Three-dimentional Convolution Neural Network based Encrypted Traffic Classifier for Wireless Communications

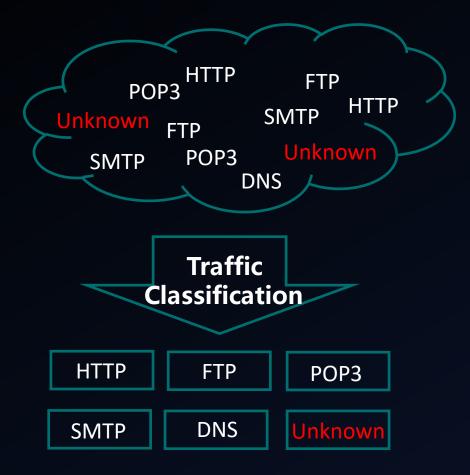
Jing Ran, Yexin Chen, Shulan Li Beijing University Of Posts And Telecommunications

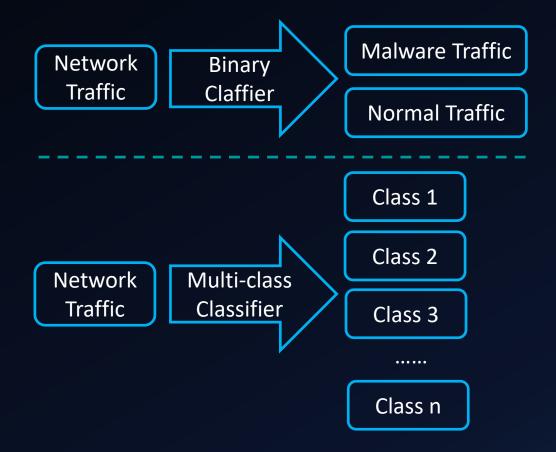
## Outline

- Traffic Classification
- Preprocessing
- 3D CNN Classifier
- Simulation

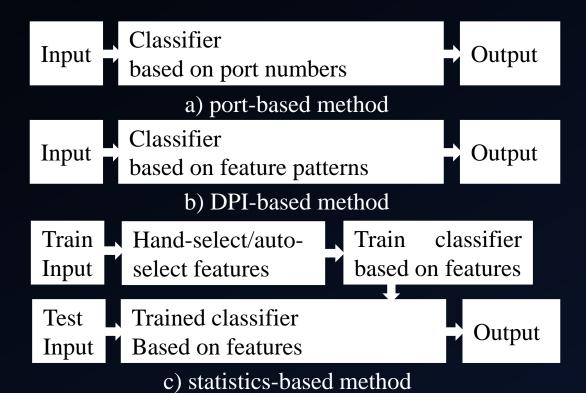
I Traffic Classification

#### Traffic Classification





#### Traffic Classification

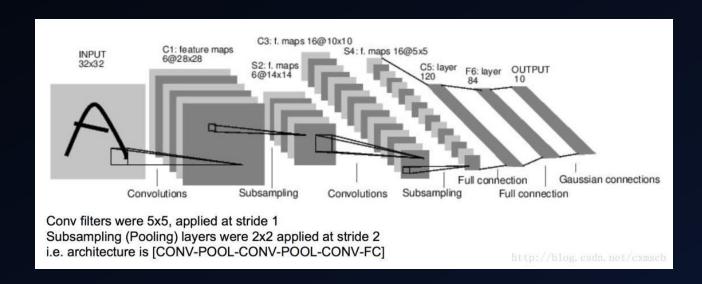


Methods of network traffic classification

II Preprocessing

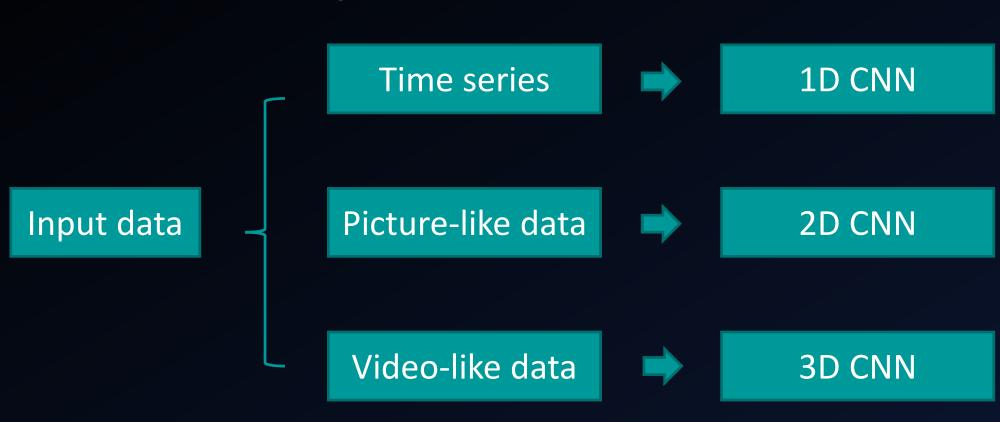
## Preprocessing

- Convolution Neural Network
  - Convolutional layer
  - Pooling layer
  - Fully-connected layer
  - Output layer
- Main Application
  - Computer Vision
- Features
  - Sparse connection
  - Weight sharing
  - Multi feature graph

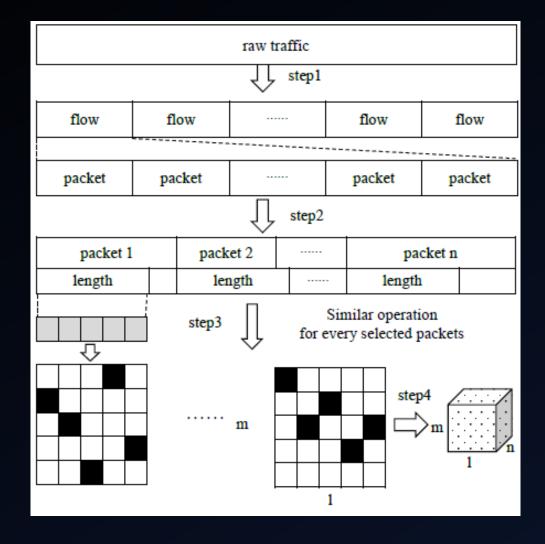


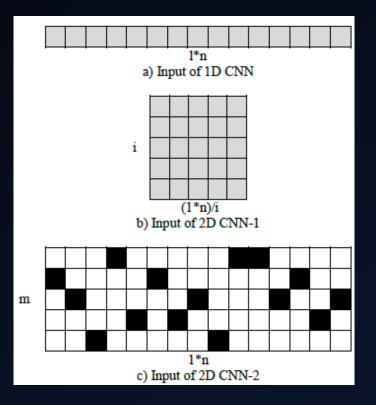
LeNet-5

# Preprocessing



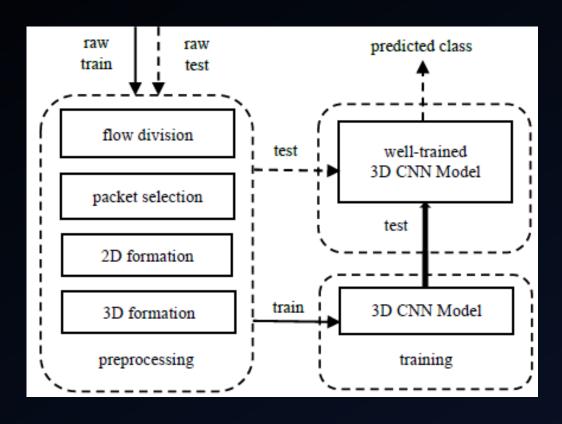
# Preprocessing

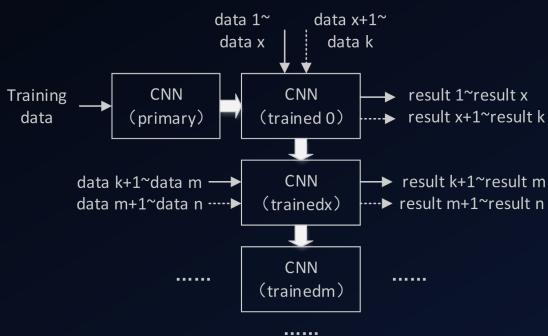




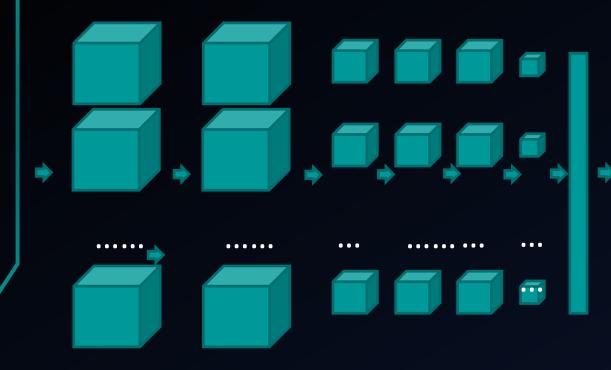
III 3D CNN Classifier

## 3D CNN Classifier





# 3D CNN Classifier



layer	type	filter	output
1	conv+Relu	3*3*3*32	28*28*16*32
2	conv+Relu	3*3*3*32	28*28*16*32
3	max pooling	2*2*1	14*14*16*32
4	conv+Relu	3*3*3*64	14*14*16*64
5	conv+Relu	3*3*3*64	14*14*16*64
6	max pooling	2*2*2	7*7*8*64
7	dense	-	1024
8	dense	-	10
9	softmax	-	10

result

IV Simulation

#### Simulation

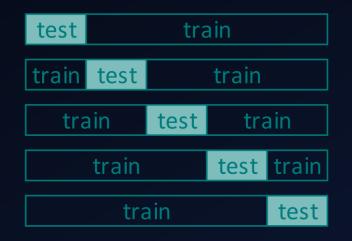
Normal Traffic					Malware Traffic				
BitTorrent	Gmail	MySQL	Outlook	Weibo	Htbot	Neris	Nsis-ay	Shifu	Virut

Dataset : part of USTC-TFC2016(raw)

real traffic collected by CTU researchers simulated traffic collected by IXIA BPS

Experimental frameworks: Keras and Tensorflow

#### Five-fold cross-validation



## Simulation

#### **Parameters Selection**

#### Learning rate & Optimizer:

learning rate optimizer	0.001	0.0001
GradientDescentOptimizer	98.73%	86.02%
AdamOptimizer	10.52%	99.77%

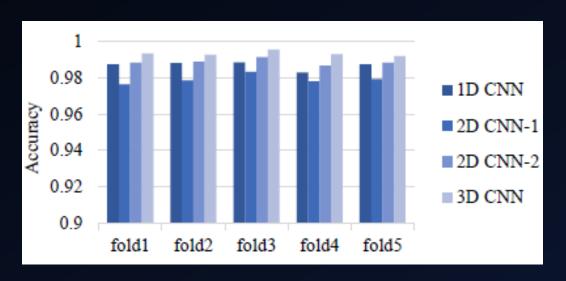
Packets per flow: 16

Training time: 40 epoch

Cost function: cross entropy

Mini-batch size: 50

#### Accuracy



CNN-1: cut according to fixed length

CNN-2: one-hot encoding of 1D data

## Simulation

#### **Evaluation Criterion**

Classified as →	X	$\overline{X}$	
X	ТР	FN	
$\overline{X}$	FP	TN	

$$Accuracy = \frac{TP + TN}{TP + TN + FP + FN}$$

$$Precision = \frac{TP}{TP + FP}$$

$$Recall = \frac{TP}{TP + FN}$$

$$F1 = 2 \frac{Precision \times Recall}{Precision + Recall}$$

class	precision	recall	Fl
BitTorrent	1	0.9962	0.9981
Gmail	0.9898	0.9898	0.9898
MySQL	1	1	1
Outlook	0.9888	0.9925	0.9906
Weibo	1	1	1
Htbot	0.9954	1	0.9977
Neris	0.9433	0.9659	0.9545
Nsis-ay	1	0.9952	0.9976
Shifu	1	0.9971	0.9985
Virut	0.9515	0.927	0.9391

Thank you!