



UNIVERSITY OF  
CAMBRIDGE



北京邮电大学

Beijing University of Posts and Telecommunications

# Edge Network opens the door of Heterogeneous Network Federation

Yuchao Zhang  
Oct. 7<sup>th</sup>, 2021

# Self-introduction

## Yuchao Zhang



Associate Professor

School of Computer Science (National Pilot Software Engineering School)

Beijing University of Posts and Telecommunications.

I'm the Director of Network and Software Communication Center, CS, BUPT.

I'm the Director of Academic Committee in Joint Laboratory of BUPT and Chuangcache.

B.S. degree (in CS) from Jilin University in 2012

Ph.D. degree (in CS) from Tsinghua University (Advisor: Prof. Ke Xu) in 2017

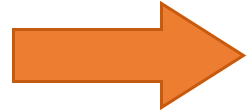
Postdoc in the Hong Kong University of Science and Technology in 2018 (with Prof. Kai Chen)

Visiting Researcher in Hong Kong Huawei 2012 Theory Lab in 2019.

Visiting Scholar in Cambridge University since Sept. 2021 (with Prof. Jon Crowcroft).

<http://yuchaozhang.weebly.com/>

# Outline



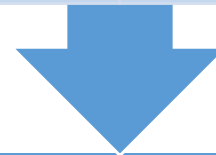
## 1. Cache on the edge

Exponentially growing data amount **VS.** Limited storage on the edge



## 2. Transmission among the edges

Large scale transmission **VS.** Real-time traffic requirements

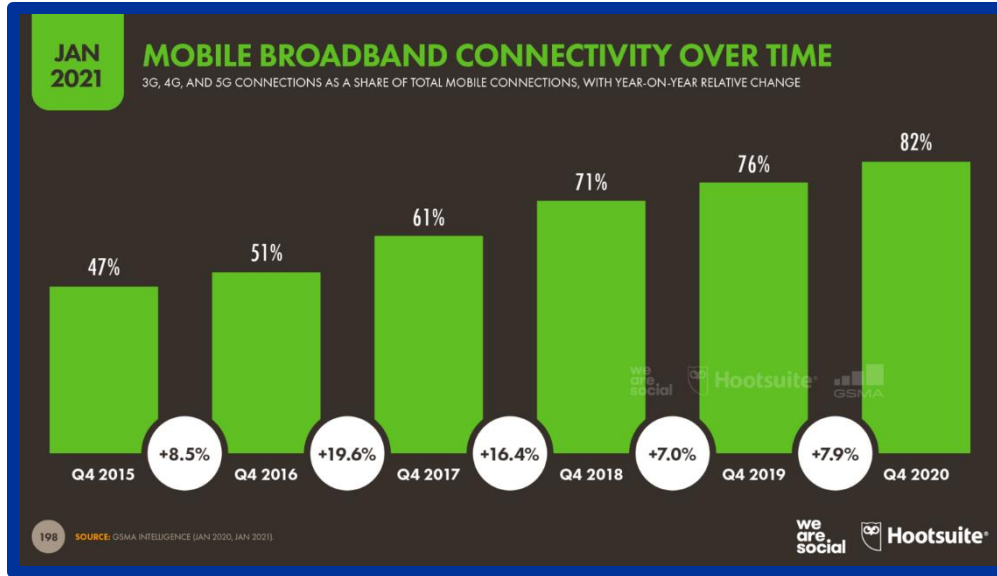


## 3. Federation of multiple edges

Strict data privacy **VS.** Multiple edge network collaboration



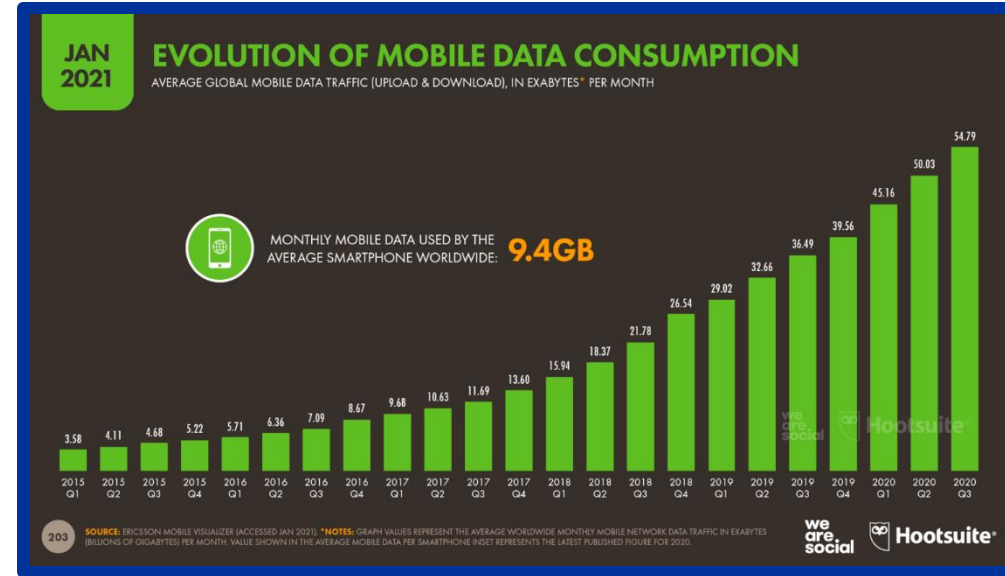
# Data from edges are ever increasing



## TIME

- **2015** **47%**
- **2020** **82%**
- **2015->2020** **+74.39%**
- **FUTURE** **?**

**3G,4G, AND 5G CONNETCION AS SHARE OF TOTAL MOBILE CONNETIONS, YEAR-ON-YEAR RELATIVE CHANGE**

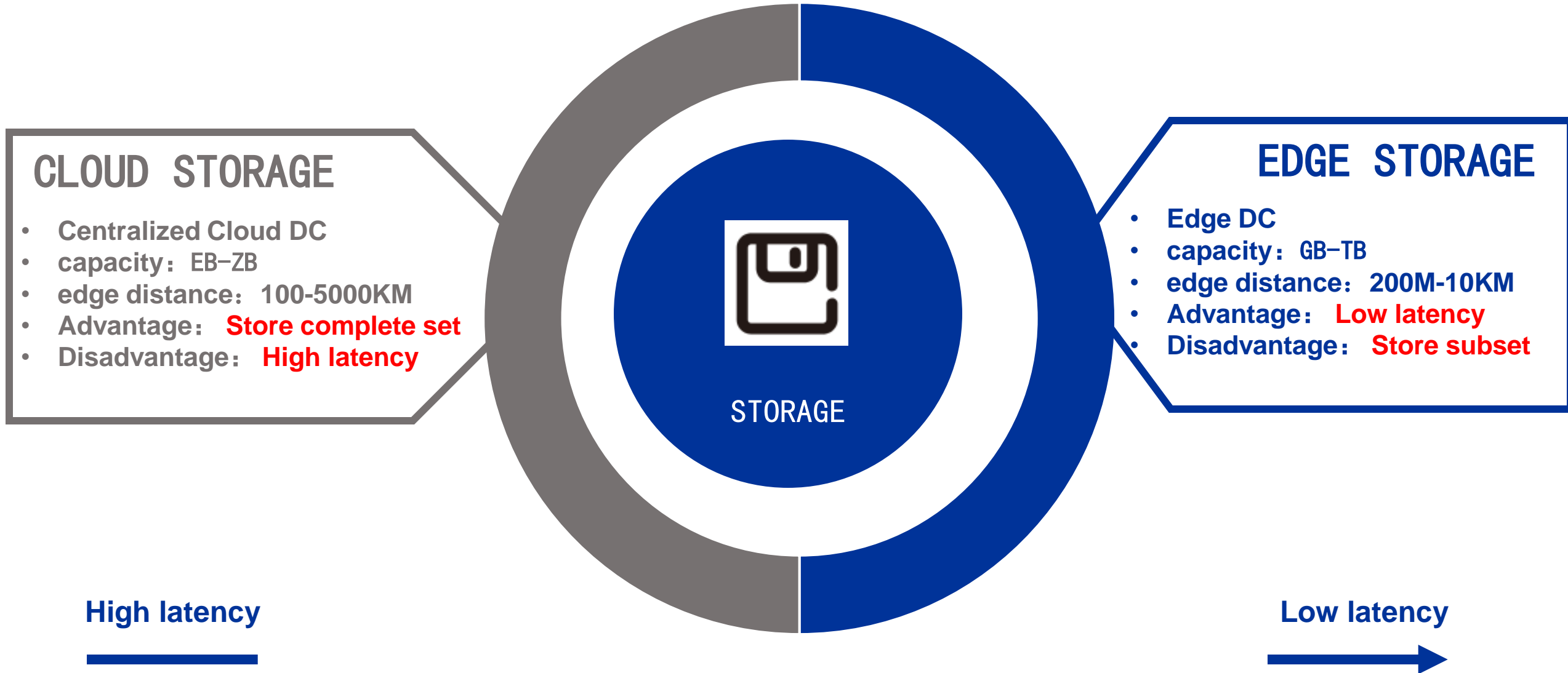


## DATA

- **2015 Q1** **3.58 EB**
- **2020 Q3** **54.79 EB**
- **2015->2020** **+51.21 EB**
- **FUTURE** **?**

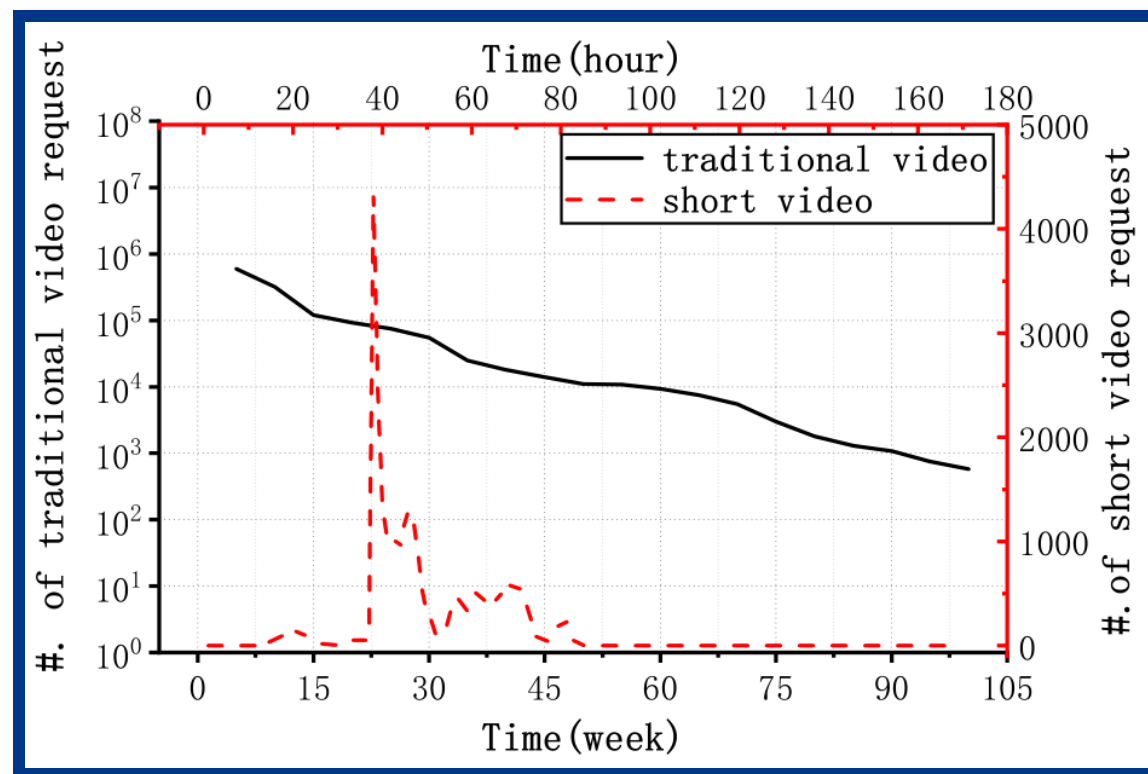
**AVERAGE GLPBAL MOBILE DATA TRAFFIC (UPLOAD&DOWNLOAD), IN EXABYTES\*PER MONTH**

# 1. Cache on the edge



# Characteristics

Comparison	All Online Video Platforms	One Short Video Platform	Times
#. of videos viewed per day Toutiao (2019); Ibizuo (2018)	46 million (from top 10 in China)	10 billion (only from Toutiao)	217
Length of video up-	0.1 million minutes (from all TV	340 million minutes	2850
<b>Daily active user</b>		<b>&gt;400million</b>	
		<b>&lt;100million</b>	
Growth rate of use time QuestMobile (2019)	-12%	+521.8%	×



## Explosive video and user **quantity**

- *It is difficult to guarantee the low latency under limited cost of server.*

## Popularity changes **quickly**

- *It is difficult to predict popular short videos*

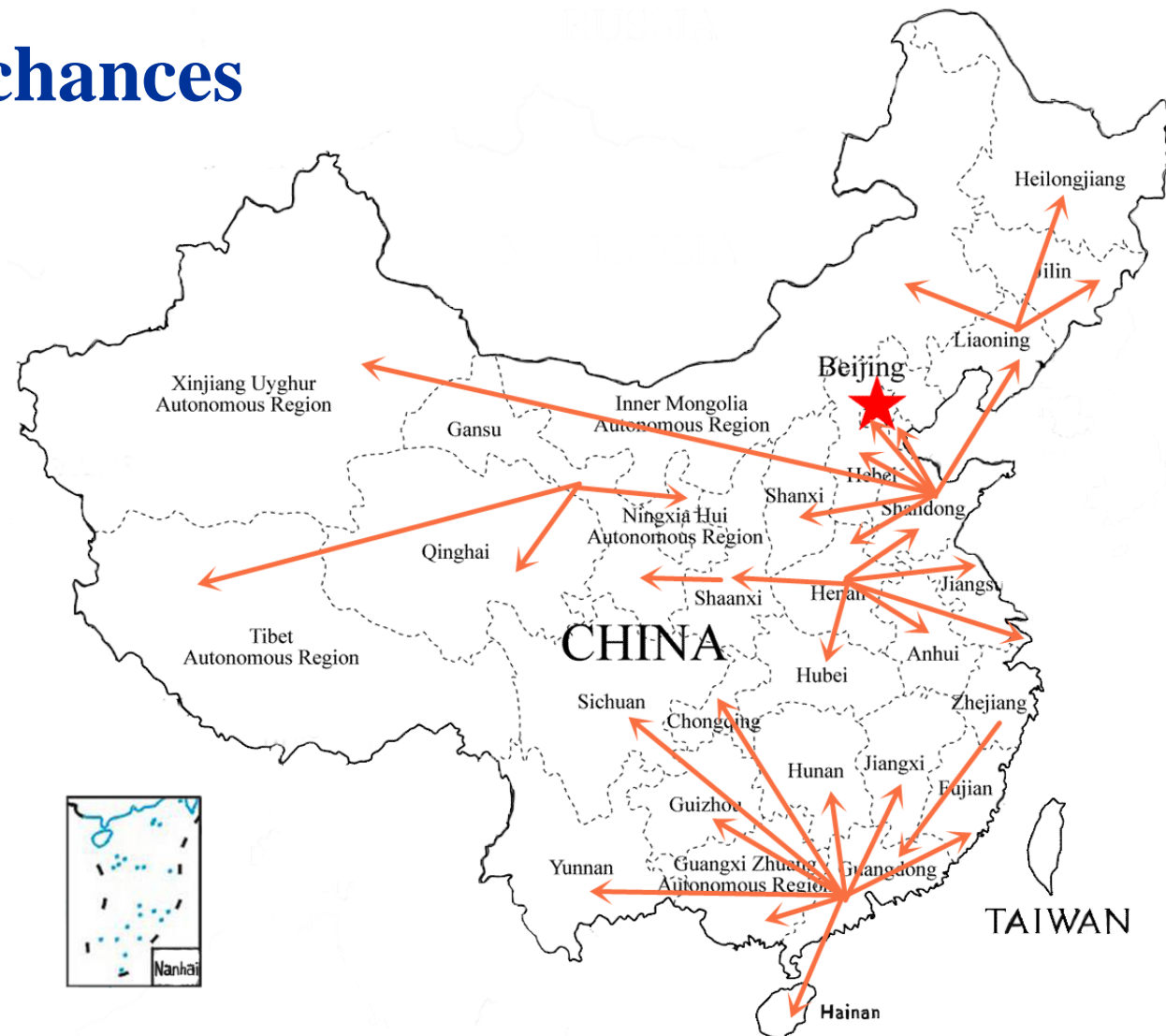


# 1.1 Potential opportunity

## Predicting popular content -- chances

- Hot Topic Propagation
- Quantity Number
- Time Sensitivity

Dataset	Cache Size	Access #.	Server #.	Video #.
City 1	3.96T	5,323,508	30	1,424,564
City 2	5.12T	9,202,618	72	1,813,058
City 3	2.51T	3,051,059	10	876,058
City 4	2.39T	2,765,419	21	843,219
City 5	2.48T	2,828,645	6	862,806
...	...	...	...	...
Total	78.75T	105,231,883	488	12,089,887

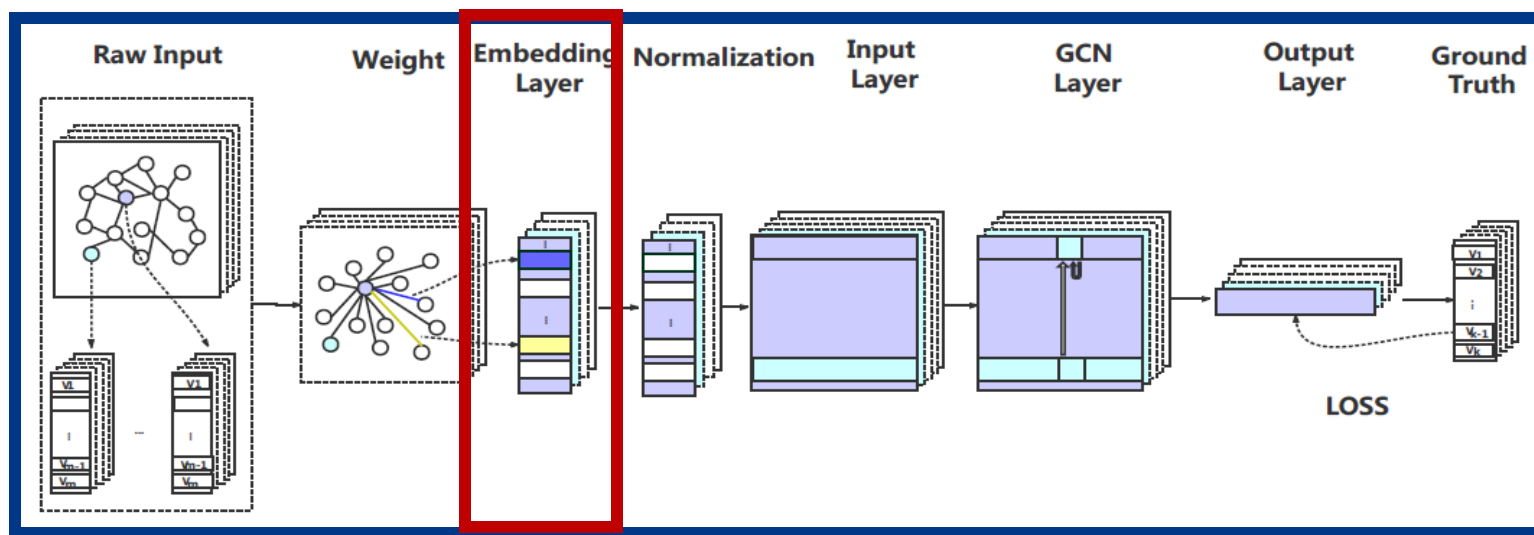




# 1.1 GraphInf architecture

## Predicting popular content -- GraphInf

We propose *GraphInf* based GCN network eight Layers, designing an embedding layer to capture the propagation relationship of popular videos.



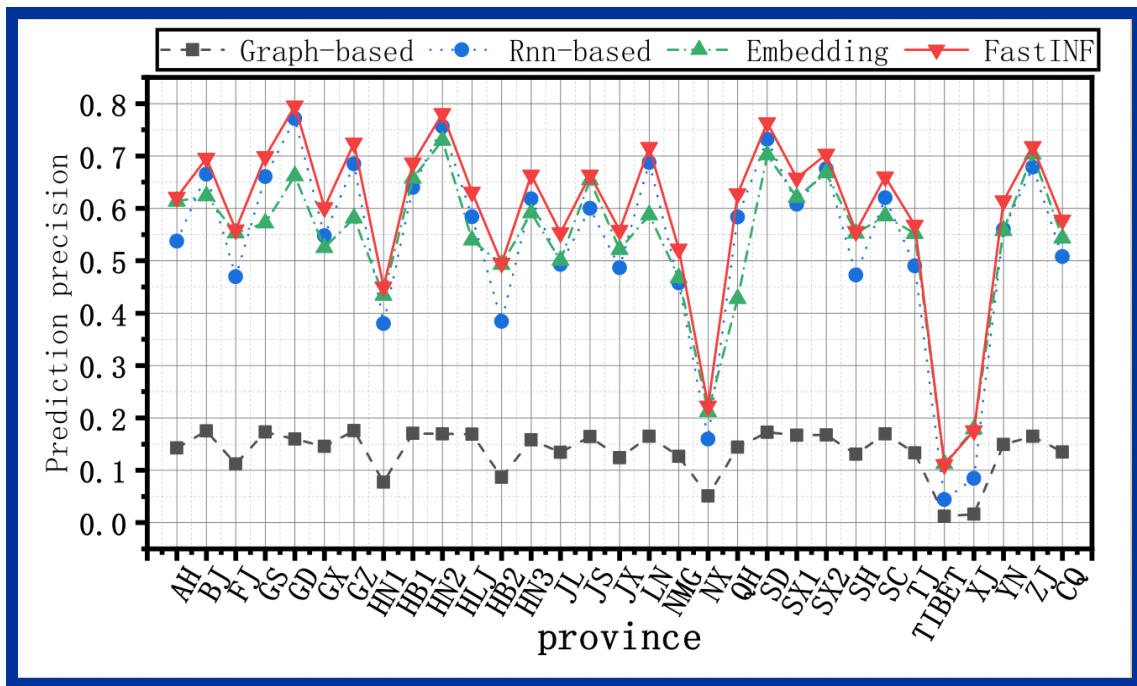
### Input

- Time slice
- Regions/provinces
- Hot items

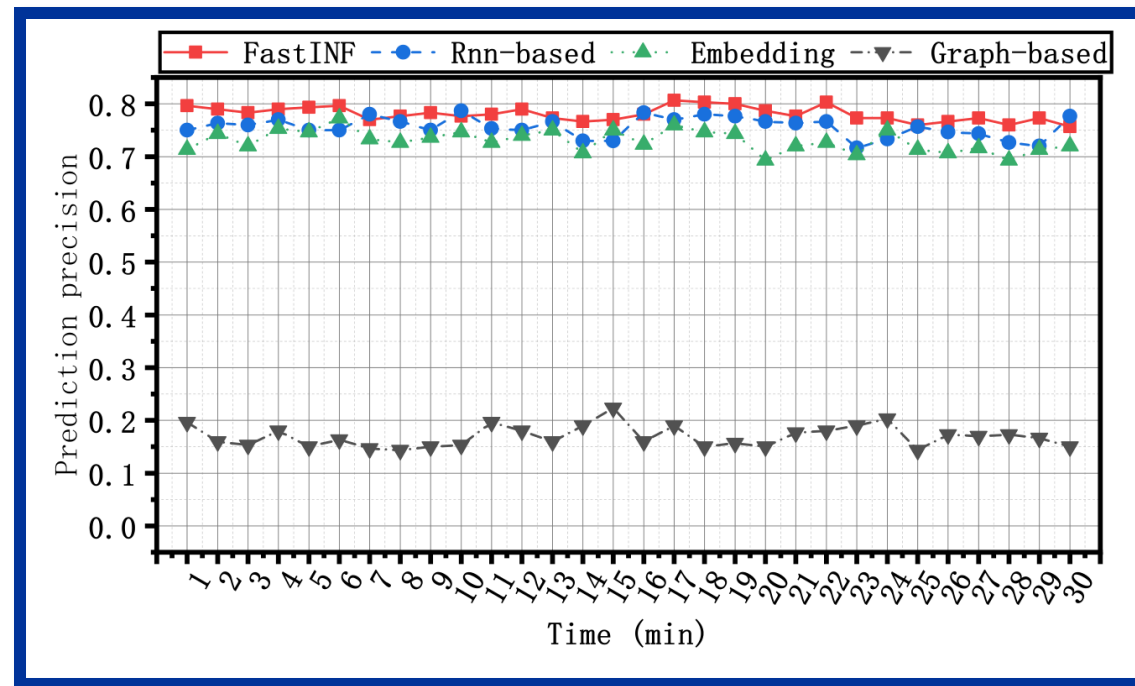
### Output

- Hot videos per region

# 1.1 GraphInf evaluation



**31 PROVINCES**

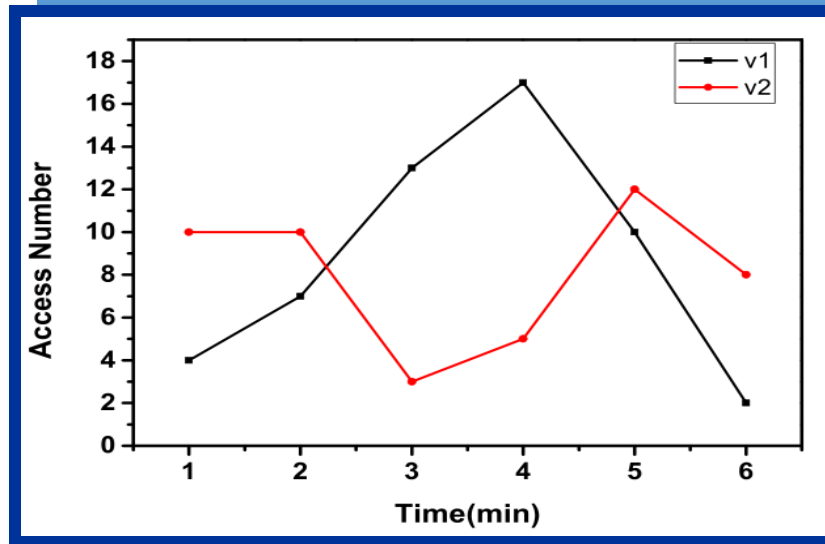


**Henan**

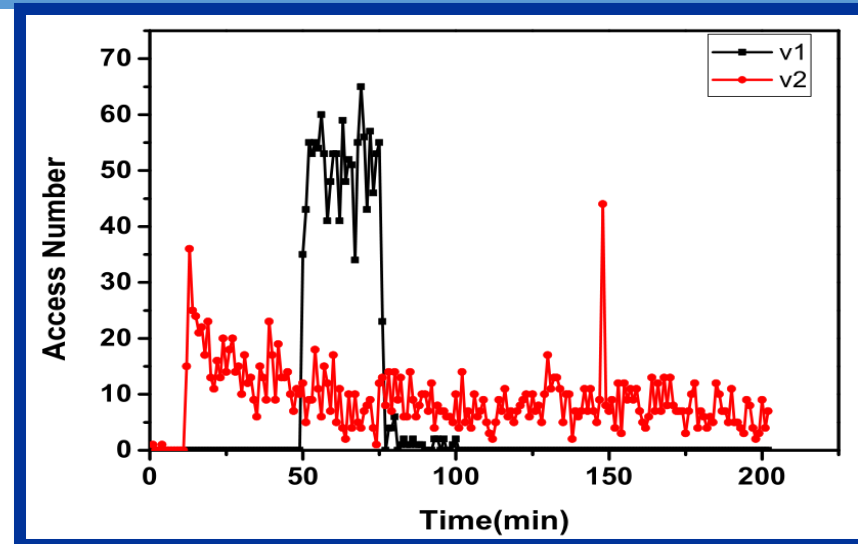
# 1.2 How to evict?

Prior solutions (such as LFU and DeepCache<sup>1</sup>) work well in traditional centralize-controlled CDNs, but become invalid in the emerging short video networks due to:

Non-stationary video access pattern.



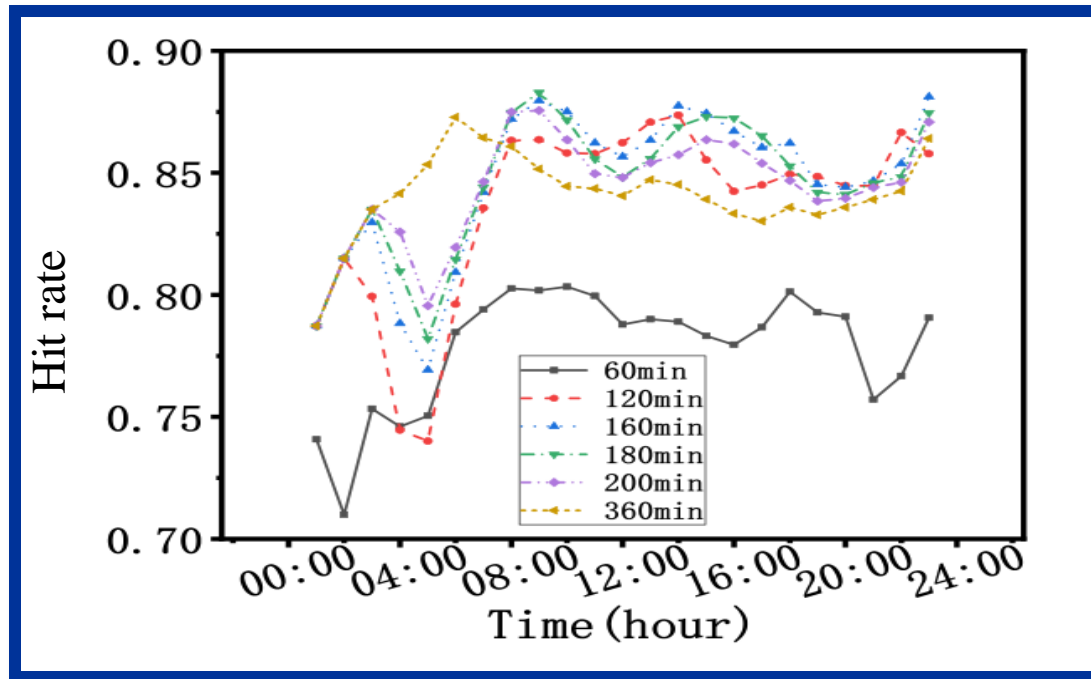
Temporal and spatial video popularity pattern.



1. J. Qiu, J. Tang, H. Ma, Y. Dong, K. Wang, and J. Tang, "Deepinf: Social influence prediction with deep learning," in Proceedings of the 24th ACM SIGKDD International Conference on Knowledge Discovery & Data Mining. ACM, 2018, pp. 2110–2119.

# 1.2 How to evict?

## Replacement



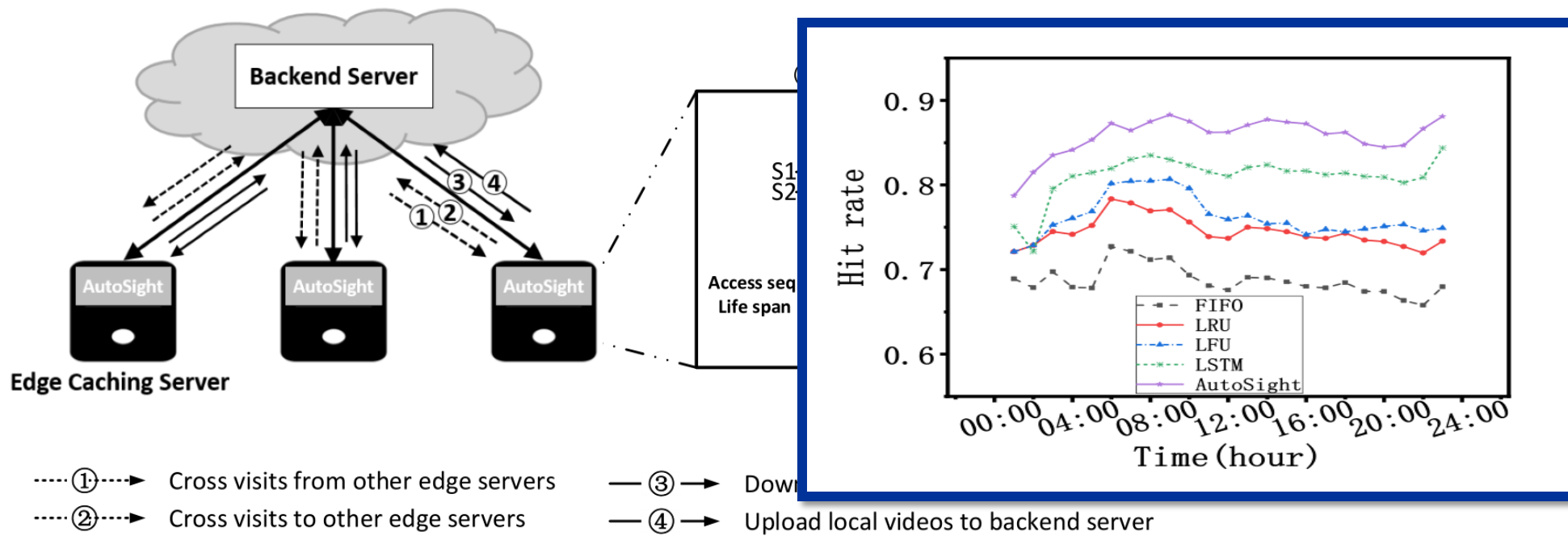
**It is difficult to guarantee a constant high hit rate with static Future Horizon**

## Different Future Horizon

# 1.2 How to evict?

## AutoSight

*AutoSight* consists of a propagation-based predictor (**CoStore**) that predicts the number of times a video will be requested, and a caching engine (**Viewfinder**) with adaptive future horizon to make caching decisions.



# 1. Cache on the edge

## PUBLICATION

- [1] **Yuchao Zhang**, Pengmiao Li, Zhili Zhang, Bo Bai, Gong Zhang, Wendong Wang, Bo Lian. **Challenges and Chances for the Emerging Short Video Network**. IEEE International Conference on Computer Communications (Infocom'2019) Poster. 29 April - 2 May, Paris, France.
- [2] **Yuchao Zhang**, Pengmiao Li, Zhili Zhang, Bo Bai, Gong Zhang, Wendong Wang, Bo Lian, Ke Xu. **AutoSight: Distributed Edge Caching in Short Video Network**. IEEE Network. 2020, 34(3), pp. 194-199. IF: 7.503, JCR Q1.
- [3] **Yuchao Zhang**, Pengmiao Li, Zhili Zhang, Chaorui Zhang, Wendong Wang, Yishuang Ning, Bo Lian. **GraphInf: A GCN-based Popularity Prediction System for Short Video Networks**. 2020 International Conference on Web Services, Sept. 18-20, 2020, Honolulu, Hawaii, USA. (Virtual Conference)
- [4] Pengmiao Li, **Yuchao Zhang**#, Huahai Zhang, Wendong Wang, Ke Xu, Zhili Zhang. **CRATES: A Cache Replacement Algorithm for Low Access Frequency Period in Edge Servers**. The 17th International Conference on Mobility, Sensing and Networking (MSN), December 13-15, 2021, **Exeter, UK**.
- [5] Bo Yi, Fuliang Li, **Yuchao Zhang**#, Xingwei Wang. **TPA based Content Popularity Prediction for Caching and Routing in Edge-Cloud Cooperation Network**. 2021 IEEE Global Communications Conference (GlobeCom), December 7-11, 2021. Madrid, Spain. (Hybrid: In-Person and Virtual Conference)

## ONGOING

2021/10/7

### **Under submission**

FastINF: Popularity Prediction in Short Video Network by a Fast Graph Convolutional Neural Network

Pengmiao Li, Yuchao Zhang, Member, IEEE, Zhili Zhang, Fellow, IEEE.

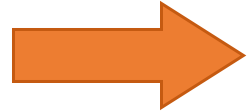
### *Delayed Evicting for Dynamic Caching*

### **Ongoing**

*Abstract*—The recent advances in the development of capable smart devices and mobile Internet services have resulted in rapidly escalating levels of data traffic over the social network. To reduce the tremendous pressure to the backbone network & data centers and provides better QoE with lower latency, popular



# Outline



## 1. Cache on the edge

Exponentially growing data amount **VS.** Limited storage on the edge



## 2. Transmission among the edges

Large scale transmission **VS.** Real-time traffic requirements



## 3. Federation of multiple edges

Strict data privacy **VS.** Multiple edge network collaboration



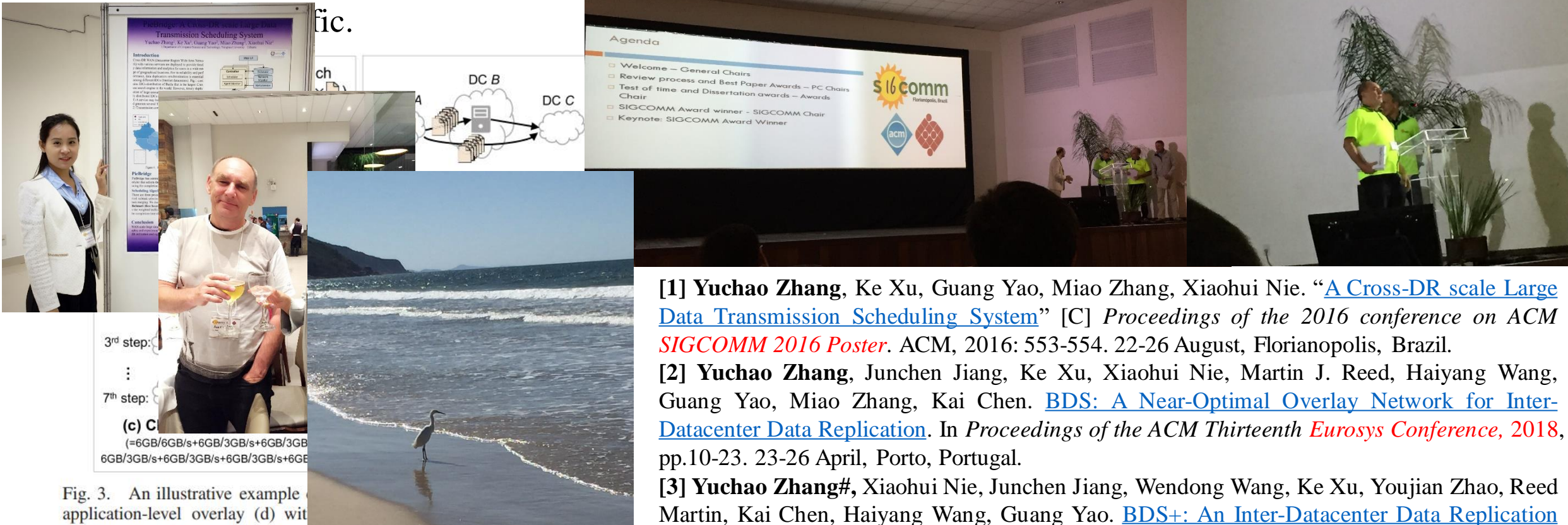
## 2. Transmission among the edges

- As described in the previous part, the demand for inter-domain transmission (cross multiple edge networks) is exploding.



# 2.1 Data Replication

■ BDS+ is an application-level **multicast overlay network** with a fully centralized architecture, which fully utilizes the available overlay paths and leverages **dynamic bandwidth separation** to make



[1] Yuchao Zhang, Ke Xu, Guang Yao, Miao Zhang, Xiaohui Nie. “[A Cross-DR scale Large Data Transmission Scheduling System](#)” [C] *Proceedings of the 2016 conference on ACM SIGCOMM 2016 Poster*. ACM, 2016: 553-554. 22-26 August, Florianopolis, Brazil.

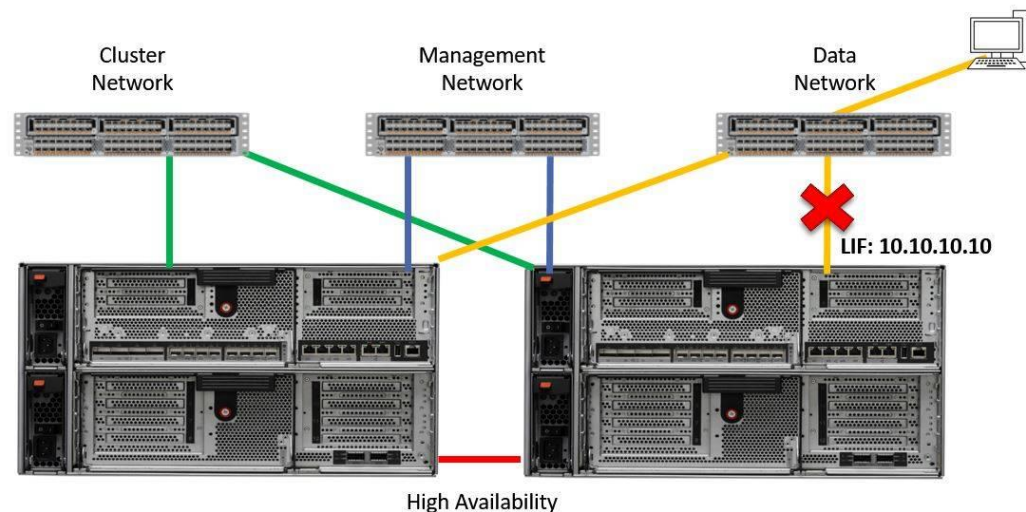
[2] Yuchao Zhang, Junchen Jiang, Ke Xu, Xiaohui Nie, Martin J. Reed, Haiyang Wang, Guang Yao, Miao Zhang, Kai Chen. [BDS: A Near-Optimal Overlay Network for Inter-Datacenter Data Replication](#). In *Proceedings of the ACM Thirteenth Eurosys Conference, 2018*, pp.10-23. 23-26 April, Porto, Portugal.

[3] Yuchao Zhang#, Xiaohui Nie, Junchen Jiang, Wendong Wang, Ke Xu, Youjian Zhao, Reed Martin, Kai Chen, Haiyang Wang, Guang Yao. [BDS+: An Inter-Datacenter Data Replication System with Dynamic Bandwidth Separation](#). *IEEE/ACM Transaction on Networking (ToN)*. 2021, 29(2): 918-934. DOI: 10.1109/TNET.2021.3054924.

Fig. 3. An illustrative example of application-level overlay (d) with overlay (c) and no overlay (b).

## 2.2 When link fails?

- With the scale of inter-DC networks growing exponentially, failures are quite common.
- In a production inter-DC wide area network with more than 200 routers and 6000 links, the link failure frequency is close to **25%** in every 5 minutes, and close to **40%** in every 10 minutes[2].

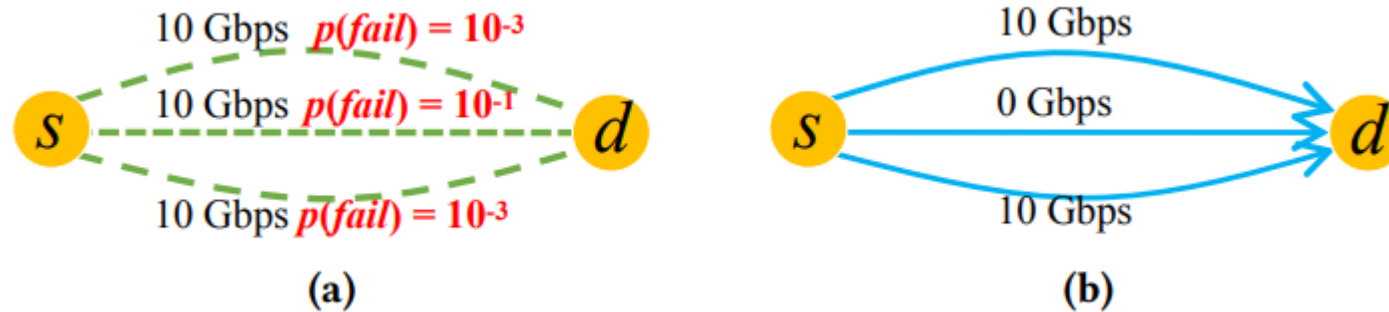


[2]. H. H. Liu, S. Kandula, R. Mahajan, M. Zhang, and D. Gelernter, "Traffic engineering with forward fault correction," in Proceedings of the 2014 ACM Conference on SIGCOMM, 2014, pp. 527–538.

## 2.2 Existing Solutions & Drawbacks

### ■ *Availability*

- TEAVaR[3] concentrates on striking the balance between availability and utilization, but it cost too much time for it takes the whole network into computation, while in inter-DC situations, fast failure recovery is needed.



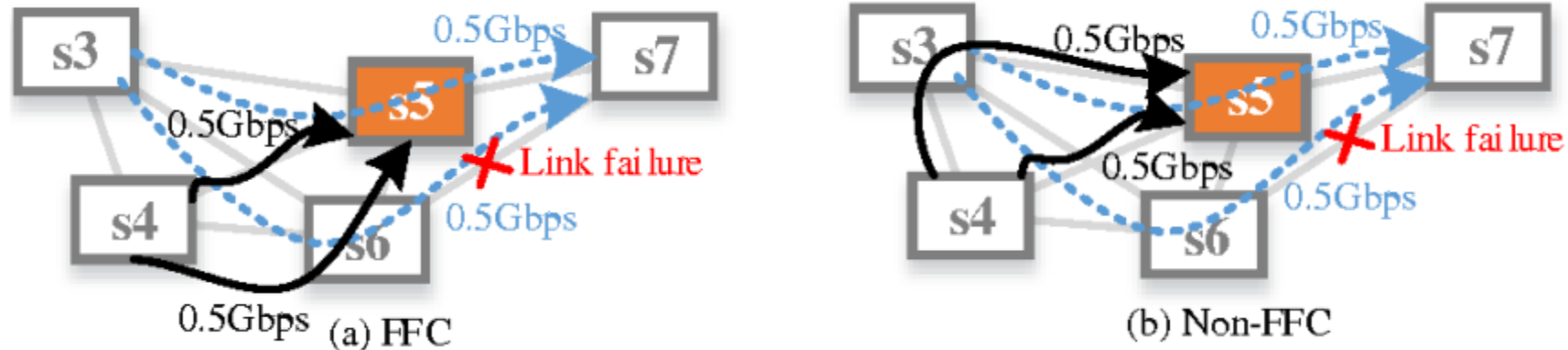
TEAVaR example



## 2.2 Existing Solutions & Drawbacks

### ■ *Efficiency*

- FFC[2] achieves fast failure recovery, but sacrifices network throughput, and no guarantee on future availability/performance.



FFC for link failures( $k=1$ )

## 2.2 Motivation

---

- Challenges – we want both

- ◆ *Efficiency*: Fast recovery against frequent network failures

- ◆ *Availability*: Keep high throughput and avoid future failure

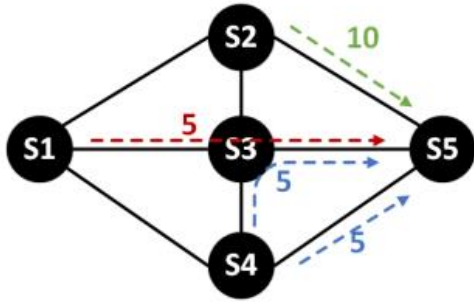
- Ideal

- **Less shuffle**: The less traffic we reroute, the fast recovery we achieve.

- **Lowest risk**: Reroute traffic in the way with lowest risk to encounter future failures.

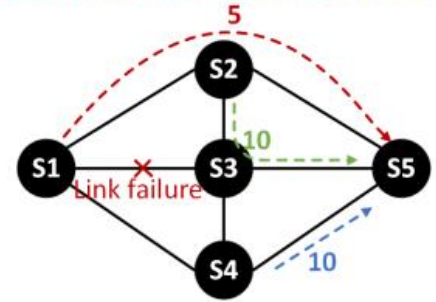
# 2.2 A Motivating Example

Link capacity: 10  
 3 flows:  $s1 \rightarrow s5$ ,  $s2 \rightarrow s5$ ,  $s4 \rightarrow s5$



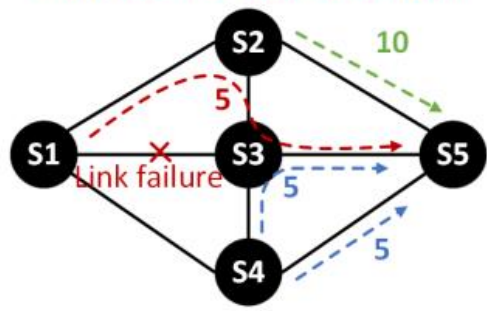
(a) Initial traffic distribution

All the flows have to be re-scheduled



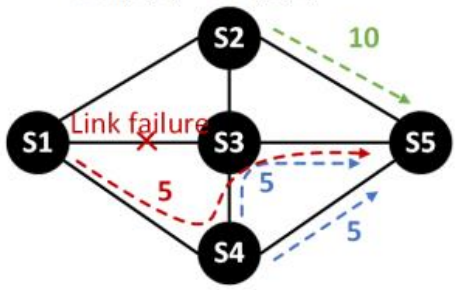
(b) Traffic distribution after global scheduling

Only one flow is re-scheduled



(c) Incremental scheduling

Low risk of rerouting in the future  
 $P_{\text{down}(s1,s2)}=0.1$   
 $P_{\text{down}(s1,s4)}=P_{\text{down}(s3,s4)}=0.001$



(d) FRAVaR scheduling

## ■ Potential solution

### ◆ Incremental scheduling:

- ◆ Fast computation: reroute from the source of failure link. (See Fig(c) ).
- ◆ Reduce impact on other links: reconfiguration of the whole network brings unnecessary traffic reroute. (See Fig(b) ).

### ◆ Guarantee future performance:

- ◆ Risk: prefer to route to a lower failure probability route. (See Fig(d) ).



## 2.2 Quantify the Recovery Efficiency

---

### ■ Global schemes:

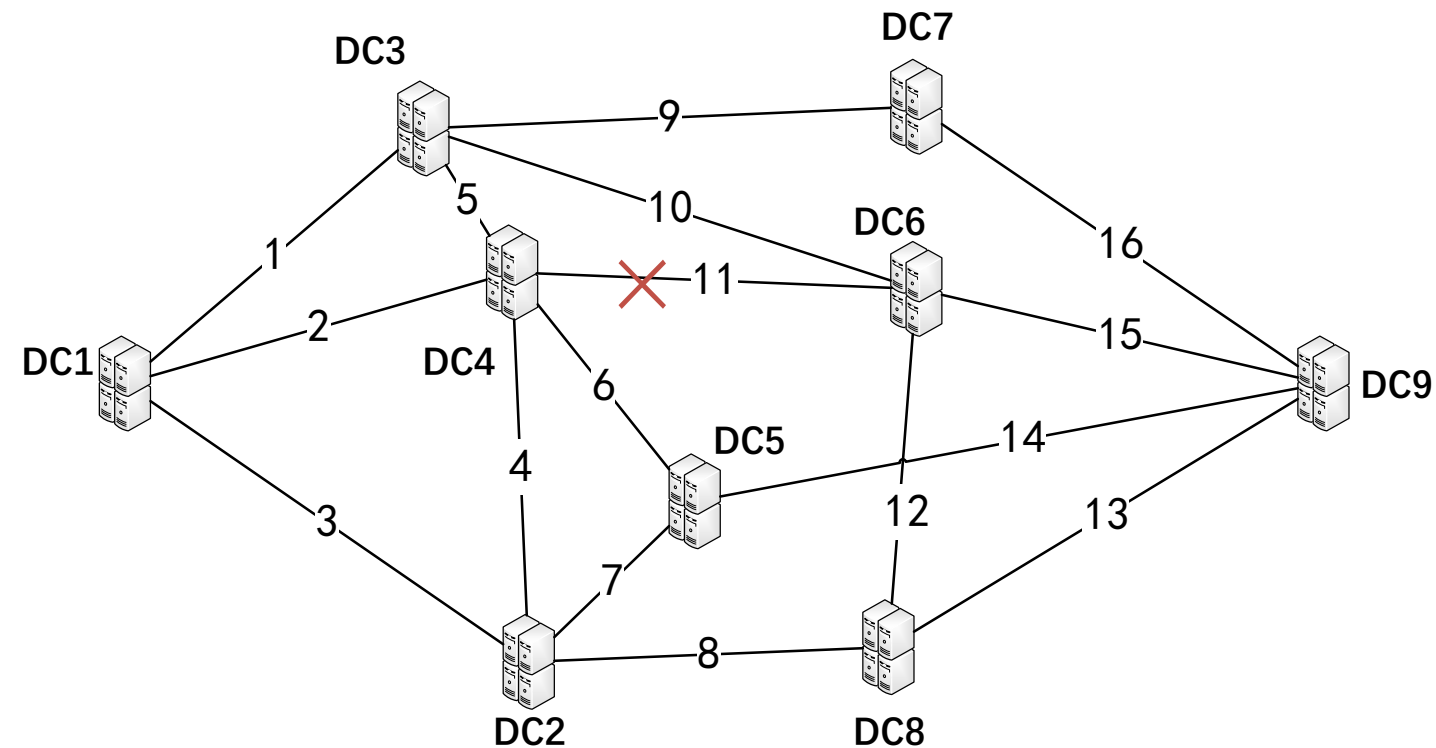
1. The controller has to calculate a global optimization scheme with high complexity.
2. The network has to reconfigure the routing rules on the shuffled paths.
3. A measure in [4] shows when a new routing rules inserting, up to 466 moves are required and each move consumes hundreds of milliseconds.

■ Intuitively, a smaller scale of traffic shuffle requires shorter computation time.



## 2.2 Incremental-scheduling

- Present a non-backtrace algorithm to search paths for recovery from the source to the end of the failed link.

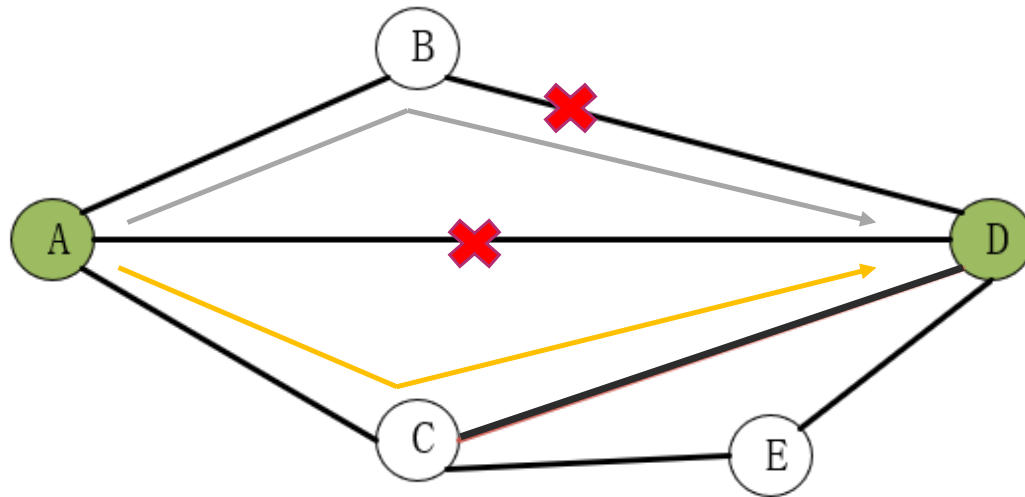


Link11 fails and traffic on it needs to be rerouted.  
Search ends when we **get 3 paths**.

Queue	Results
{5,10},{DC3,DC6}	VALID, {5,10}
{4,8,12}, {DC2,DC8,DC6}	VALID, {4,8,12}
{4,8,13}, {DC2,DC8,DC9}	VISITED
{5,9,16}, {DC3,DC7,DC9}	VISITED
{6,14,15}, {DC5,DC9,DC6}	VALID, {6,14,15}
{5,1}, {DC3,DC1}	VISITED
{5,10},{DC3,DC6}	VALID, {5,10}
{5,9},{DC3,DC7}	
{6,7}, {DC5,DC2}	VISITED
{6,14}, {DC5,DC9}	

## 2.2 Overlapping Failure

- Guarantee future availability
  - ◆ After rerouted caused by one failure, the traffic encounters another failure later when transmitting. We call it the “*Overlapping Failure*” (OF).
  - ◆ An OF increases the scale of rerouted traffic while it should have been unnecessary.

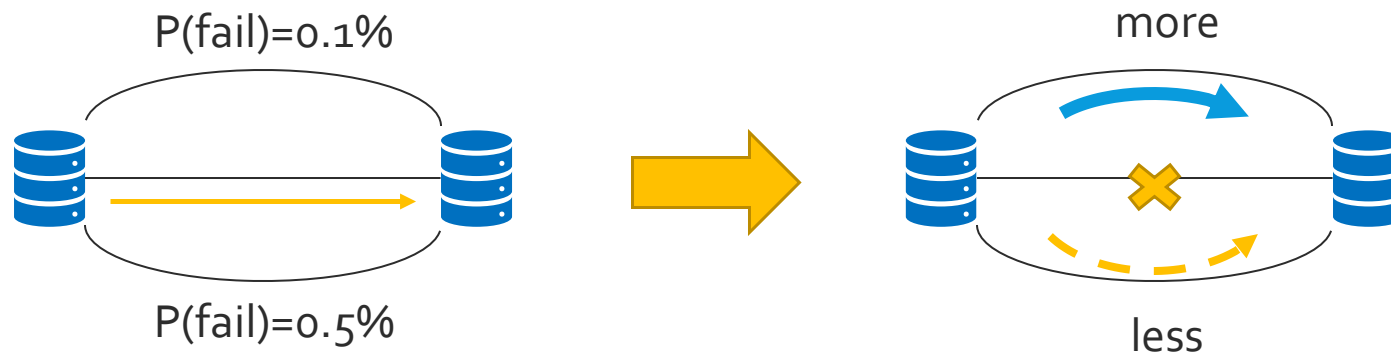


The OF causes recovery once more and the scale of influenced traffic raises overall.

## 2.2 Overlapping Failure

- Avoid Overlapping Failures

- ◆ Allocate less traffic to a high failure probability route to avoid OF.



# 2.2 Evaluation

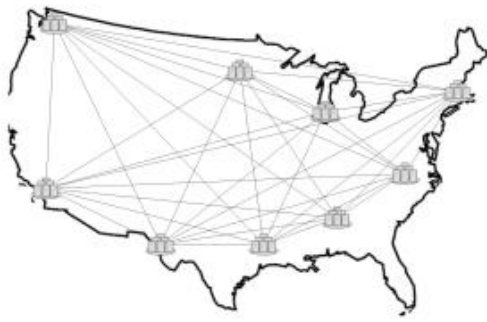
## ■ Average Recovery Time



(a) AGIS



(b) ATT

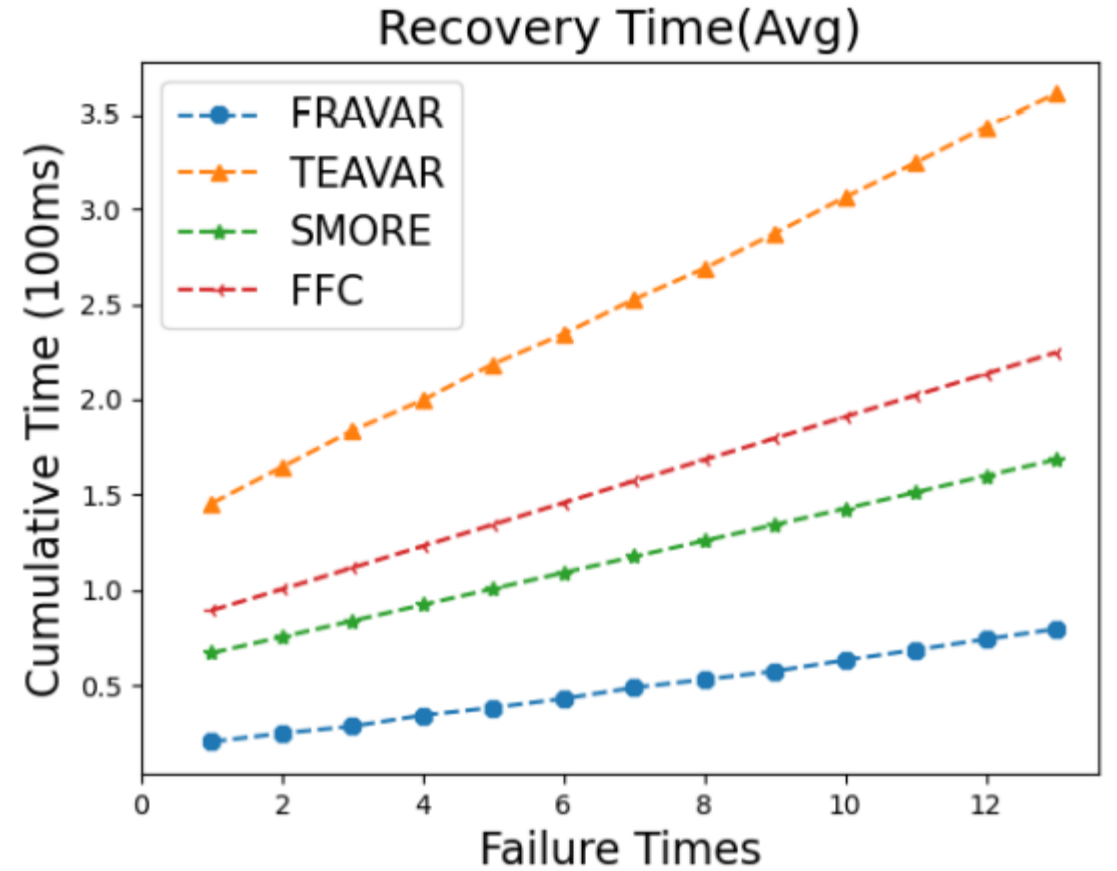


(c) GLOBAL CENTER



(d) IBM

Network topologies applied in simulation.



Average accumulated recovery time cost over 4 topologies.

## 2.2 Evaluation

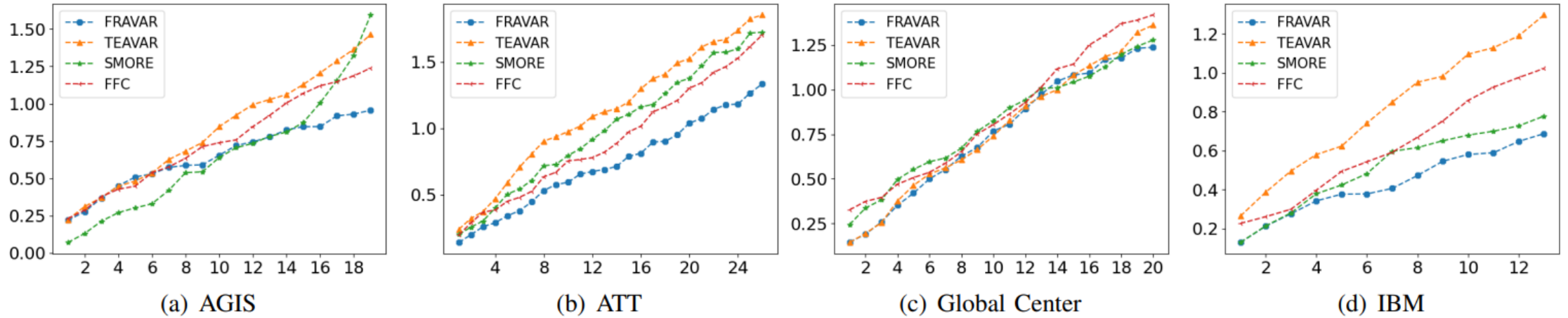


Fig1. Comparison of the scale of traffic rerouted by four failure recovery algorithms over four network topologies. The x-axis is the **failure times** order by epoch and the y-axis is the **cumulative scale of traffic rerouted/shuffled** (metric:10GB).

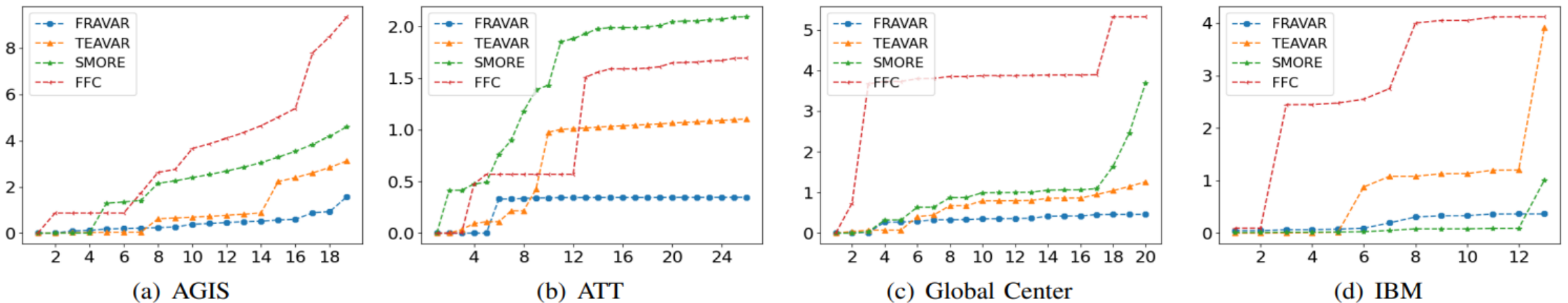


Fig2. Comparison of the scale of rescheduled traffic meeting overlapping failures in four failure recovery algorithms over four network topologies. The x-axis is the **failure times** order by epoch and the y-axis is the cumulative scale of traffic that meets **overlapping failures** (metric:10GB).

# Outline

## 1. Cache on the edge

Exponentially growing data amount **VS.** Limited storage on the edge



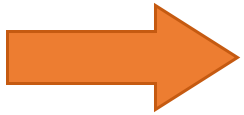
## 2. Transmission among the edges

Large scale transmission **VS.** Real-time traffic requirements



## 3. Federation of multiple edges

Strict data privacy **VS.** Multiple edge network collaboration





### 3. Routing among multiple edge networks

**Wrong config./status?**  
**→ Network Failures/Attack**

**No network status?**  
**→ Data island**

Facebook outage on 4<sup>th</sup> Oct.

Our engineering teams have learned that configuration changes on the backbone between our data centers caused issues that interrupted this communication. This disruption to network traffic had a significant effect on the way our data centers communicate, bringing our services to a halt.

-- Facebook

Mark Zuckerberg  
23 小时前

Facebook, Instagram, WhatsApp and Messenger are coming back online now. Sorry for the disruption today – I know how much you rely on our services to stay connected with the people you care about.

1.9M 621K 449K

# 3. Existing Solutions and Drawbacks

## ■ How to use **correct** edge network inform./config. **safely**?

- Non-shared intra-domain state information cannot assist in cross-domain routing decisions, i.e., the **data islands** problem.
- Intra-domain data sharing will lead to privacy **data leakage** issues.

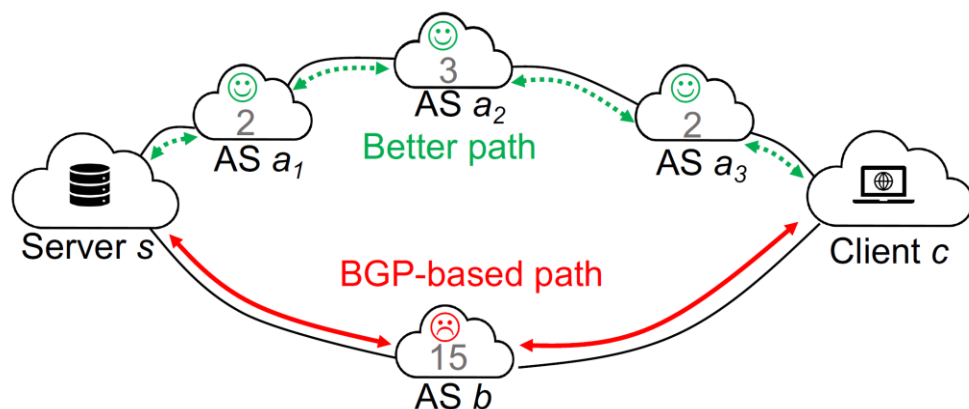


Figure 1. AS hop-based inter-domain routing policies may miss the optimal path

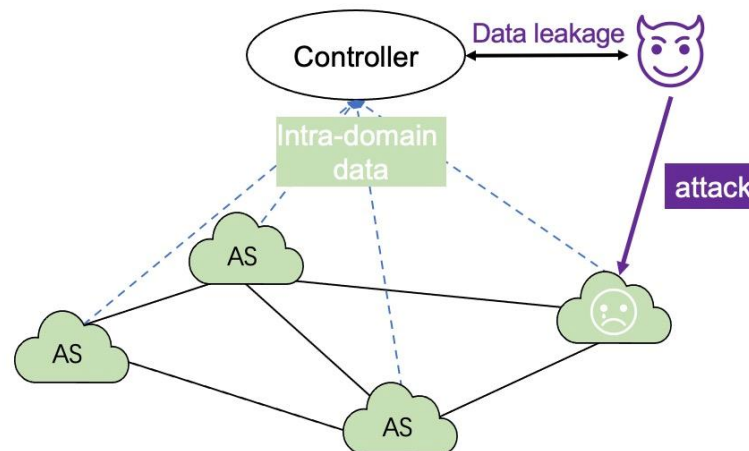


Figure 2. Naive data export will cause data privacy issues

# Research Progress

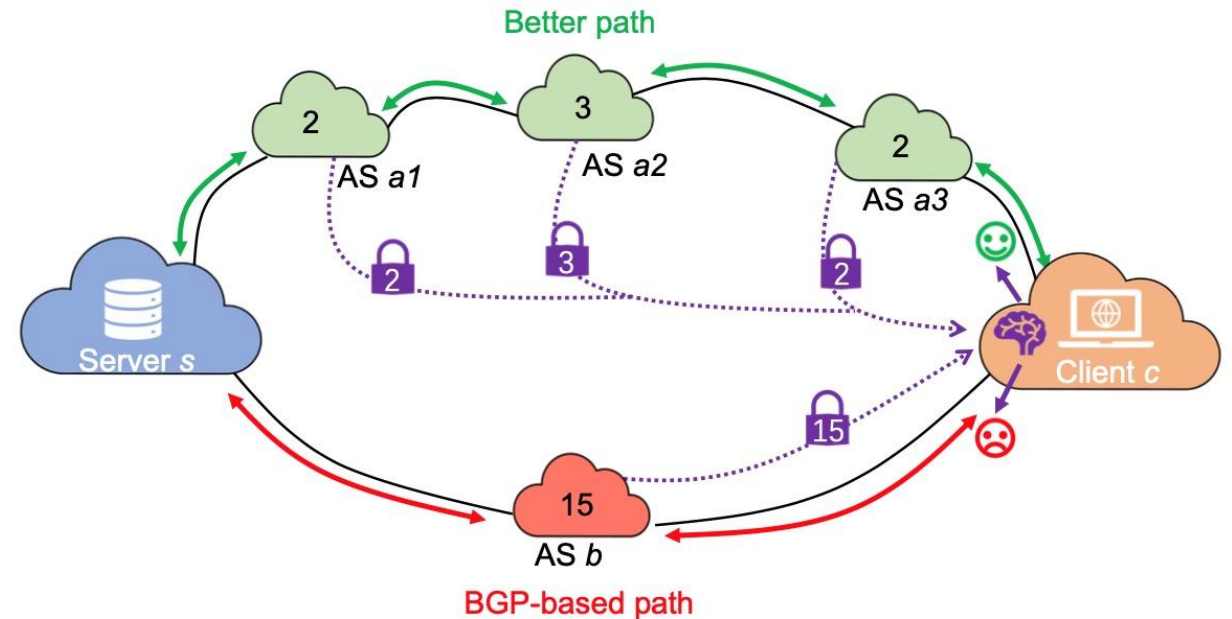
## I<sup>2</sup>BGP (Intra-domain Information-aware Border Gateway Protocol)

### ■ Motivation

If the operations and comparisons are performed without knowing the specific values, then the contradictory problems in inter-domain routing can be solved to certain extent.

### ■ Overview

Homomorphic encryption has the characteristics that we need. Therefore, we propose I<sup>2</sup>BGP, an Intra-domain Information-aware Border Gateway Protocol, which can privately export intra-domain states.



Path	AS-path	Priority	State	Hops
[c, b, s]	2	high	selected	hops= 15
[c, a3, a2, a1, s]	4	low	discarded	hops= 2 + 3 + 2

# I<sup>2</sup>BGP

## I<sup>2</sup>BGP (Intra-domain Information-aware Border Gateway Protocol)

### ■ Routing Diffusion Example

D (d.1)->B (b.2)					D (d.1)->C (c.2)				
Des	Next-hop	AS-path	MED	Des-AS	Des	Next-hop	AS-path	MED	Des-AS
d.2	d.1	1	10	D	d.2	d.1	1	10	D

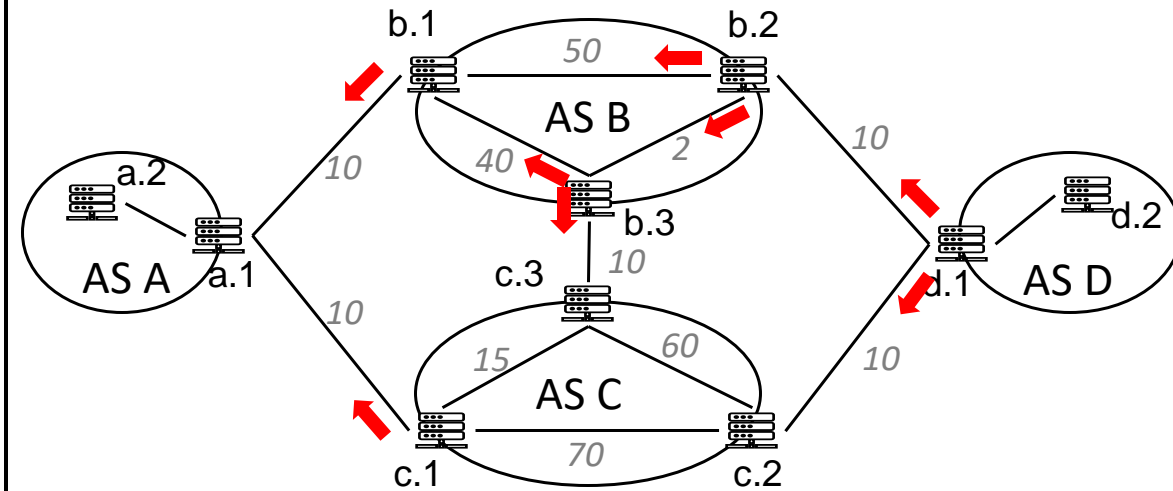
b.2->b.1					b.2->b.3				
Des	Next-hop	AS-path	MED	Des-AS	Des	Next-hop	AS-path	MED	Des-AS
d.2	b.2	1	60	D	d.2	b.2	1	12	D

b.3->b.1					B (b.3)->C (c.3)				
Des	Next-hop	AS-path	MED	Des-AS	Des	Next-hop	AS-path	MED	Des-AS
d.2	b.3	1	52	D	d.2	b.3	2	22	D

B (b.1)->A (a.1)					C (c.1)->A (a.1)				
Des	Next-hop	AS-path	MED	Des-AS	Des	Next-hop	AS-path	MED	Des-AS
d.2	b.1	2	62	D	d.2	c.1	3	47	D

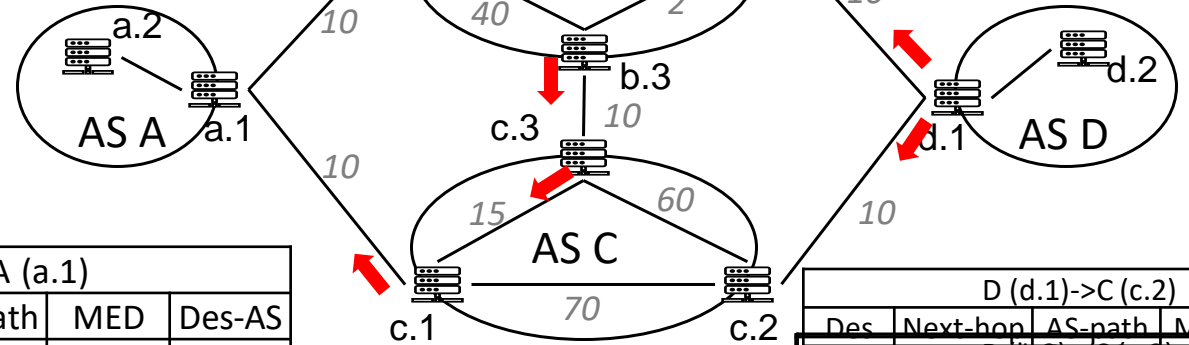


# I<sup>2</sup>BGP workflow

## I<sup>2</sup>BGP (Intra-domain Information-aware Border Gateway Protocol)

- Example of a local RIB update.

B (b.1)->A (a.1)				
Des	Next-hop	AS-path	MED	Des-AS
d.2	b.1	2	62	D



C (c.1)->A (a.1)				
Des	Next-hop	AS-path	MED	Des-AS
d.2	c.1	3	47	D

D (d.1)->B (b.2)				
Des	Next-hop	AS-path	MED	Des-AS
d.2	d.1	1	10	D

BGP table of a.1				
Des	Next-hop	AS-path	MED	Des-AS
d.2	c.1	3	47	D
	b.1	2	62	D

BGP table of b.1				
Des	Next-hop	AS-path	MED	Des-AS
d.2	b.2	2	60	D
	b.3	2	52	D

BGP table of b.2				
Des	Next-hop	AS-path	MED	Des-AS
d.2	d.1	1	update	D
	b.3	2	80	D

BGP table of b.3				
Des	Next-hop	AS-path	MED	Des-AS
d.2	b.2	1	12	D
	c.3	2	70	D

BGP table of c.1				
Des	Next-hop	AS-path	MED	Des-AS
d.2	c.3	2	47	D
	c.2	1	80	D

BGP table of c.2				
Des	Next-hop	AS-path	MED	Des-AS
d.2	c.3	3	82	D
	d.1	1	consistent	D

BGP table of c.3				
Des	Next-hop	AS-path	MED	Des-AS
d.2	b.3	2	22	D
	c.2	1	70	D

D (d.1)->C (c.2)				
Des	Next-hop	AS-path	MED	Des-AS
d.2	d.1	1	10	D
d.2	b.3	2	47	D

# Research Progress

## I<sup>2</sup>BGP (Intra-domain Information-aware Border Gateway Protocol)

### ■ Privacy Protection

- **Topology Abstraction:** abstracting each domain into a characteristic topology graph with its border routers (Figure 1), so as to mask specific intra-domain states.

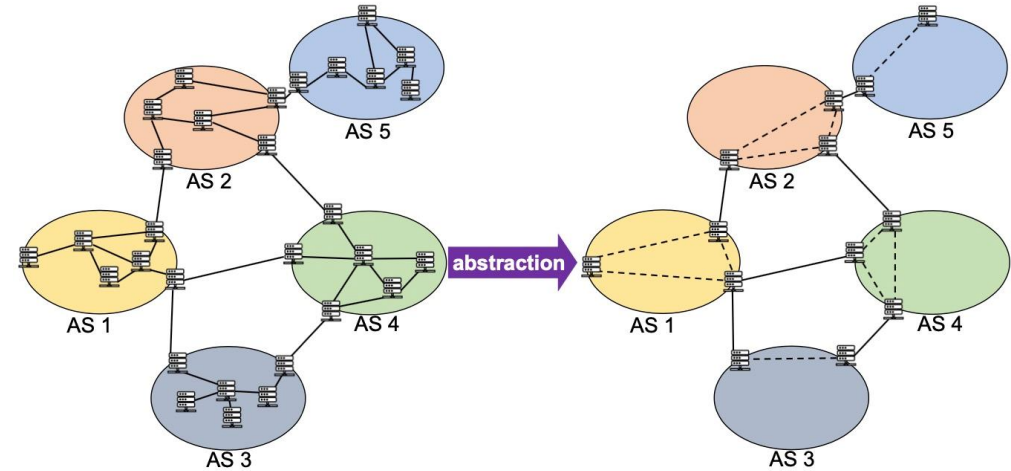
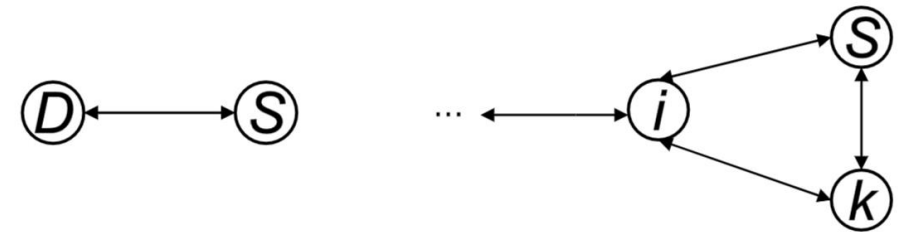


Figure 1. Topology Abstraction

- **Random Number Confusion:** for the directly connected case as shown in Figure 2(a), we confuse the source properties with random numbers to protect its privacy. It allows us get the hops/latency accumulated on the **whole path**.

$$d^{Exported} = d^{Authentic} + \delta_d$$



(a) Direct Connection (↔)

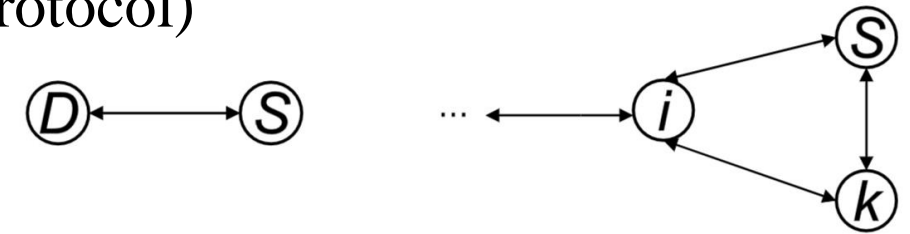
(b) Triangular Connection (Δ)

Figure 2. Connection Description

# Research Progress

## I<sup>2</sup>BGP (Intra-domain Information-aware Border Gateway Protocol)

- Random Number Confusion:** for the triangular connected case as shown in Figure 1(b), we use a **homomorphic encryption-based privacy number comparison** approach to constrain the route diffusion to protect the relevant data.



(a) Direct Connection (↔)      (b) Triangular Connection (Δ)

Figure 1. Connection Description

- Comparing Paths:**

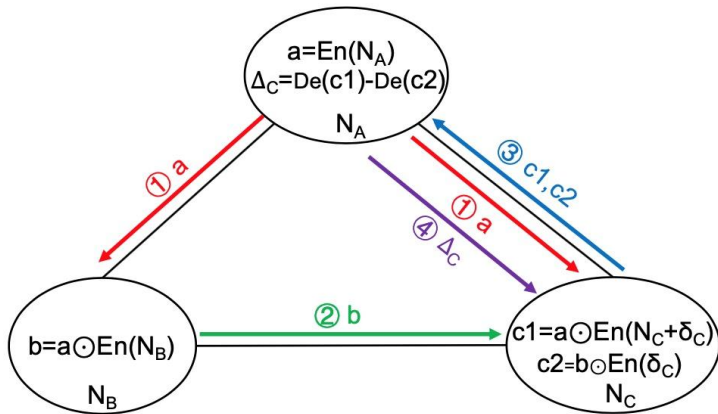


Figure 2. Comparison Example

- Constraining Diffusion:**

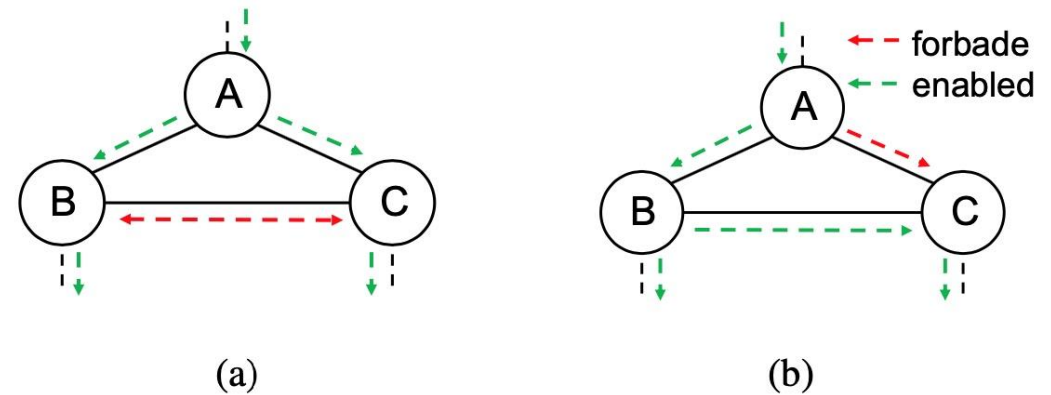


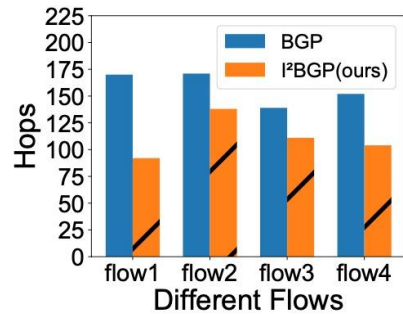
Figure 3. Two types of diffusion constraint



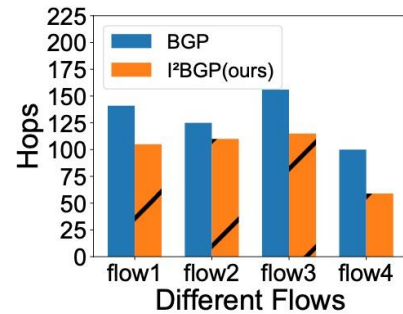
# Research Progress

## I<sup>2</sup>BGP (Intra-domain Information-aware Border Gateway Protocol)

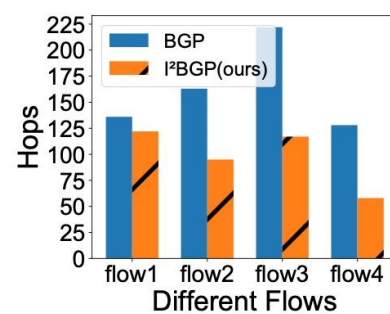
### ■ Performance



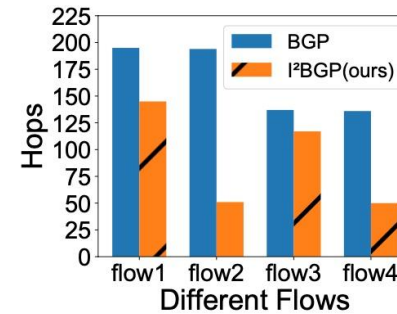
(a) ATMnet



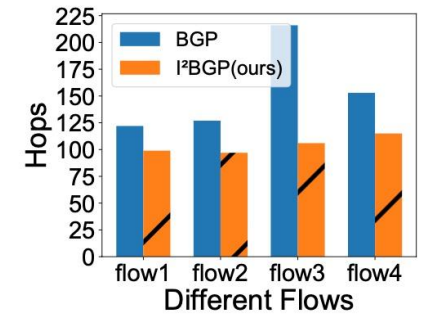
(b) Claranet



(c) Compuserve

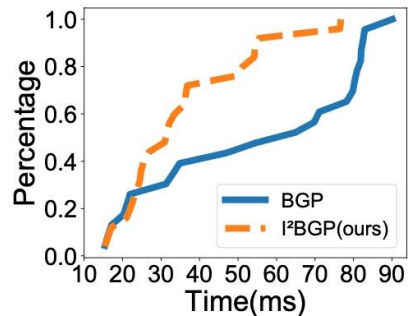


(d) NSFnet

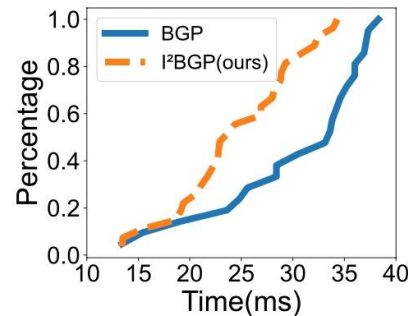


(e) Peer1

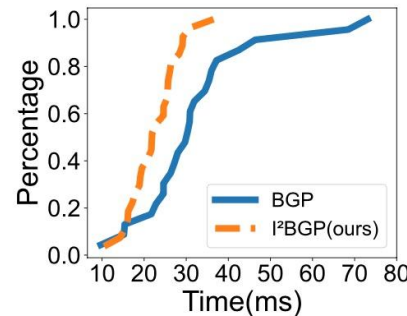
Figure 1. Hops improvements under 5 topologies



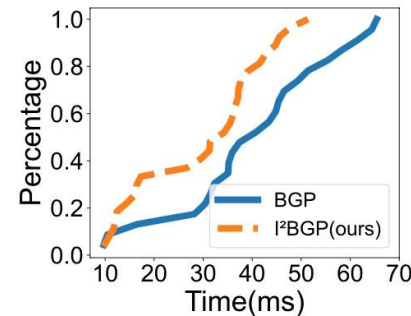
(a) ATMnet



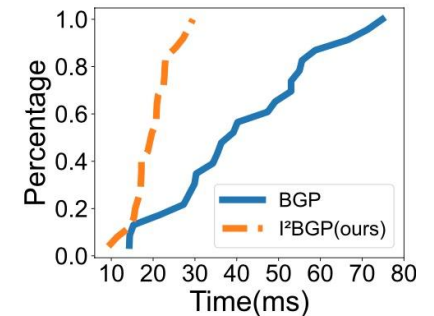
(b) Claranet



(c) Compuserve



(d) NSFnet



(e) Peer1

Figure 2. Latency improvements under 5 topologies

■ This work, I<sup>2</sup>BGP, has been submitted to INFOCOM2022

# Recall the challenges

## ■ Core challenge

- Information from multiple network **VS.** Safety & privacy

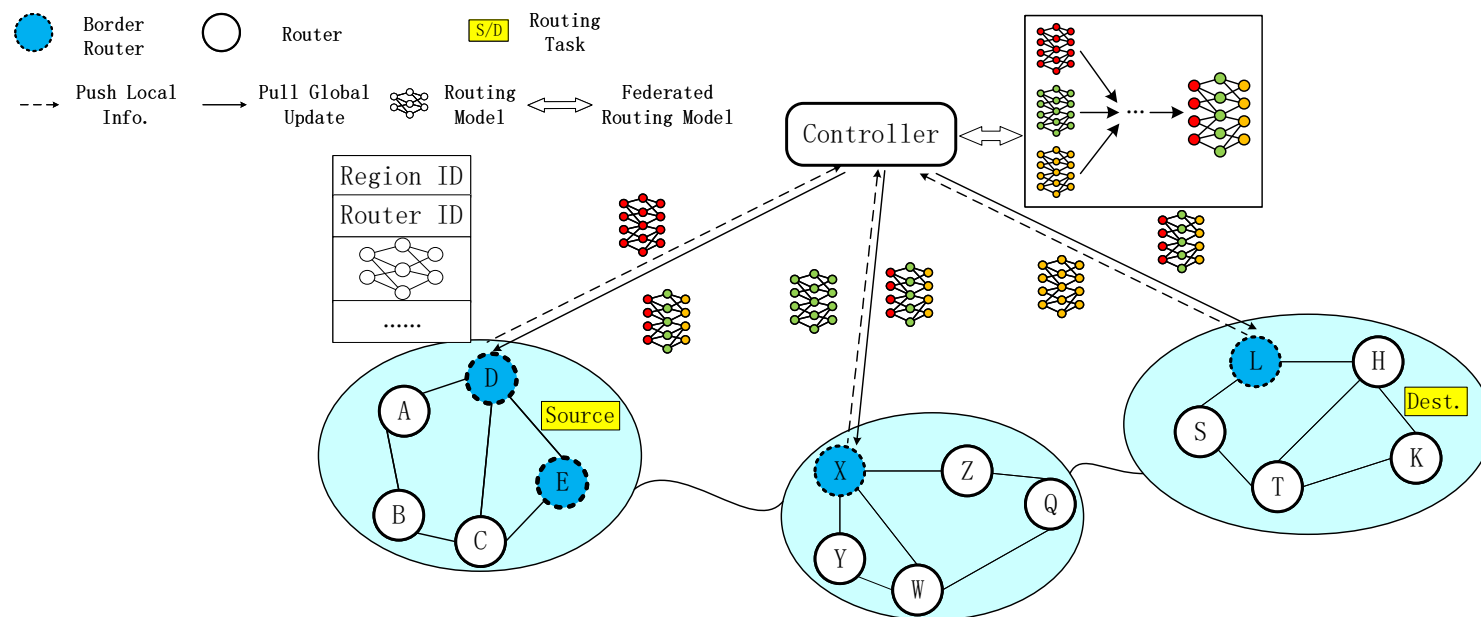
## ■ Opportunity

- Federated Learning, a machine learning framework, which can effectively assist multiple organizations in data usage while meeting user privacy protection and data security requirements.

# Potential Solutions

## ■ Federated Routing

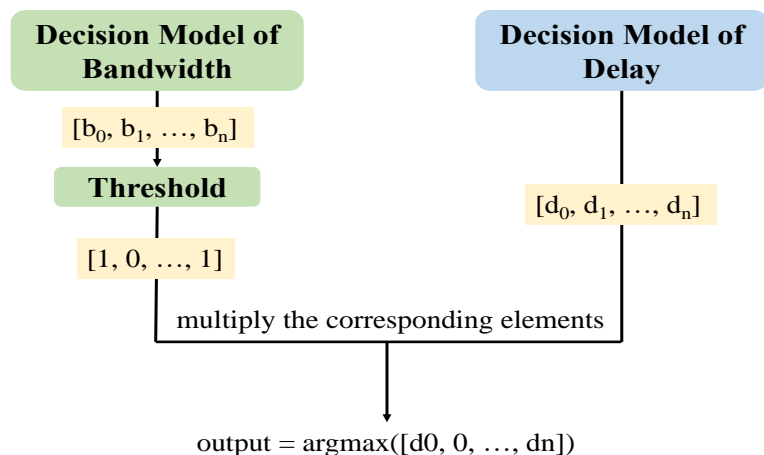
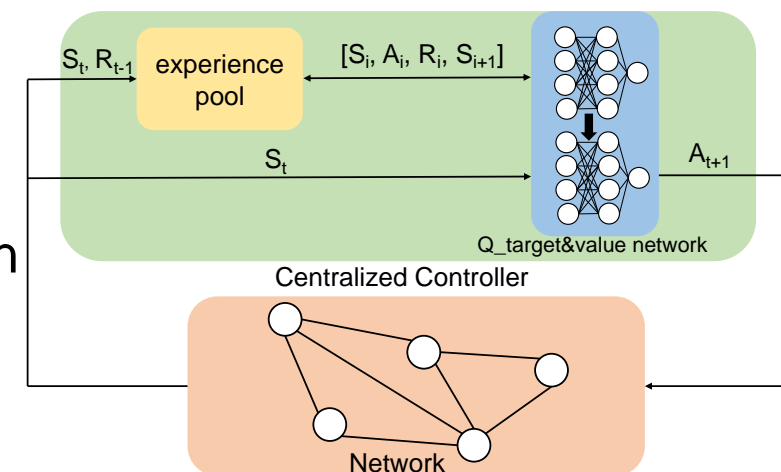
- a federated learning-based inter-domain routing mechanism
- uses the model parameters from each AS
- to co-construct a federated routing model
- achieve optimal inter-domain routing decisions with intra-domain state
- without disclosing private data.



# Research Progress

## ■ A centralized DQN-based intra-domain control strategy:

1. Getting network *status* and action *rewards*
2. *Q\_value* network outputs strategies and sends them out for execution
3. *Q\_target* network trains the network and updates the value network
4. Iterating the above process



## ■ Model fusion-based multi-constraint routing strategy:

1. Defining link priority rules
2. Calculating the best path by model policy fusion

# Research Progress

- This solution enables sharing link attributes flexibly with significant performance improvement.

With the help of intra-domain information, to co-build inter-domain federated routing model.

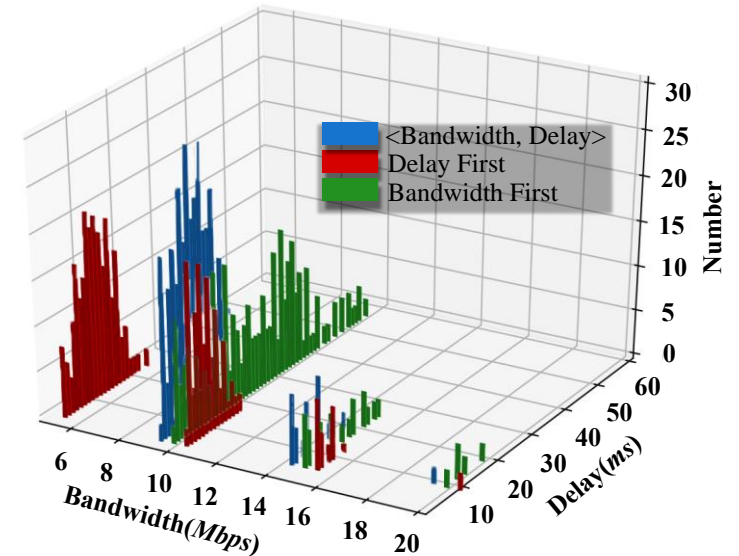


Figure 1. Routing decision performance.

## ■ Published Papers:

- Cong P, **Zhang Y**, Wang W, et al. A Deep Reinforcement Learning-based Routing Scheme with Two Modes for Dynamic Networks[C]//ICC 2021-IEEE International Conference on Communications. IEEE, 2021: 1-6.
- Cong P, **Zhang Y**, Liu Z, et al. A deep reinforcement learning-based multi-optimality routing scheme for dynamic IoT networks[J]. Computer Networks, 2021, 192: 108057.

## 2. Transmission among edges

### PUBLICATION

- [1] **Yuchao Zhang**, Ye Tian, Wendong Wang, Peizhuang Cong, Chao Chen, Dan Li, Ke Xu. **Federated Routing Scheme for Large-scale Domain Network**. IEEE International Conference on Computer Communications (Infocom'2020) Poster. 6-9 July, 2020, Toronto, Canada. (Virtual Conference)
- [2] Peizhuang Cong, **Yuchao Zhang#**, Wendong Wang, Ke Xu, Ruidong Li, Fuliang Li. **A Deep Reinforcement Learning-based Routing Scheme** with Two Modes for Dynamic Networks. 2021 IEEE International Conference on Communications (ICC). 14-23 June 2021, Montreal. (Virtual Conference)
- [3] Peizhuang Cong, **Yuchao Zhang#**, Zheli Liu, Thar Baker, Hissam Tawfik, Wendong Wang, Ke Xu, Ruidong Li, Fuliang Li. A Deep Reinforcement Learning-based Multi-Optimality **Routing Scheme for Dynamic IoT Networks**. Computer Networks (CN), 2021, 192: 108057.

### 2021 IEEE infocom submission

### ONGOING

#### Straightforward? Why Not! A Desensitization-based Inter-domain Routing Protocol

Anonymous

*Abstract*—With the rapid development of Internet technology and the diverse requirements of services, the inter-domain transmission scenario becomes more and more common, which accompanied with ever increasing amount of data. The de facto inter-domain routing protocol, BGP, cannot perceive the intra-domain information, which takes each domain as indiscriminate blackbox. On this basis, the inter-domain routing path that only considers the number of AS is not optimal, which will lead to unbalanced link utilization, unsatisfied transmission perfor-

[6]. The strategy of treating all domains as *indiscriminate blackbox* exists pitfalls, which completely ignores valuable intra-domains states. For example, when an AS provides poor performance of transmission services, e.g., link congestion, high packet loss rate, etc., then performance of all paths that contains this AS will be greatly impaired. Or, it will introduce a large amount of traffic to a popular AS in shortest paths,

# Outline

## 1. Cache on the edge

Exponentially growing data amount **VS.** Limited storage on the edge



## 2. Transmission among the edges

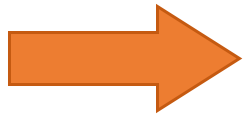
Large scale transmission **VS.** Real-time traffic requirements



**More...**

## 3. Federation of multiple edges

Strict data privacy **VS.** Multiple edge network collaboration





# Some other ongoing work

- ***When the model on the **federated nodes** are heterogeneous?***
  - Network structure
  - Data characteristics
  - Non-IID (Independent and Identical Distribution)
  - .....
- ***How to make the complex federated model **lightweight**?***
  - Knowledge distillation
  - Identify the difficulty of samples
  - Adaptive training sequence
  - .....

**Project of National Natural Science Foundation of China (PI) 2021.08 - 2025.07**

# Some other ongoing work

- **Blockchain: *Nested-chain Architecture* between Massive Off-chain and On-chain Data**
  - Problem: it is difficult to upload the data of application networks to the blockchain system, due to:
    - the massive scale
    - high real-time performance
  - This project:
    - 1) Design of interoperable architecture based on nested chain;
    - 2) Sub-chain construction for massive real-time data;
    - 3) Collaborative consensus based on distributed hash.

**Key Project of Beijing Natural Science Foundation (PI) 2021.01 - 2023.12**



UNIVERSITY OF  
CAMBRIDGE



北京邮电大学  
Beijing University of Posts and Telecommunications

You are more than **welcome** to  
discuss and work together!