

Computer Vision I

Review & Outlook - 10.07.2013



TECHNISCHE
UNIVERSITÄT
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Review & Outlook

- Before the semester ends, I would like to give an overview / review of the things that we have learned this semester.
 - It may have seemed to go by quickly, but we have covered a lot of ground.

>100 Things We've Learned

- | | | |
|------------------------------|----------------------------|--------------------------|
| ■ Computer vision | Thin lens formula | Canny edge detector |
| ■ Application examples | Depth of field | Non-maximum suppression |
| ■ Artistic cues | Camera artifacts | Laplacian pyramid |
| ■ Human visual cues | Color cameras | Template-based matching |
| ■ Pinhole camera | Bayer pattern | SSD |
| ■ Perspective projection | | Subspaces |
| ■ Orthographic projection | Linear filtering | Dimensionality reduction |
| ■ Coordinate transformations | Convolution kernels | Correlation |
| ■ Spatial sampling | Median filter | PCA |
| | Morphology | SVD |
| | Gaussian pyramid | Eigenrepresentations |
| ■ Vanishing points | Aliasing | Eigenfaces |
| ■ Perspective distortions | | |
| ■ Homogeneous coordinates | Template-based recognition | Appearance manifolds |
| ■ Camera intrinsics | Edge detection | View-based recognition |
| ■ Extrinsic | | |



>100 Things We've Learned

■ Bag-of-Words model	Bayesian decision theory	Brightness constancy
■ Color histograms	Naive Bayes classifier	OFCE
■ Histogram distances	Discriminative & generative approaches	Aperture problem
■ Receptive field histograms	SVMs	Lucas-Kanade
■ Interest points	Kernel trick	Image registration
■ Invariant features	Histogram & Pyramid match kernels	Image warping
■ Structure tensor	Sliding window detector	Coarse-to-fine estimation
■ Harris points	HOG	Linear camera calibration
■ Scale space	Motion field	Homogeneous least squares
■ Scale selection	Optical flow	Homographies
■ Harris-Laplace points	Image interpolation	Coordinate normalization
■ SIFT features		RANSAC
■ Shape context		Panorama stitching
■ Performance evaluation		Triangulation
■ Vector quantization		

>100 Things We've Learned

- Epipolar geometry
 - Epipolar constraint
 - Essential matrix
 - Fundamental matrix
 - Eight-point algorithm
 - Estimating projection matrix
 - Binocular stereo
 - Disparity
 - Rectification
 - Baseline
 - Window-based matching
 - Normalized correlation
 - Uniqueness constraint
 - Ordering constraint
- Segmentation
 - Figure-ground labeling
 - Supixels
 - Gestalt factors
 - Occlusion
 - Clustering
 - K-Means
 - Mean Shift
 - Kernel Density Estimate
 - Graph-based clustering
 - Graph-cut segmentation
 - Normalized cuts



So are we done?



[Marr]

Computer Vision Timeline

1975-1985

Vision as AI

Early view (50's-60's): Minsky thought the vision sub-problem of AI could be solved by a single PhD student in a single summer. Done. Move on.

- Lofty goals and early excitement.

[Black]

Computer Vision Timeline

1975-1985

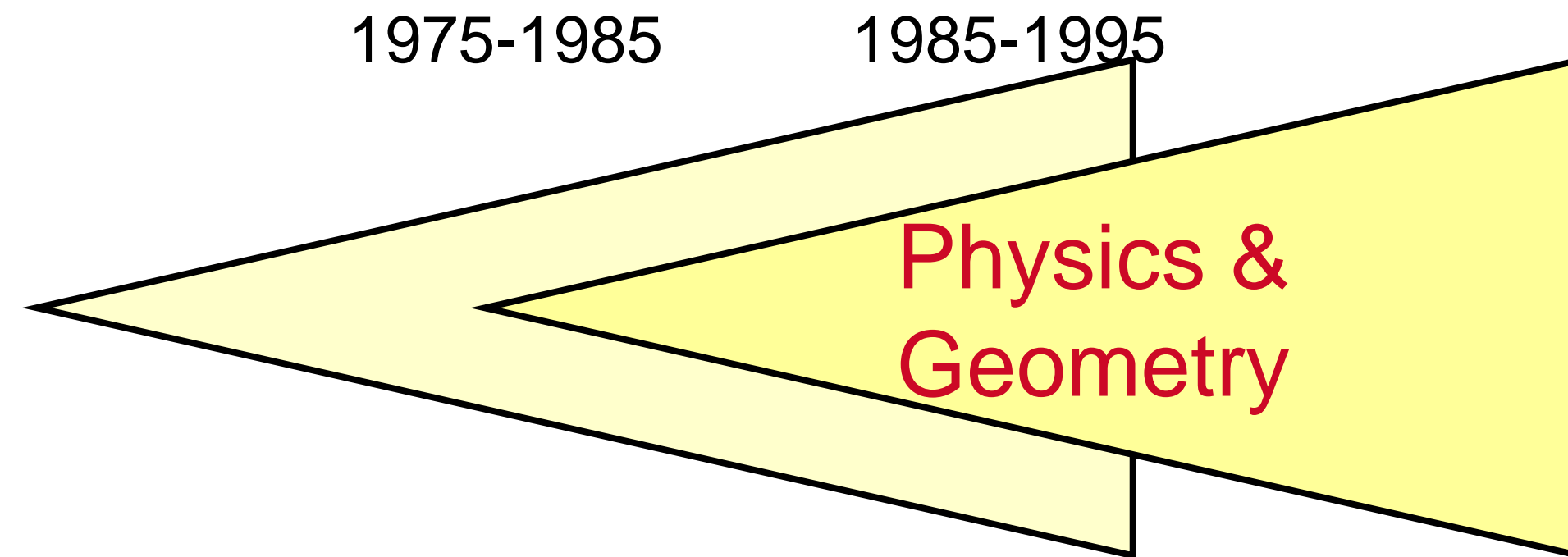
Vision as AI



- Shattered dreams and early disappointment.

[Black]

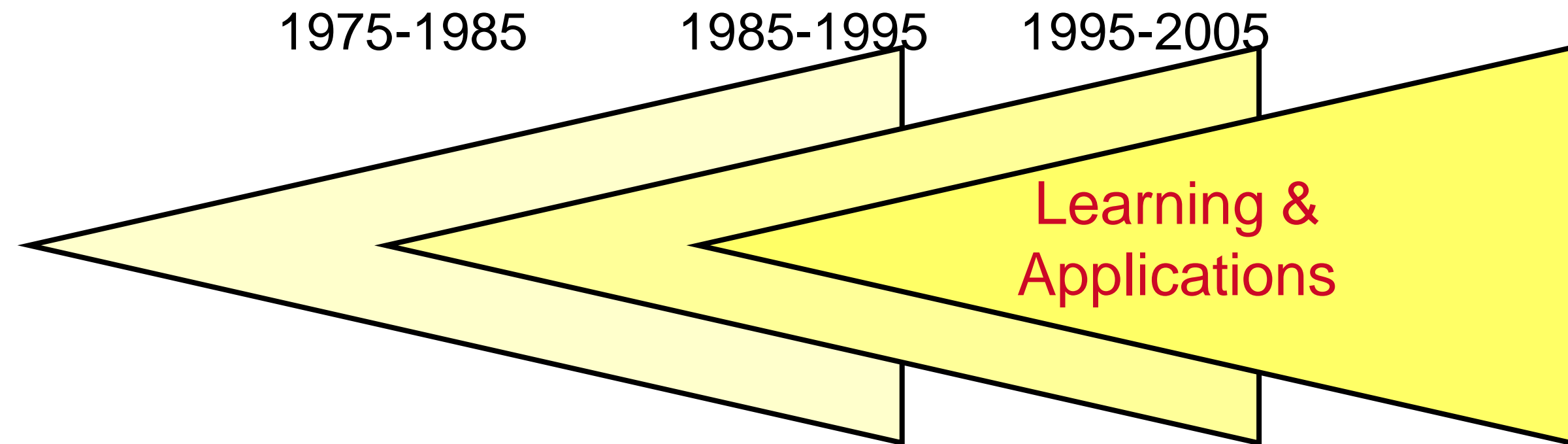
Computer Vision Timeline



- Regroup, focus on the basics:
 - Metric reconstruction, quantitative evaluation
 - Optimization methods

[Black]

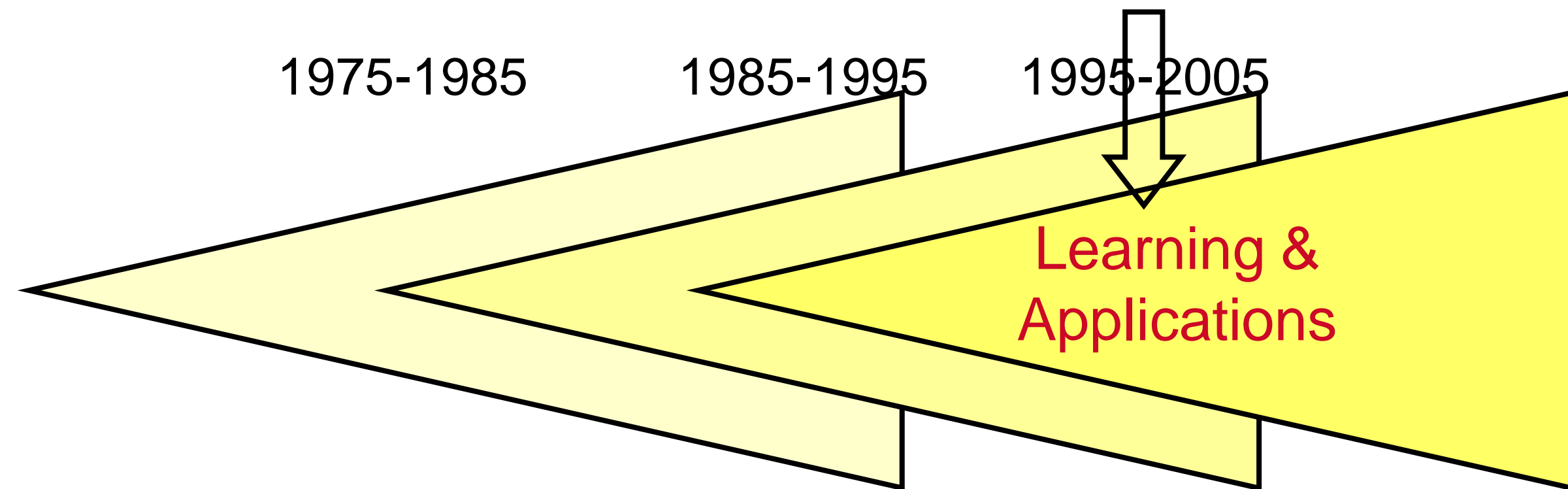
Computer Vision Timeline



- Trends: Big disks, digital cameras, fast processors, desktop video.
 - Machine learning provides a new grounding.

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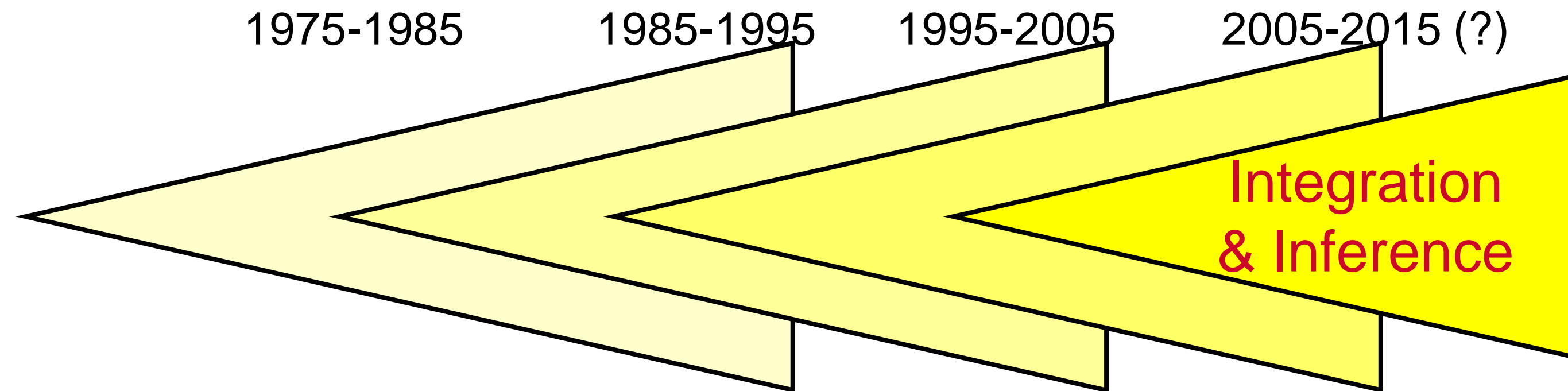
Computer Vision Timeline



- Real applications come around late:
- E.g., optical flow:
 - Horn & Schunk, early 80s
 - Application in movies (e.g. “The Matrix”), late 90s

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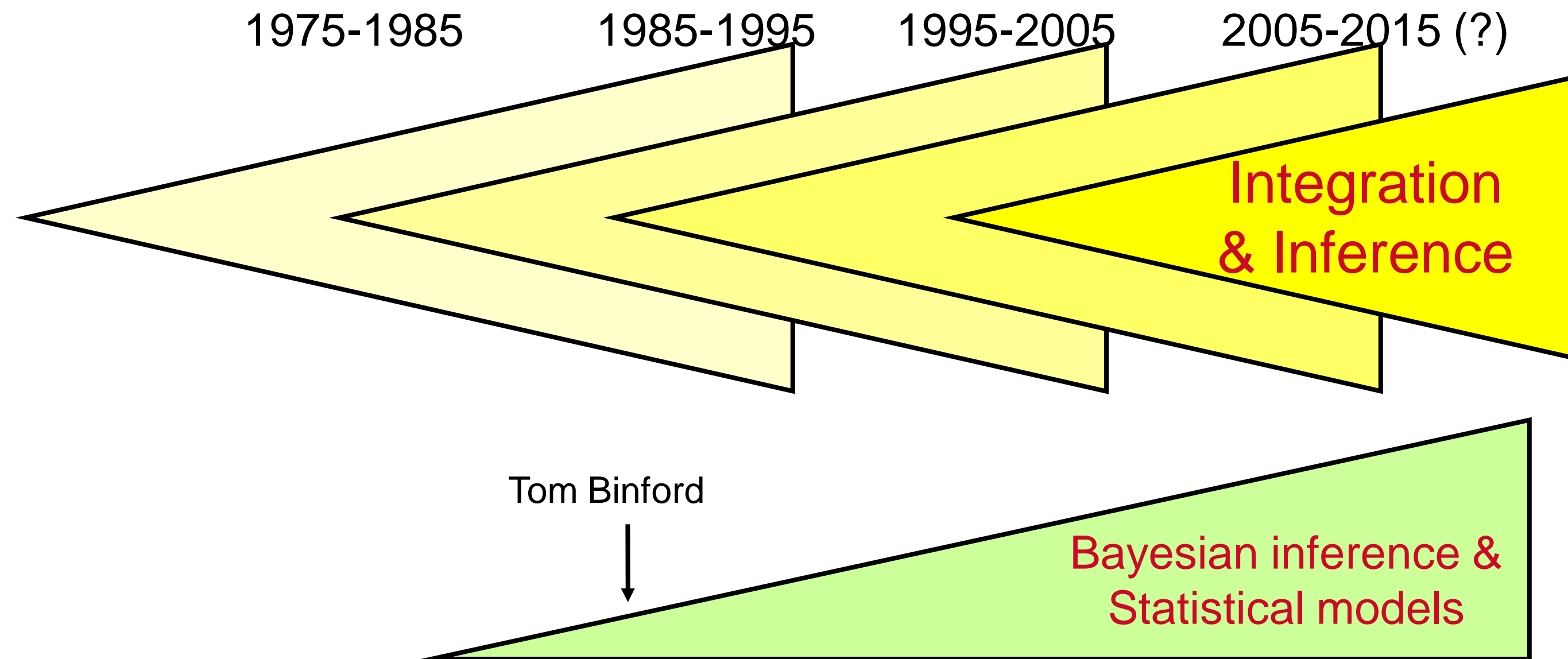
Computer Vision Timeline



Return to some of the early goals with new tools.

[Black]

Computer Vision Timeline



Probabilistic Modeling and Inference

- One trend is quite clear recently:
 - **Statistical tools & probabilistic inference** are used in more and more areas of computer vision.
 - **Machine learning** has become an important tool for vision and for many other areas in CS.



What's next

■ If you want to learn more about this:

- **Computer Vision II**
- Probabilistic methods in computer vision
- Winter term, V2 + Ü2

■ Also recommended:

- **Machine Learning: Statistical Methods I+II**
- Introductory class on statistical machine learning, V2 + Ü2
- WS: ML II, SS: ML I

Exam

■ Written exam

- Wednesday 21.08.2012, 10:00 -- 12:00am
- Place: C205@Piloty & 074@S3|05,
- Language: English or German, as you prefer

■ Exam aids:

- **closed book**, i.e. no books, smart phones etc. allowed
- But: May bring a **handwritten** A4 help sheet (no typing/printing)
- use only one side

■ Are you registered?

Exam

- Problems aim at checking the understanding of important concepts and methods.
- No extensive computation/derivation
 - but the key formulas are important
 - you should know what they say and which assumptions were made to arrive there
- No Matlab programming
- There will be some open questions

- More details will be announced by email through the mailing list