

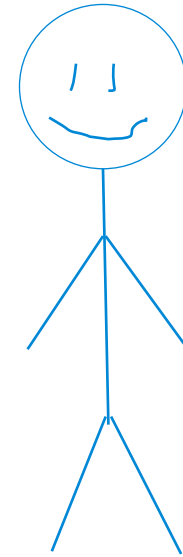
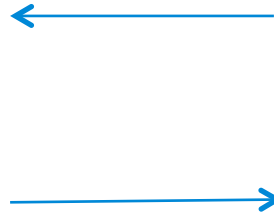
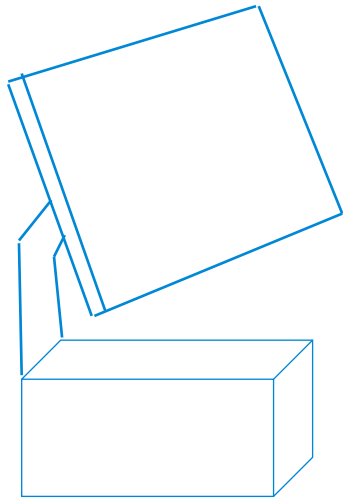
# Continuous Facial Expression Recognition for Affective Interaction with Virtual Avatar

Zhengkun Shang, Jyoti Joshi, Jesse Hoey

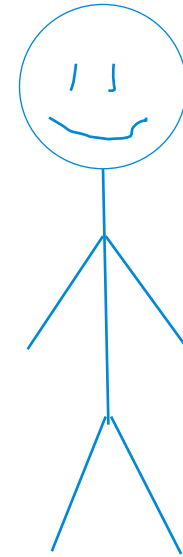
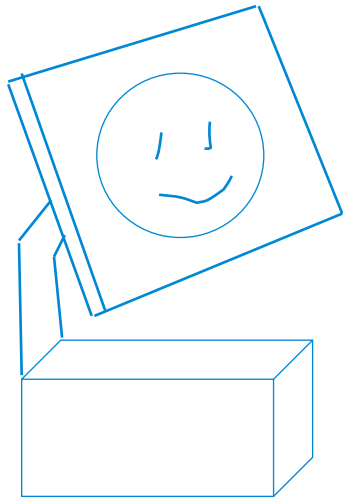


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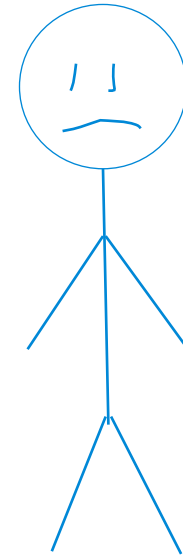
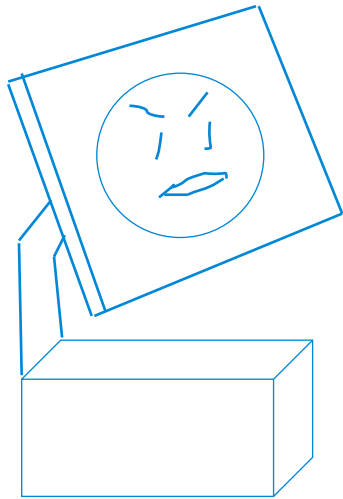
# MOTIVATION



# MOTIVATION



# MOTIVATION



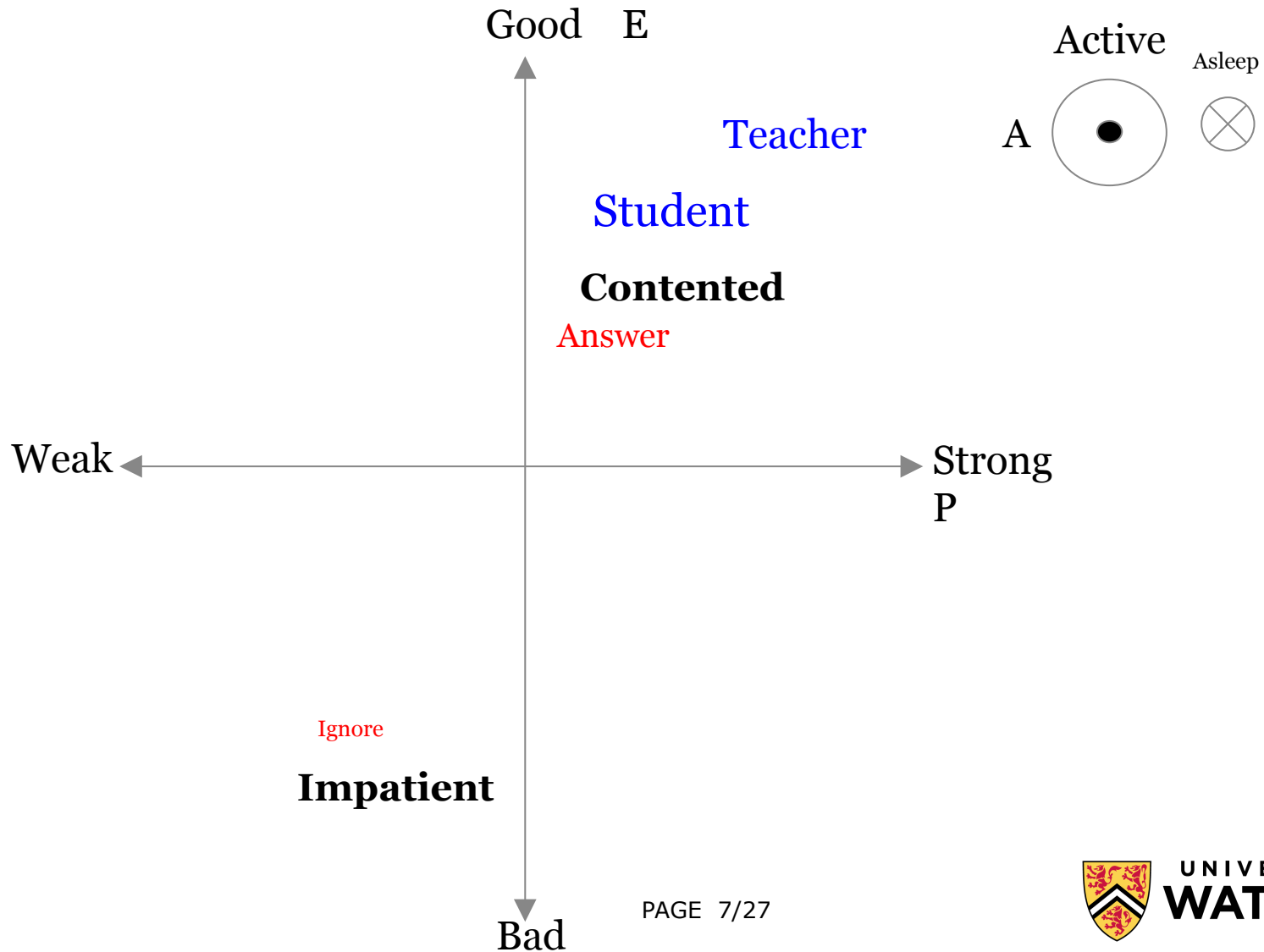
# What We Did?

- Estimated emotion in 3D continuous space from different features of facial expressions
- Used a probabilistic framework to simulate the interaction of a user and behavior-styled avatars

# Affect Control Theory - EPA Space

- Evaluation: how positive
- Potency: how powerful
- Activity: how active

# Affect Control Theory - EPA Space



# BayesACT

- A generalization of Affect Control Theory
  - keeps multiple hypothesis about both identities and behaviors as a probability distribution
- A sequential Bayesian model
  - estimates and updates variables over time from actions and observations

More information on BayesACT can be found at [bayesact.ca](http://bayesact.ca)



# Dataset

**Semaine Database** provides annotations for 93 videos

- Valence (Evaluation)
- Power (Potency)
- Arousal (Activity)



Avatar

















(Happy/Sad/Sensible/Angry)



User

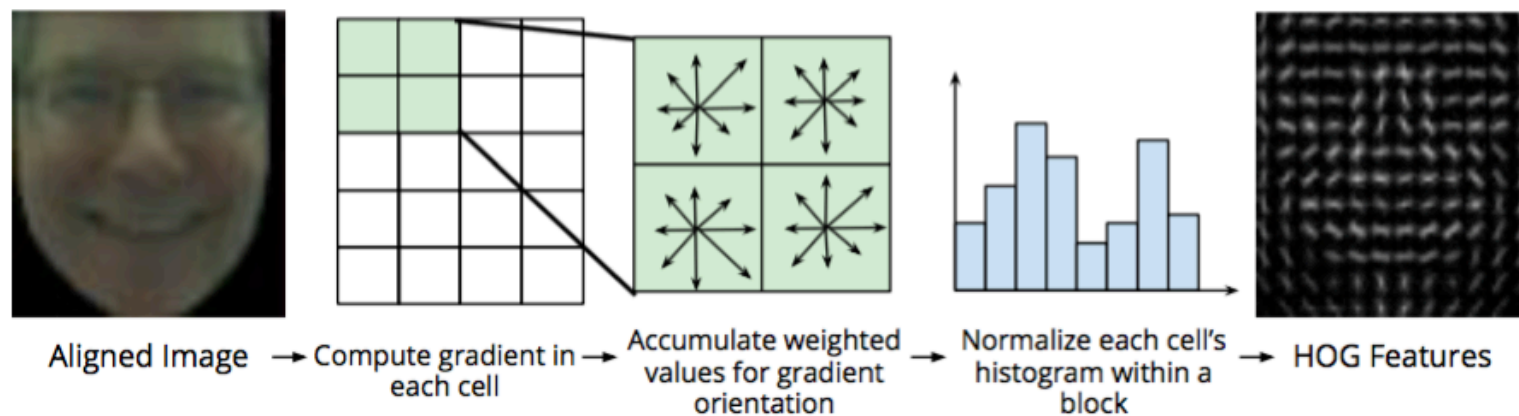
# Features

- Action Units
- Histogram of Oriented Gradient (HOG)
- Felzenszwalb's HOG (FHOG)

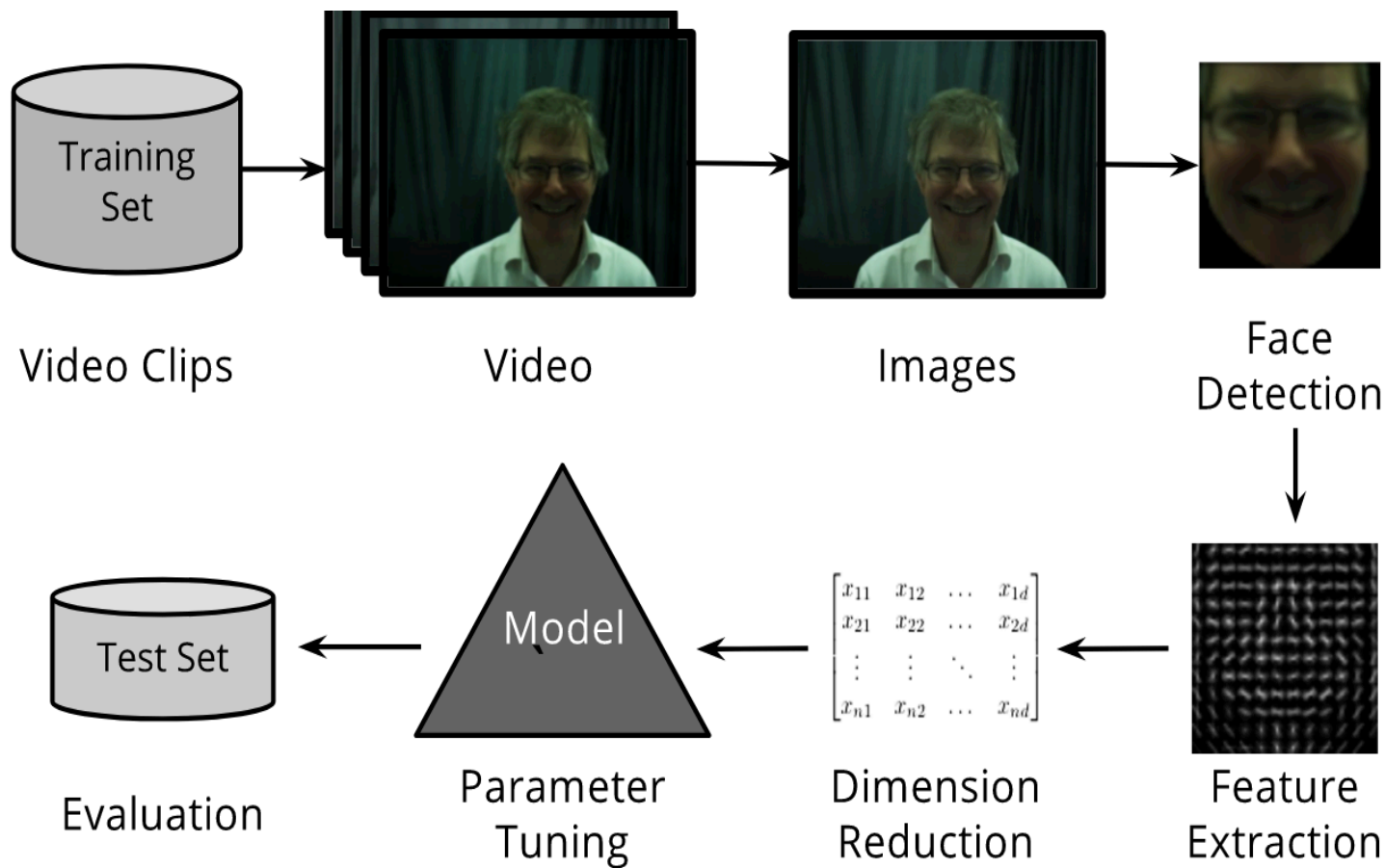
AU1  Inner Brow Raiser	AU2  Outer Brow Raiser	AU4  Brow Lowerer	AU5  Upper Lid Raiser	AU6  Cheek Raiser
AU7  Lid Tightener	AU9  Nose Wrinkler	AU10  Upper Lip Raiser	AU12  Lip Corner Puller	AU14  Dimpler
AU15  Lip Corner Depressor	AU17  Chin Raiser	AU20  Lip Stretcher	AU23  Lip Tightener	AU25  Lips Part
AU26  Jaw Drop	AU45  Blink			

# Features

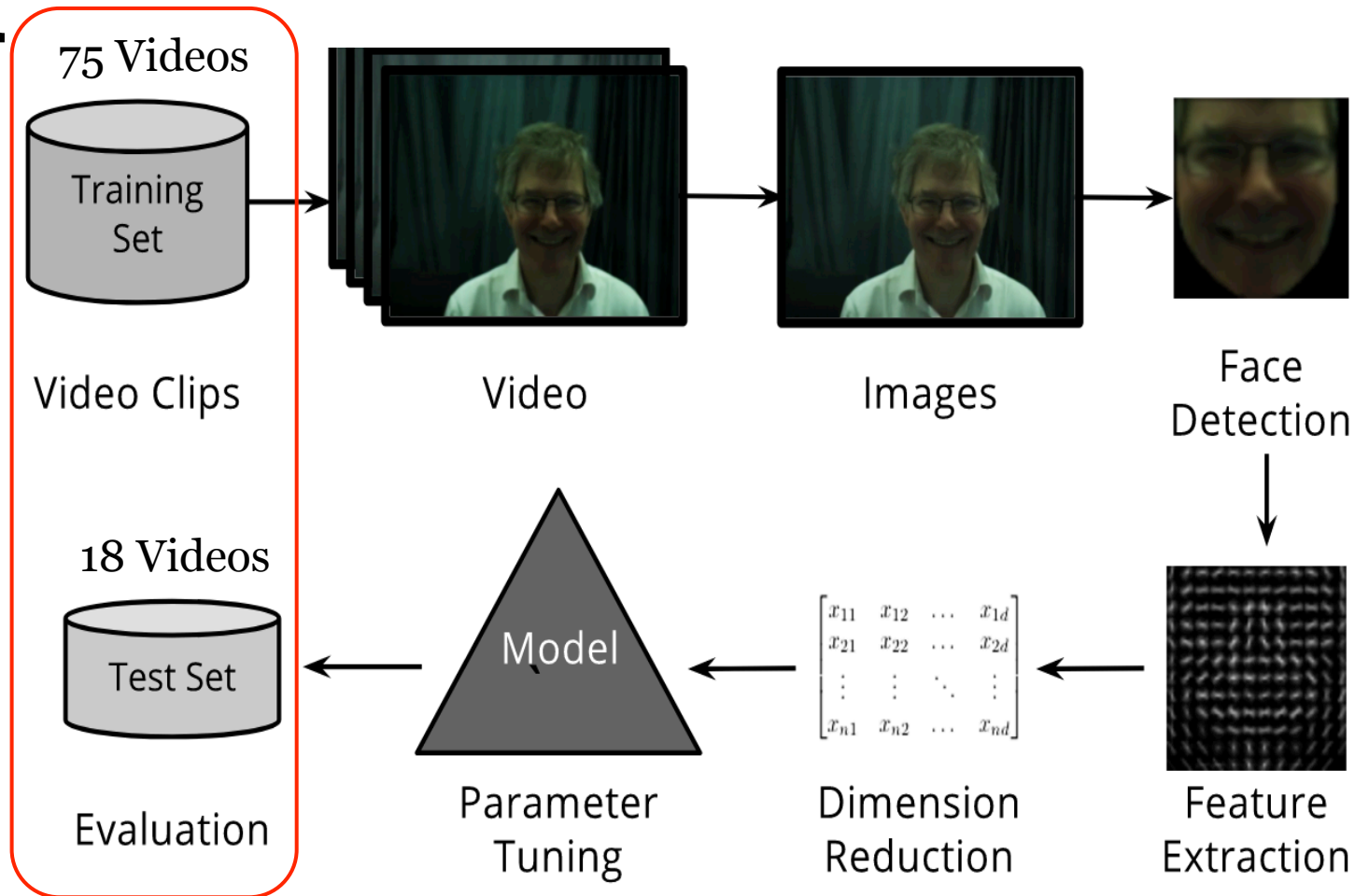
- Action Units
- Histogram of Oriented Gradient (HOG)
- Felzenszwalb's HOG (FHOG)



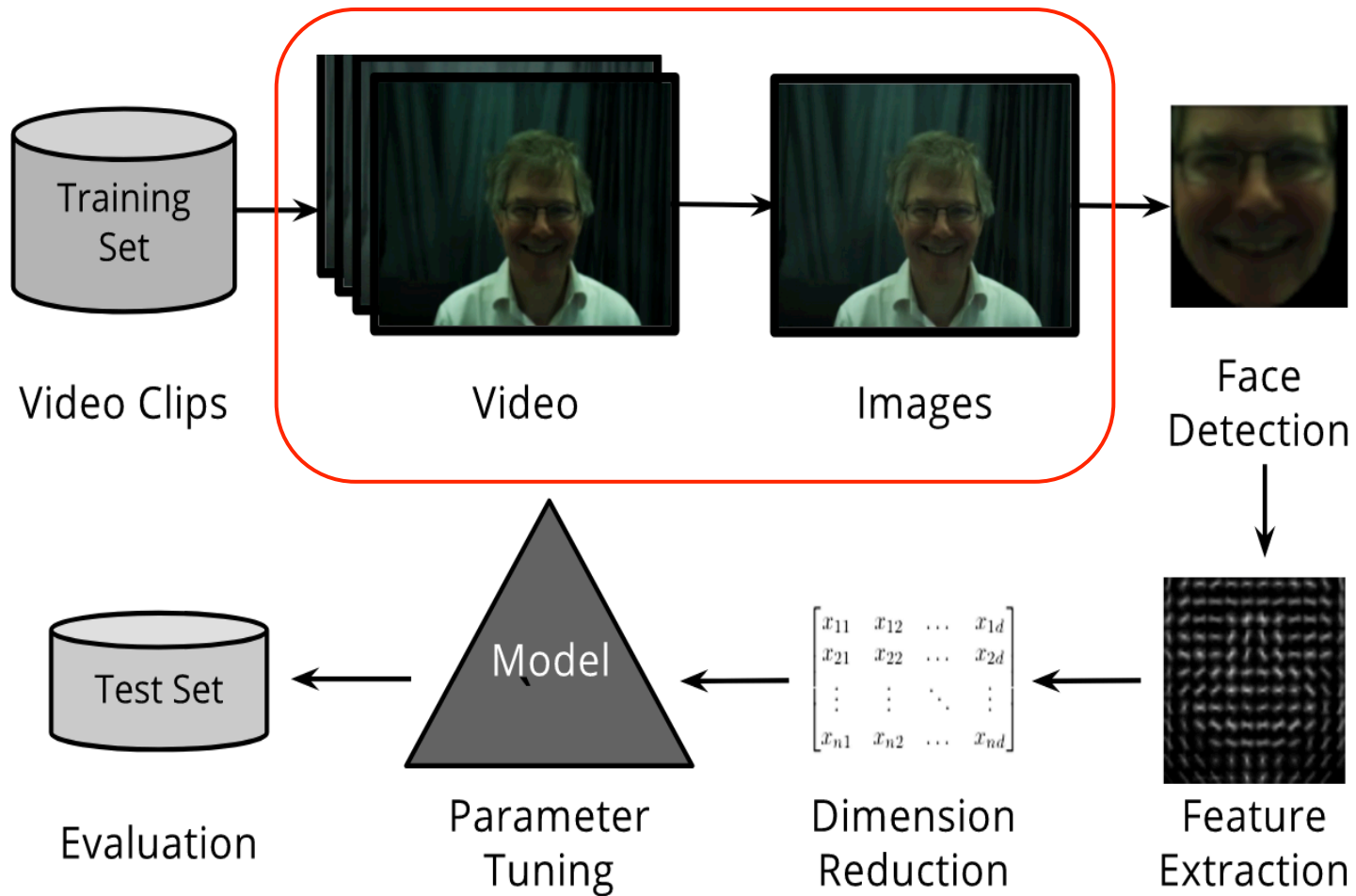
# Procedures



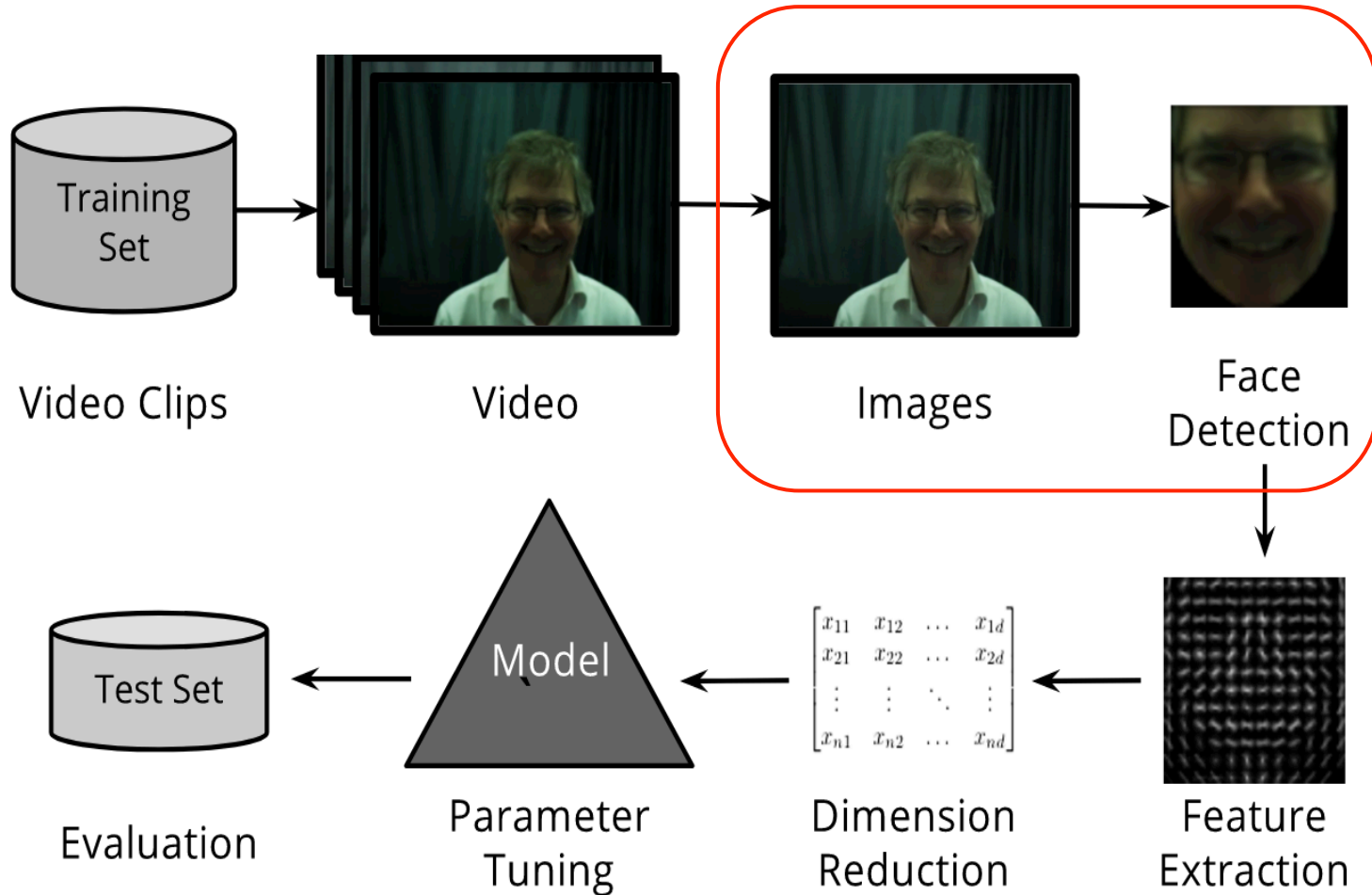
# Step 1: Split videos to training set and test set



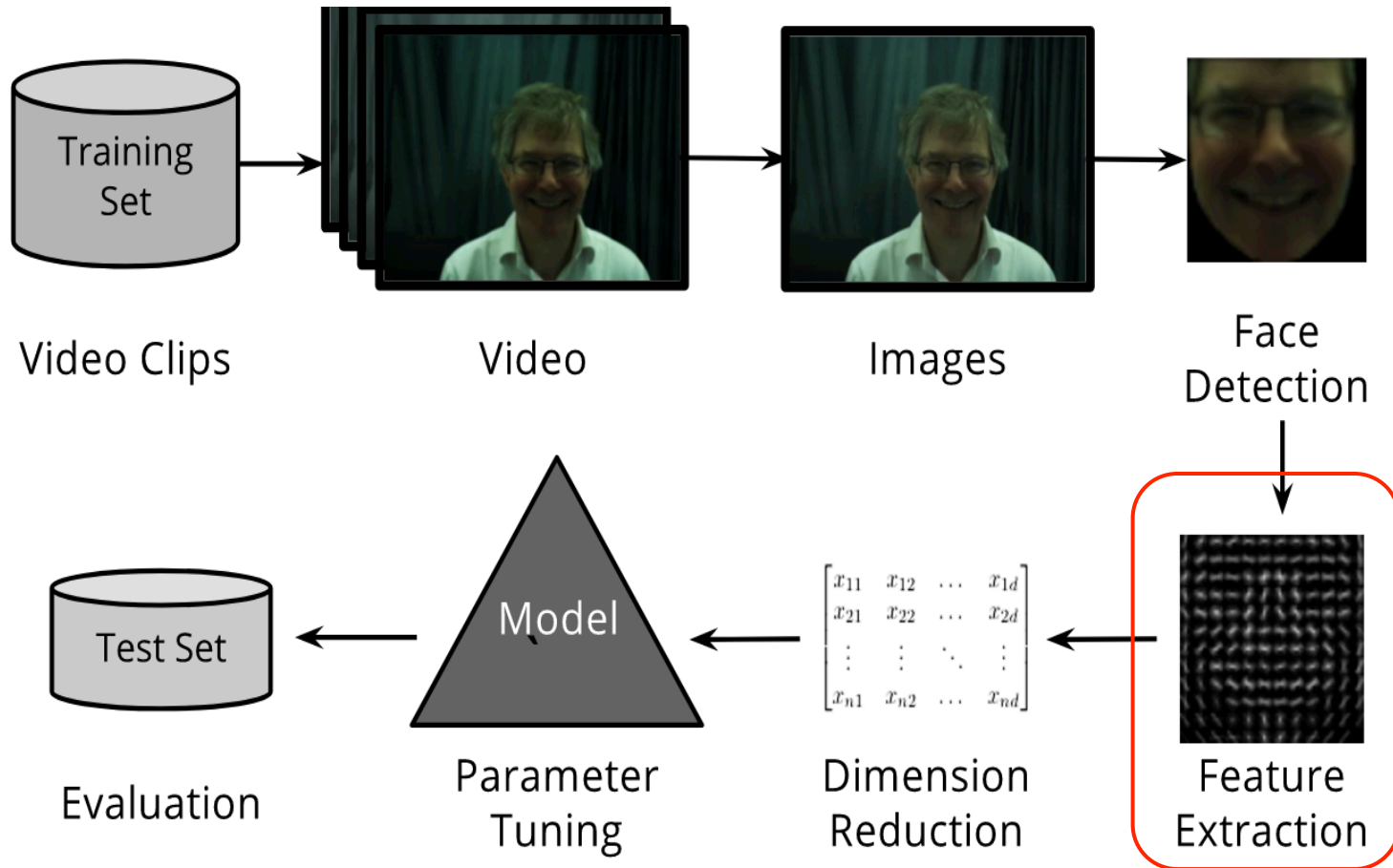
# Step 2: Downsample images



# Step 3: Get aligned face from each image

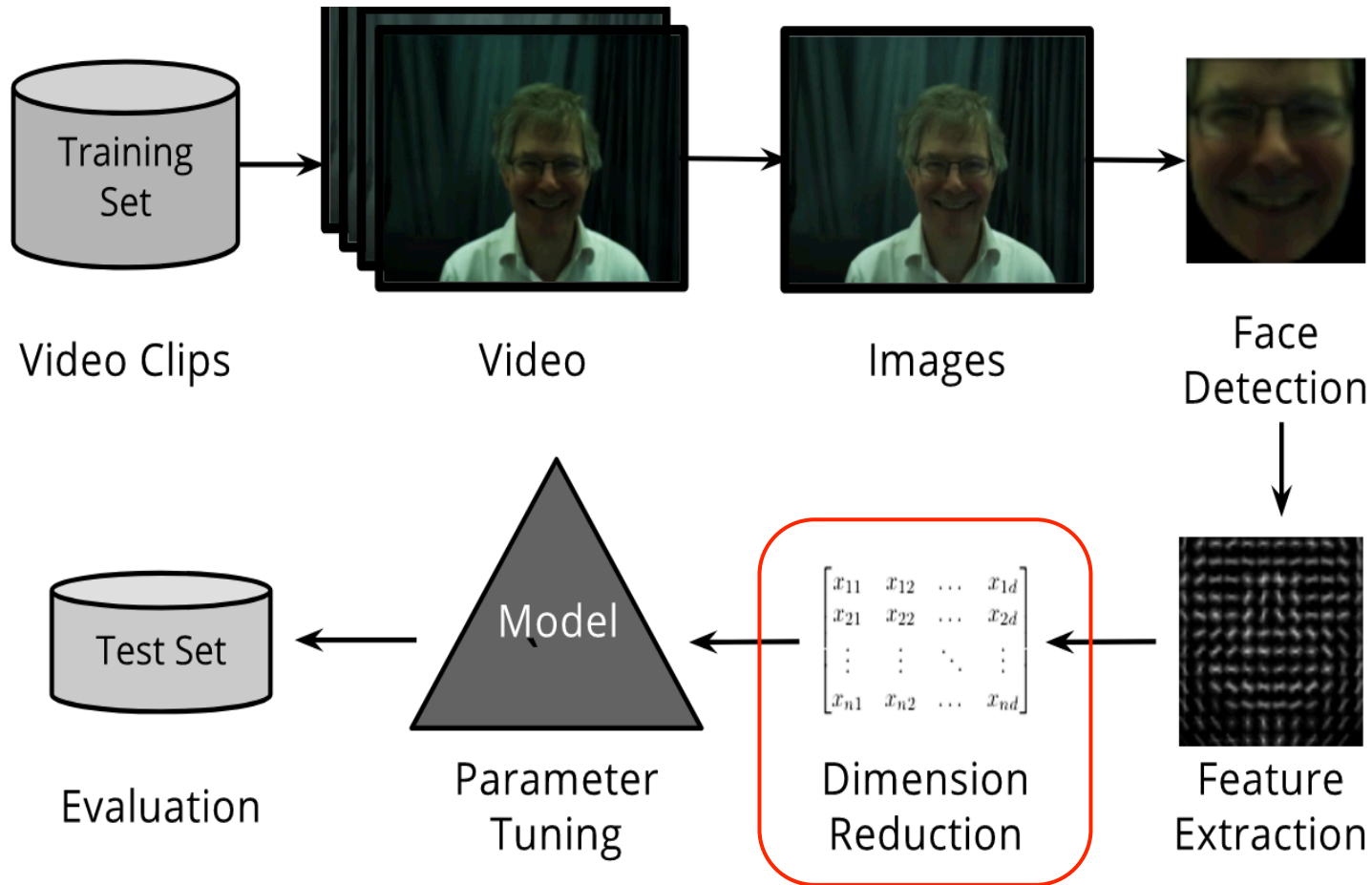


# Step 4: Extract three feature descriptors

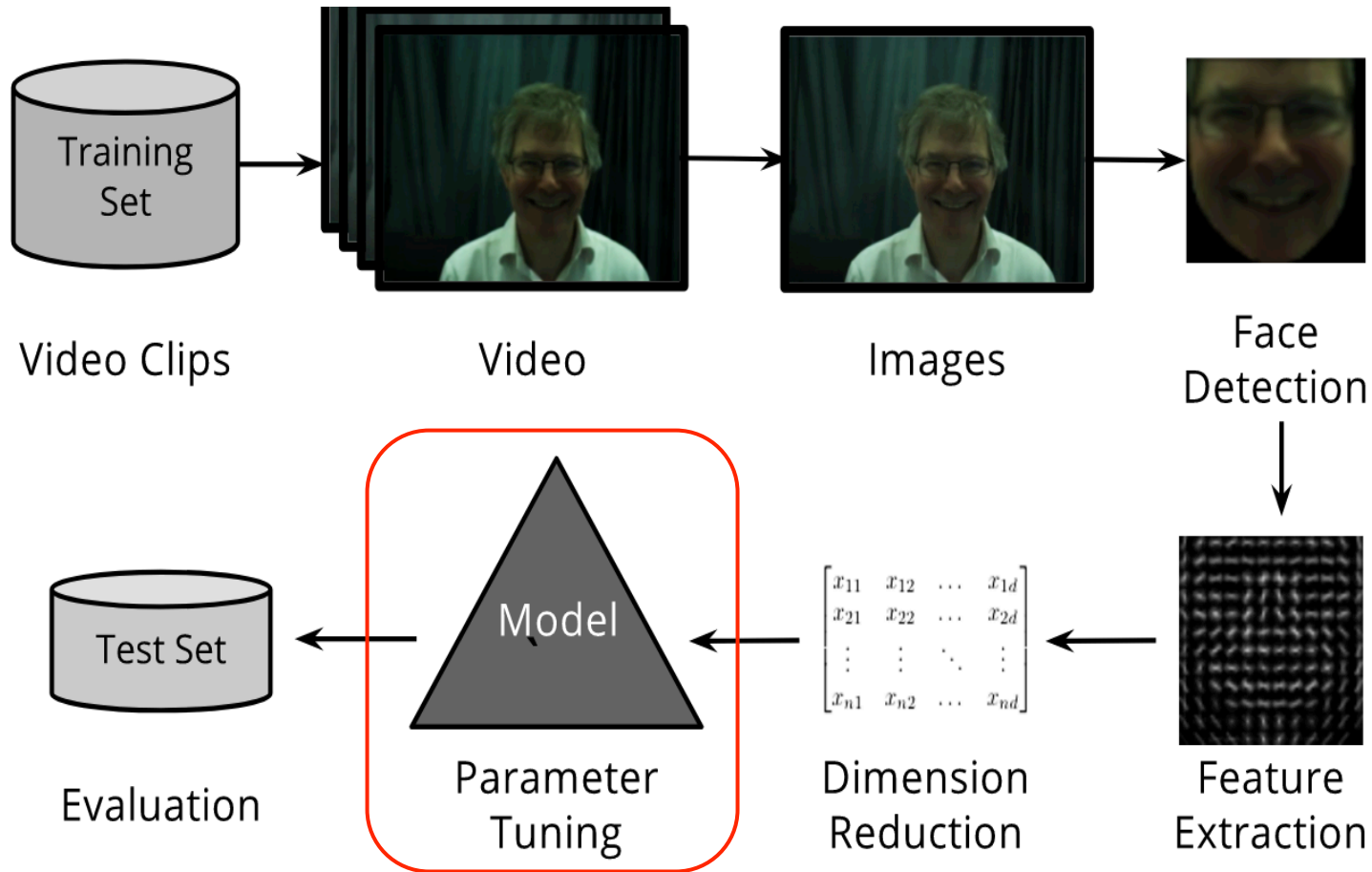




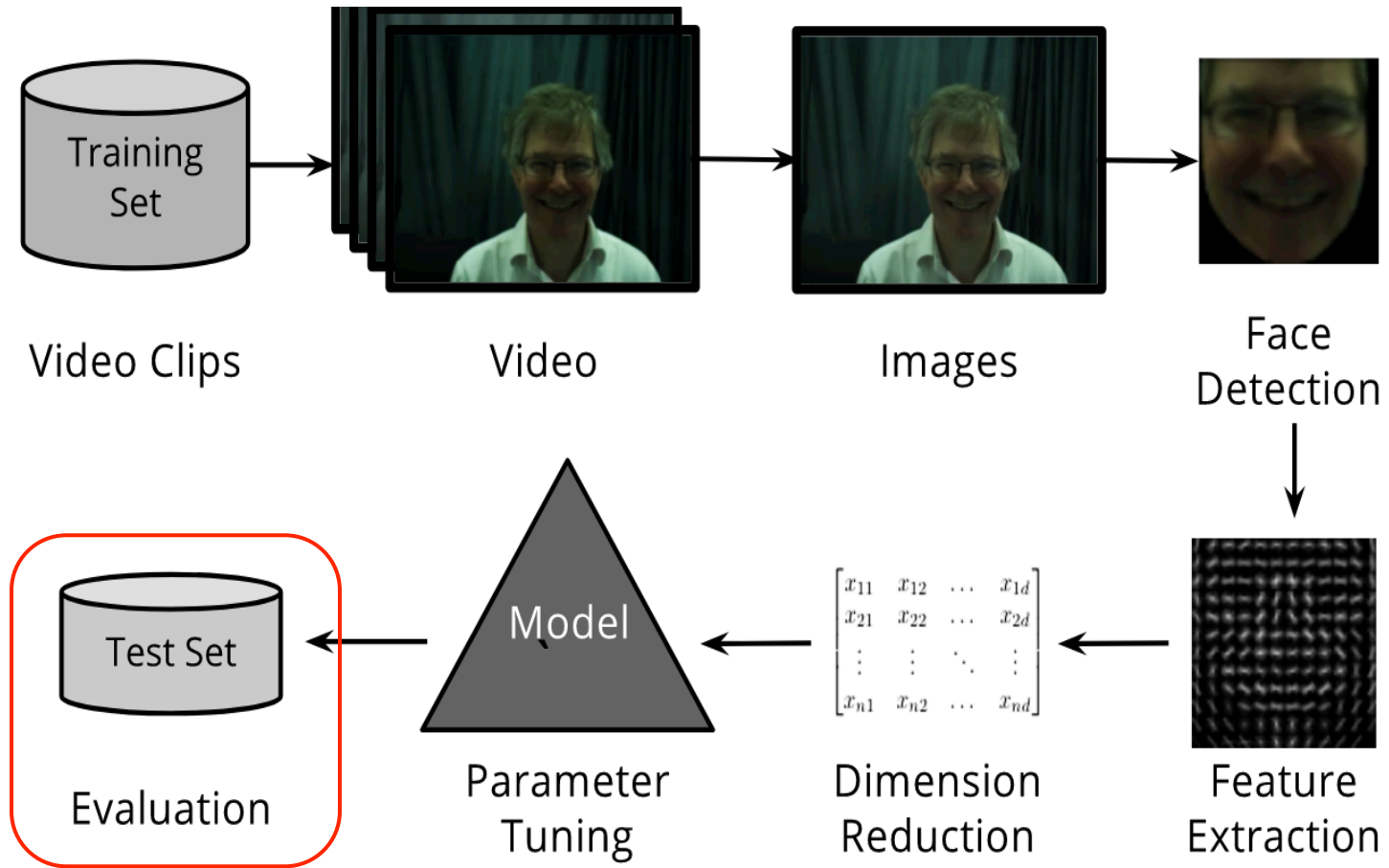
# Step 5: Reduce dimensions with PCA



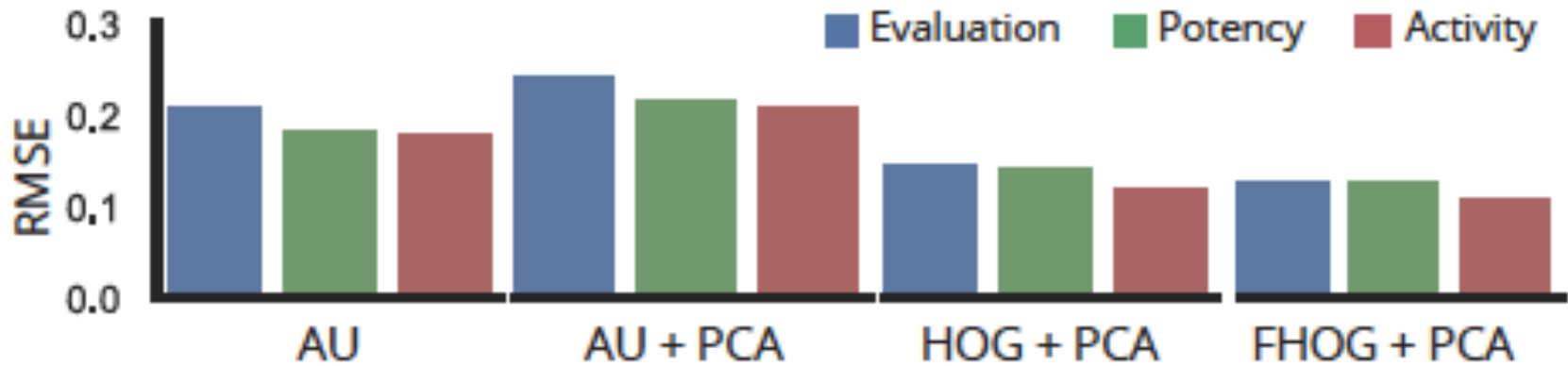
# Step 6: Train EPA models



# Step 7: Evaluate in the test set



# Results

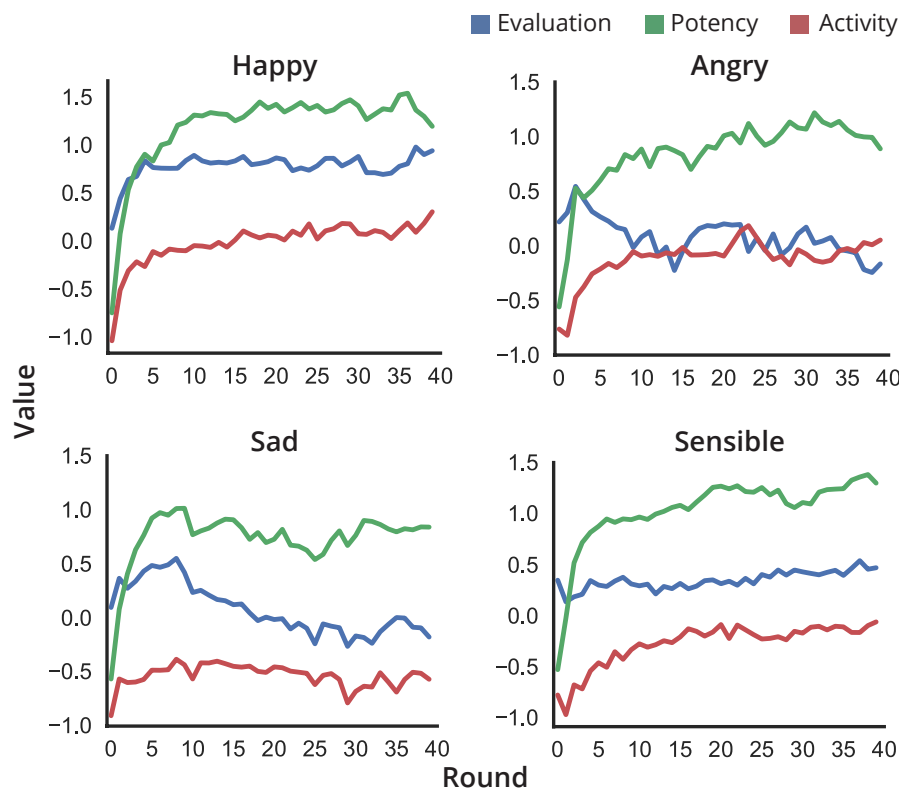


# BayesACT Simulations

- Input:
  - Actor/Object: user, avatar as “student”
  - Action: “talk to”
  - User’s emotion: facial EPA values
- Output:
  - user / avatar’s emotion in both EPA values and labels
- 40 rounds with values at every five seconds in the 18 test videos

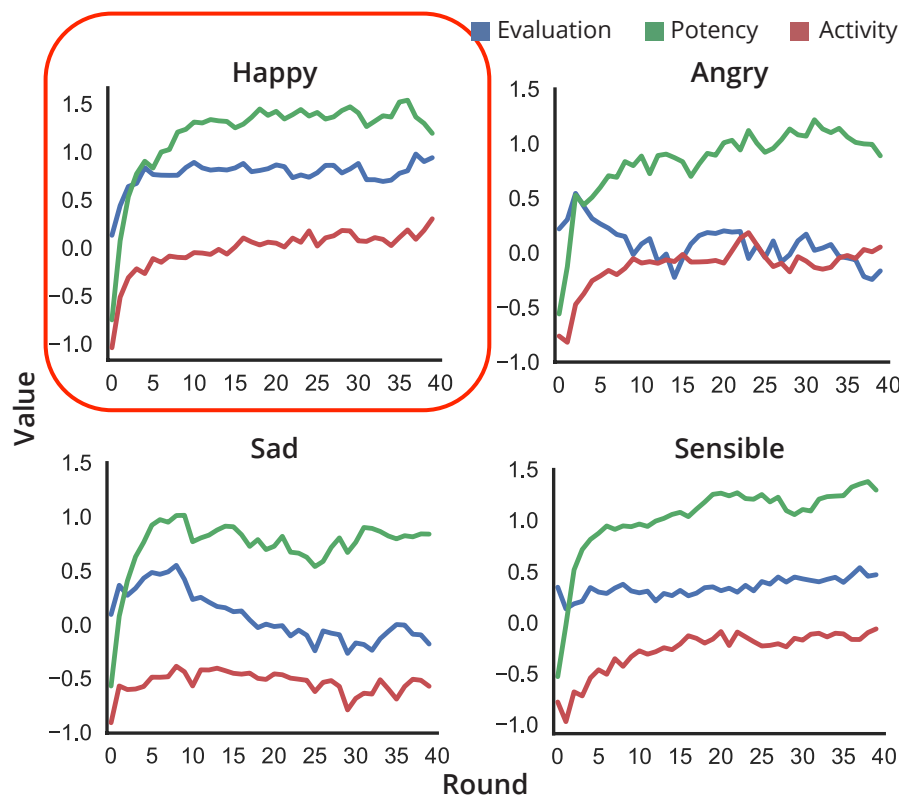
# BayesACT Simulations - Emotion Changes

Avatar's posterior estimate of user's emotion



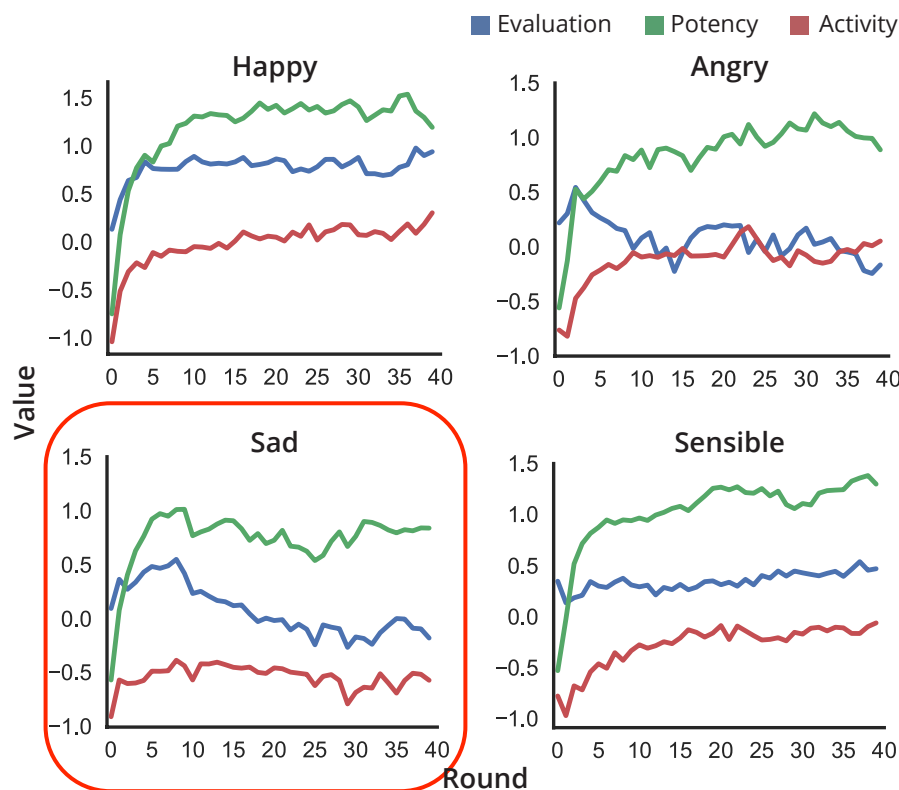
# BayesACT Simulations - Emotion Changes

Avatar's posterior estimate of user's emotion



# BayesACT Simulations - Emotion Changes

Avatar's posterior estimate of user's emotion





# BayesACT Simulations - Word Clouds

Top 15 labels that describe the user's feeling when talking to different emotional avatars



# Conclusion

- Create an automatic affect recognizer that continuously predicts user emotions in the EPA space
- Demonstrate the feasibility of using BayesACT to simulate interactions between a user and distinct behavior-styled avatars

# Thank you!



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