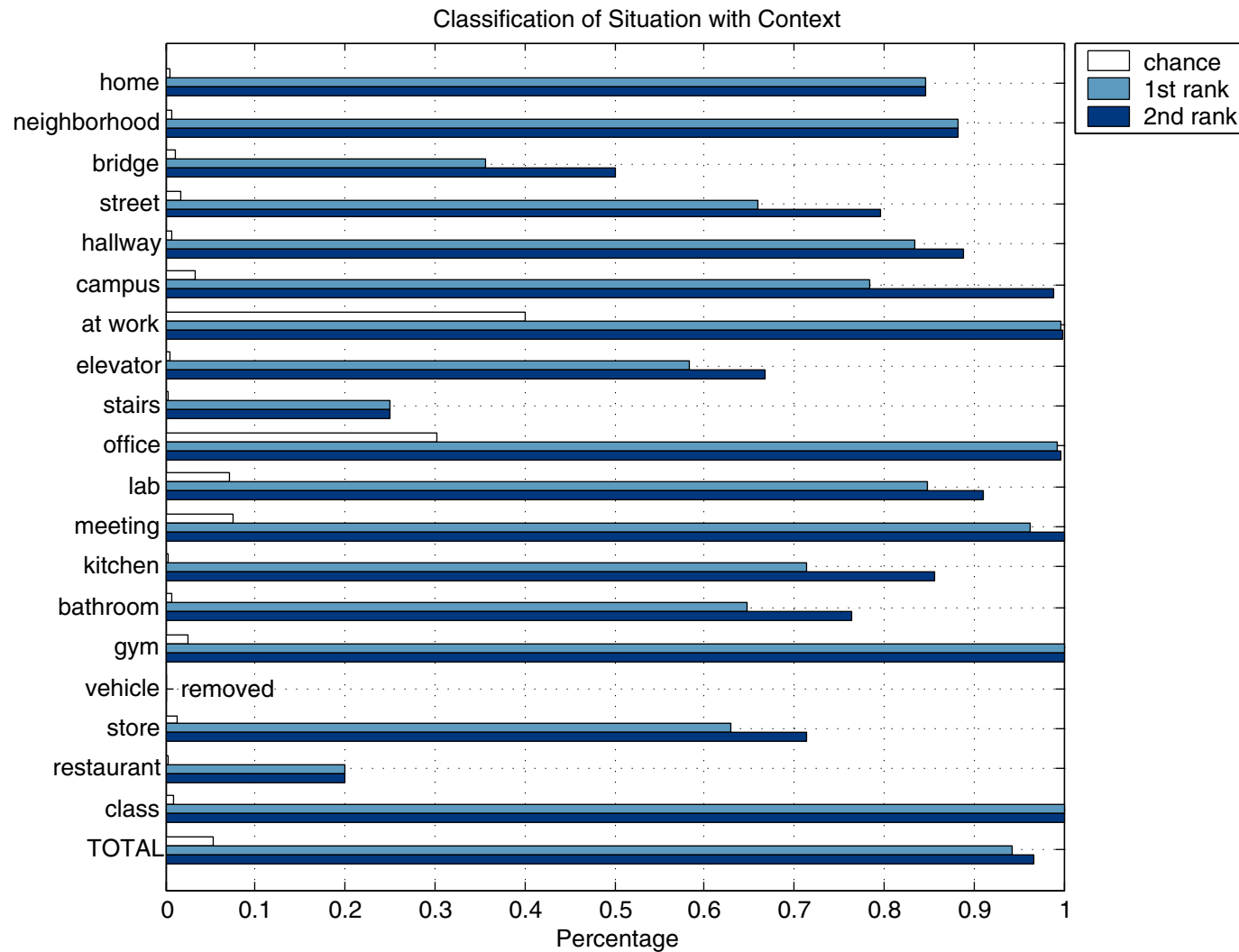
## Far vs. Near in Time



Total:  
far = 72%  
near = 95%

# Situation Classification

with Context



Total:

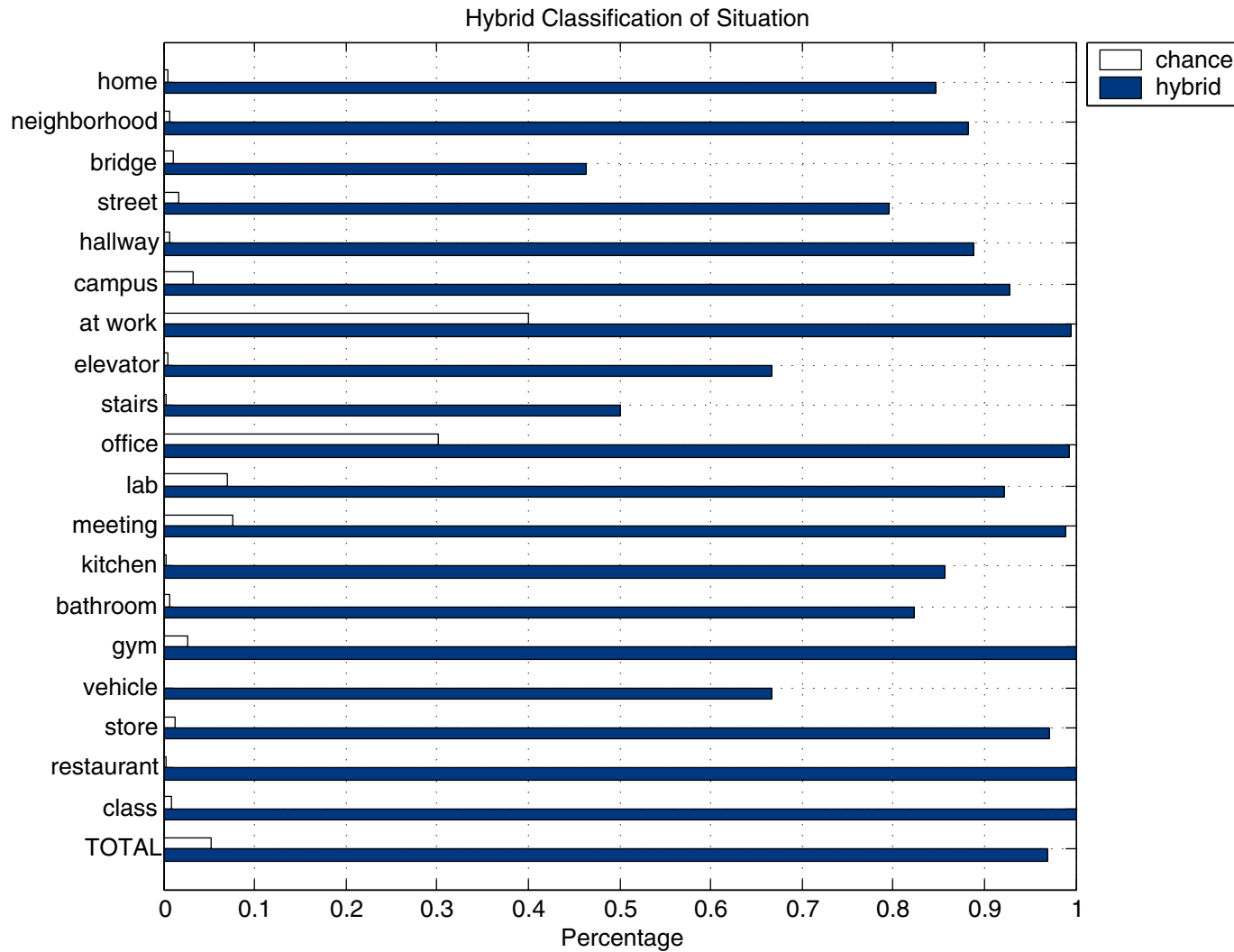
rank-1 = 94%

rank-2 = 97%



# Situation Classification

## A Hybrid Classifier



Total:  
rank-1 = 97%

What have we shown?

- Capturing complete records of an individual's day-to-day life is practical.
- The capabilities of peripheral and insect-like perception.
- Simple models can capture even the complex structure of human behavior.
- How to compare moments of an individual's life at multiple time-scales.
- That day-to-day behavior is redundant at multiple scales.
- How to classify situations.

# Epilogue

## Acknowledgements

I am indebted to...

Sandy, my patient advisor

Joe and Trevor, my patient thesis committee

Sumit, Tanzeem and Vikram, my officemates

All the Vismodders: Karen, Rich, Nathan, Steve, Liz, Tony, and many more.

All of my friends in the gymnastics team, the breakin' crew Imobilare, and Japanese class.

I dedicate this work to my little brother, Peter.

# Life Patterns

structure from wearable sensors



The End