A Computational Model for Top Down Visual Attention

Abhijit Sharang

Motivation

- Attention involves selectively processing certain aspects of the environment while ignoring others.
- Crucial for real time processing of stimuli
- Two aspects of visual attention
 - Bottom up
 - Top Down
- Top down is more dominant when the attention is goal oriented

The approach

- Sequential nature of task driven attention
- The idea is to exploit this aspect to construct a model which predicts the current state based on the previous state and the features generated in the current state
- Compare this model with other classifierbased models

Experiment

- Obtain data from a dynamic task : video games
 - 3D Driving School(player in motion)
 - Hot Dog Ambush(player static w.r.t environment)
- Identify features relevant to the task
 - GIST
 - Event in the video frame(for 3DDS)
 - Objects in the video frame(for HDB)
- Map features to the eye-gaze data for each frame to learn the mathematical model

Experiment(cont..)

- The models:
 - Mean Eye Position
 - Random Eye Position
 - Regression
 - K Nearest Neighbours
 - Dynamic Bayesian Network

Dynamic Bayesian Network

• *Two slice* Bayesian network



 Adjust m(structure of the DBN) and θ(transition matrix) to maximise P(E|m;θ)

Results(NSS)

• 3DDS



Results(NSS)

• HDB



Results(ROC)

• 3DDS



Results(ROC)

• HDB



Comparison of fixation(3DDS)

Original

Original



MEP





Regression







DBN





MEP





Regression













Rand

Regression

kNN

DBN

Original







Regression







DBN







MEP

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Rand

Regression

kNN

DBN



















Regression

DBN

















Original

MEP



Rand



Regression









Further work

- The model right now predicts eye fixation only.
- Can it be extended to saccades as well?
- Are global scene features of any importance in the model?
- Does the addition of bottom-up saliency model make any improvements in the result?

References

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- Peters, Robert J., and Laurent Itti. "Beyond bottom-up: Incorporating task-dependent influences into a computational model of spatial attention." Computer Vision and Pattern Recognition, 2007. CVPR'07. IEEE Conference on. IEEE, 2007.