

Video Future Frame Prediction of Bouncing Numerals

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Final Project

Machine Learning Fundamentals and Applications in ECE

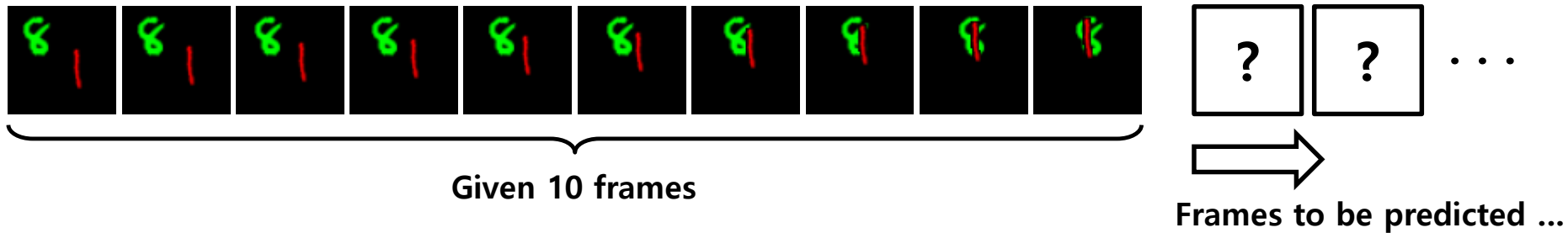
M2608.001300_001, 2019 Spring



Introduction

- Video Frame Prediction

- Given some frames, what would the next several frames look like?



- Our goal is to predict next upcoming 10 frames given past 10 frames
- We should consider both **spatial** and **temporal** information

Dataset

- Bouncing Numerals Dataset (not a public dataset)
 - 10,000 **training** sequences with 20 frames each.
 - 500 **validation** sequences with 20 frames each.
 - 500 **test** sequences with **10 frames** each.
 - You should predict the following 10 frames of each sequence.
- Data Utilization
 - Train your network **only using** training sequences
 - Find hyper-parameters, network architecture, ... with validation sequences
 - The ground truth of test sequences will not be available publicly.
 - You can check your score and current rank on the leaderboard of Kaggle

Evaluation (1/2)

- Evaluation Metric

- (Main Metric) Mean Absolute Error (MAE)

- $$\text{MAE} = \frac{1}{w \times h \times c} \sum_{w,h,c} |x_{w,h,c} - \hat{x}_{w,h,c}|$$
 - x : ground-truth frame, \hat{x} : predicted frame

- The MAE is averaged over predicted 10 frames

- (Additional Metric) Structural Similarity (SSIM)

- $$\text{SSIM}(x, y) = \frac{(2\mu_x\mu_y + c_1)(2\sigma_{xy} + c_2)}{(\mu_x^2 + \mu_y^2 + c_1)(\sigma_x^2 + \sigma_y^2 + c_2)}$$
 - SSIM will not be calculated by Kaggle.
 - You can compute SSIM with “pyssim” module on python

Evaluation (2/2)

- Sign up Kaggle, and come to project competition page
 - <https://www.kaggle.com/t/2b41fd37f29f434988c88984bfbfa816>
- You should also upload the prediction results of test sequences to ETL.
 - The format of the files **SHOULD** be as follows.
 - test_predicted/sequencexxx/framesxx.png
 - For example, the first predicted frame of sequence 123 should be test_predicted/sequence123/frames00.png
 - As a result, there must be 500 folders for each sequence, and each folder must have 10 images, from frames00 through frames09.

File description

- Unzip the given file
 - Unzip 2019Spring_ML_final.zip
- 2019Spring_ML_final/Data
 - Contains train, validation, and test sequences
 - Note that test sequences contain only **10 frames**
- 2019Spring_ML_final/data_utils.py
 - It provides train, validation, and test data loading functions
 - You can change those codes freely
- 2019Spring_ML_final/make_submission_file.py

Evaluation

- Make submission csv file for Kaggle
 - To submit your results to Kaggle, you need to create a csv file.
 - Run `make_submission_file.py`
 - Do not forget to change “`root_dir`” and “`save_dir`”
 - » `root_dir` : path to test_predicted folder
 - » `save_dir` : path where the csv file for submission will be saved
 - Do not change `make_submission_file.py`
 - If you have any questions or concerns, please let TA know.

submission.csv

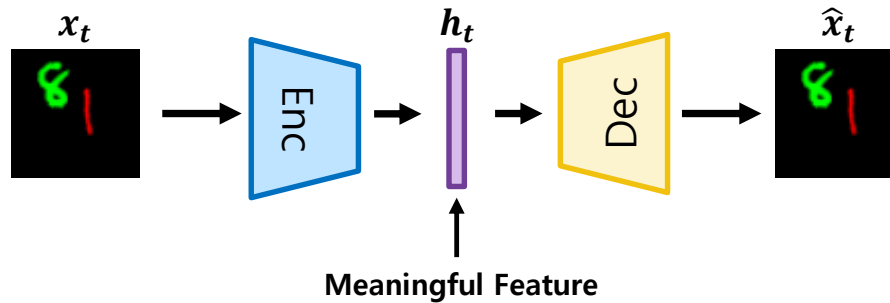
2584	0
2585	0
2586	45
2587	0
2588	0
2589	133
2590	0
2591	0
2592	180
2593	0
2594	0
2595	177
2596	0
2597	0
2598	162
2599	0
2600	0
2601	130
2602	0
2603	0
2604	108
2605	0
2606	0
2607	98
2608	0

⋮

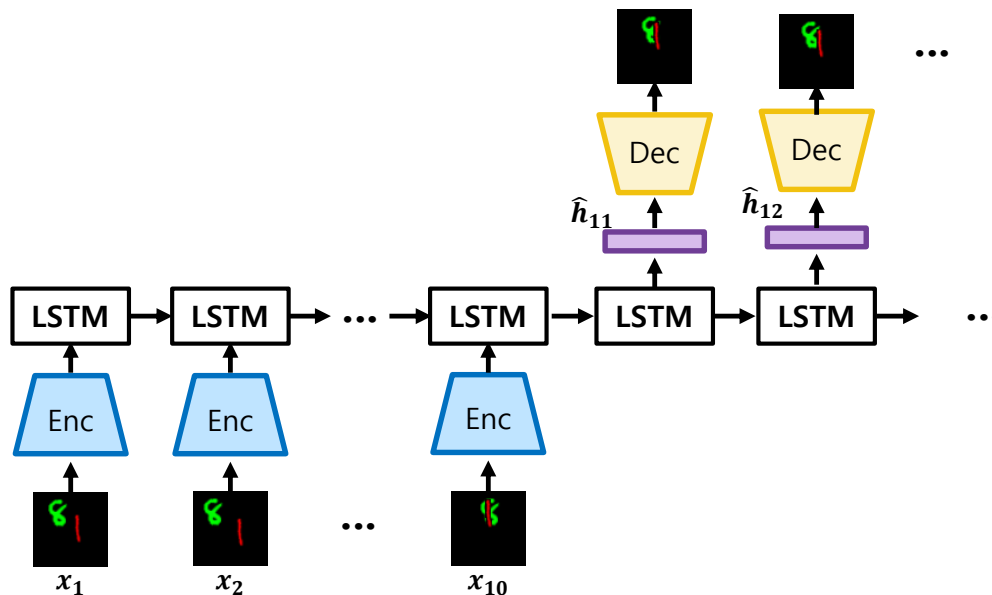
Hint (1/2)

- How was the baseline created?

- Step1 : Train an autoencoder



- Step2 : Each encoded feature is fed into LSTM sequentially



Loss functions

$$\text{Step1: } \|x_t - \hat{x}_t\|_2^2$$

$$\text{Step2: } \sum_{t=11}^{20} \|h_t - \hat{h}_t\|_2^2$$

Hint (2/2)

- Hidden Markov Model (HMM) can replace LSTM
 - HMM will soon be covered in class.

Three basic problems for HMM

problem 1:

given observation sequence $O = O_1 O_2 \cdots O_T$ and model $\lambda = (A, B, \pi)$,
how do we **efficiently compute** $P(O | \lambda)$, the probability of the observation sequence, given the model?

problem 2:

given observation sequence $O = O_1 O_2 \cdots O_T$ and model λ ,
how do we **choose a corresponding state sequence** $Q = q_1 q_2 \cdots q_T$ which is optimal in some meaningful sense?

problem 3:

how do we **adjust model parameters** $\lambda = (A, B, \pi)$ to maximize $P(O | \lambda)$?

Timeline

- Timeline (시간, 장소 반드시 확인)
 - 5/20: Project Out
 - 5/22: Project 소개 (7pm, 모랩)
 - 6/5: Project Q&A (by Jungbeom Lee, 7pm, 301동 102호)
 - 6/10: Project Q&A (by Eunji Kim), Final Q&A (by all TAs), 기존 수업시간
 - 6/12: Final Exam (3:30 pm, TBD)
 - **6/14: Final Exam Claim (TBD)
 - ** 6/17: Project Due (ETL에 file 업로드 꼭)
- TA schedule
 - Jungbeom Lee attends the conferences in the US from June 7 to June 23.
 - Quick answers to your project questions may be difficult.
 - Eunji Kim can give you a quick answer, but if not, Jungbeom Lee may be able to answer in **12:00 PM ~ 2:00 PM Korea time** (online).