



# 2D Image Processing & Augmented Reality Winter Semester 2019/2020 Survey on Face Tracking with Deep Learning

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#### **Outline**



Face tracking is a computer vision task that involves tracking a specific number of landmarks on the face detected across all frames of a video.



Applications include Face analysis, Person identification, Activity recognition, Expression analysis, Face modeling etc.



It is a challenging problem as the videos can be captured in unconstrained conditions which may include illumination variations, large head poses, occlusions, etc.







#### **Outline**

- Image-based methods use models trained on still images in each frame.
- Video-based methods make use of temporal information to predict facial landmarks in each frame.
- Various approaches
  - Regression-based methods
  - Video-based face alignment
  - Encoder-Decoder Networks





# Recurrent Encoder-Decoder Network for Video-based Face Alignment

- Employs recurrent learning at both spatial and temporal dimensions.
- At spatial level, face alignment is done in a coarse-fine manner.
- At temporal level, Temporal-variant features such as pose and expression are separated from Temporal-invariant features such as facial identity.
- The network consists of 4 modules Encoder-Decoder, Spatial recurrent learning, Temporal recurrent learning and Supervised identity disentangling.





# **Dynamic Facial Analysis: From Bayesian Filtering to Recurrent Neural Network**

- Improvises on previous approaches for dynamic facial analysis that use Kalman/Particle filters.
- Bayesian filters require problem-specific design and tuning.
- This RNN based method avoids tracker engineering by learning from data (large data).
- CNN layers followed by recurrent layers as dense layers.
- Uses FC-RNN to exploit generalization from a pre-trained CNN.





# **Dual-Agent Deep Reinforcement Learning for Deformable Face Tracking**

- Exploits the fact that bounding box tracking and landmark detection tasks are dependent. The accuracy of the latter depends on how good the former is.
- The two tasks are modeled in a probabilistic manner by following a Bayesian model.
- The architecture consists of a *Tracking agent, Alignment agent* and *Communication channels* between the agents.





# Two-stream Transformer Networks for Video-based Face Alignment

- Two stream deep learning method to capture spatial as well as temporal information.
- Spatial stream captures information on still frames.
- Temporal stream captures temporal consistency information across successive frames. It is followed by a RNN to model the sequential information over consecutive frames.
- Finally, facial landmarks are determined by a weighted fusion of spatial and temporal streams.





#### Face Alignment Recurrent Neural Network

- Recurrent regression based approach
- Uses LSTM to exploit both spatial and temporal information.
- Spatial The predicted landmark location is used as basis for estimation in the next stage.
- Temporal The predicted landmark location is used as basis for estimation in the next frame.





#### **Comparison metrics**

- Following are some of the metrics that will be used to compare the various methods in this survey:—
  - Dataset used for training (In the wild vs constrained)
  - Evaluation metrics
  - Number of landmarks tracked
  - Kind of landmarks retrieved (2D or 3D)
  - Robustness to large pose variations, illumination changes





#### **Thank You**