## Deep Learning for Electronic Health Record Prediction

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### Outline

- Problem Description
  - Task Description
  - Data Description
- 2 Data Preprocess
- 3 Preliminary Result and Future Work

#### **Feature**

- At day0, every people are given injected a virus. And given the electronic health record of several days before day0, we want to know these people's symptom after day0.
- real value time series data stored in HDF5 format

#### **Feature**

- 19 people, every person has a time series data of length about 200000, which is a record of 3 days.
- time series record contains TEMP.csv:temperature in Celsius EDA.csv:electrodermal activity in microsiemens BVP.csv:blood volume pressure ACC.csv:accelerometer,3-axis,acceleration(1/64g) IBI.csv:inter beat interval in seconds HR.csv:hearth rate from BVP in Hz

 $\bullet~1\times4$  vector for 4 days' shedding titer after day0

mean	[0.22, 1.74, 2.05, 1.63]
variance	[0.91, 2.78, 2.44, 2.17]
median	[-0.5, 1.45, 2.7, 1.7]
min	[-0.5, -0.5, -0.5, -0.5]
max	[2.2, 4.7, 4.7, 4.2]

## **Data Preprocess**

#### Data Augmentation

- upsampling to augment the data
- normalize the feature
- get 374 data, each with a  $20000 \times 7$  feature, and a  $1 \times 4$  label.

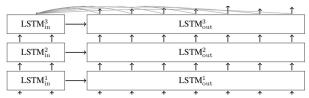
## Already Done

- Read the data from the dataset
- Data Preprocess
- Implementing seq2seq, 1D CNN, 1D ResNet and get a preliminary result

# Already Done

#### Model

• Seq2Seq(depth = 3)



- 1D CNN (10 layers)
- 1D ResNet (8 layers)

## Already Done

#### Preliminary Result

• 10 fold cross validation

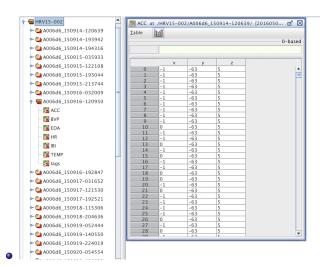
	without Normalization on data feature					
		Loss(MAE)	Correlation	Num of Parameters		
•	Seq2Seq	$1.33 \pm 0.12$	$0.42 \pm 0.05$	976		
	1D CNN	$1.74 \pm 0.70$	$0.26 \pm 0.34$	1175		
	1D ResNet	$1.25 \pm 0.12$	$0.42 \pm 0.06$	160652		

	with Normalization on data feature				
		Loss(MAE)	Num of Parameters		
•	Seq2Seq	$1.33 \pm 0.11$	976		
	1D CNN	$1.28 \pm 0.16$	1175		
	1D ResNet	$1.60 \pm 0.53$	160652		

#### Future work

- view it as a semi supervised problem, use the sequence before the labels' time as the feature
- better ways for data preprocessing
- other ways for data augmentation
- try some other kinds of networks, such as wavenet
- try attention

#### **Feature**



Label

subject_id	studyDate	studyDay	sheddingTiter
HRV15-002	2015/9/18	Day1	1.7
HRV15-002	2015/9/19	Day2	3.95
HRV15-002	2015/9/20	Day3	4.45
HRV15-002	2015/9/21	Day4	3.7
HRV15-003	2015/9/18	Day1	0
HRV15-003	2015/9/19	Day2	0
HRV15-003	2015/9/20	Day3	0
HRV15-003	2015/9/21	Day4	0
HRV15-004	2015/9/18	Day1	0.7
HRV15-004	2015/9/19	Day2	1.45
HRV15-004	2015/9/20	Day3	2.7
HRV15-004	2015/9/21	Day4	1.7
HRV15-005	2015/9/18	Day1	0.7
HRV15-005	2015/9/19	Day2	1.7
HRV15-005	2015/9/20	Day3	2.7