## Revisiting Traffic Splitting for Software Switch in Datacenter

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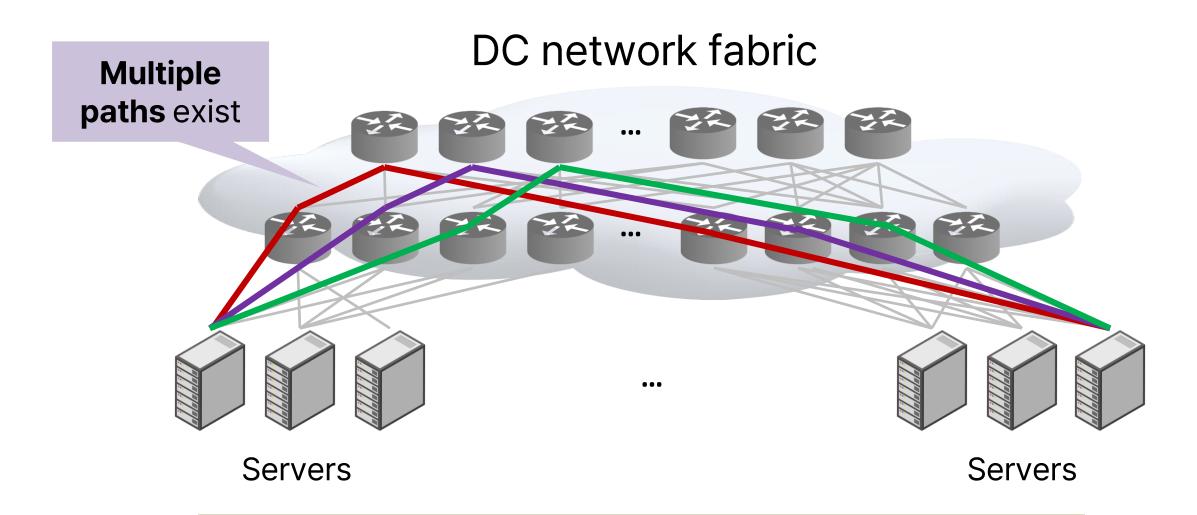
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#### Multipath Networking in Datacenter

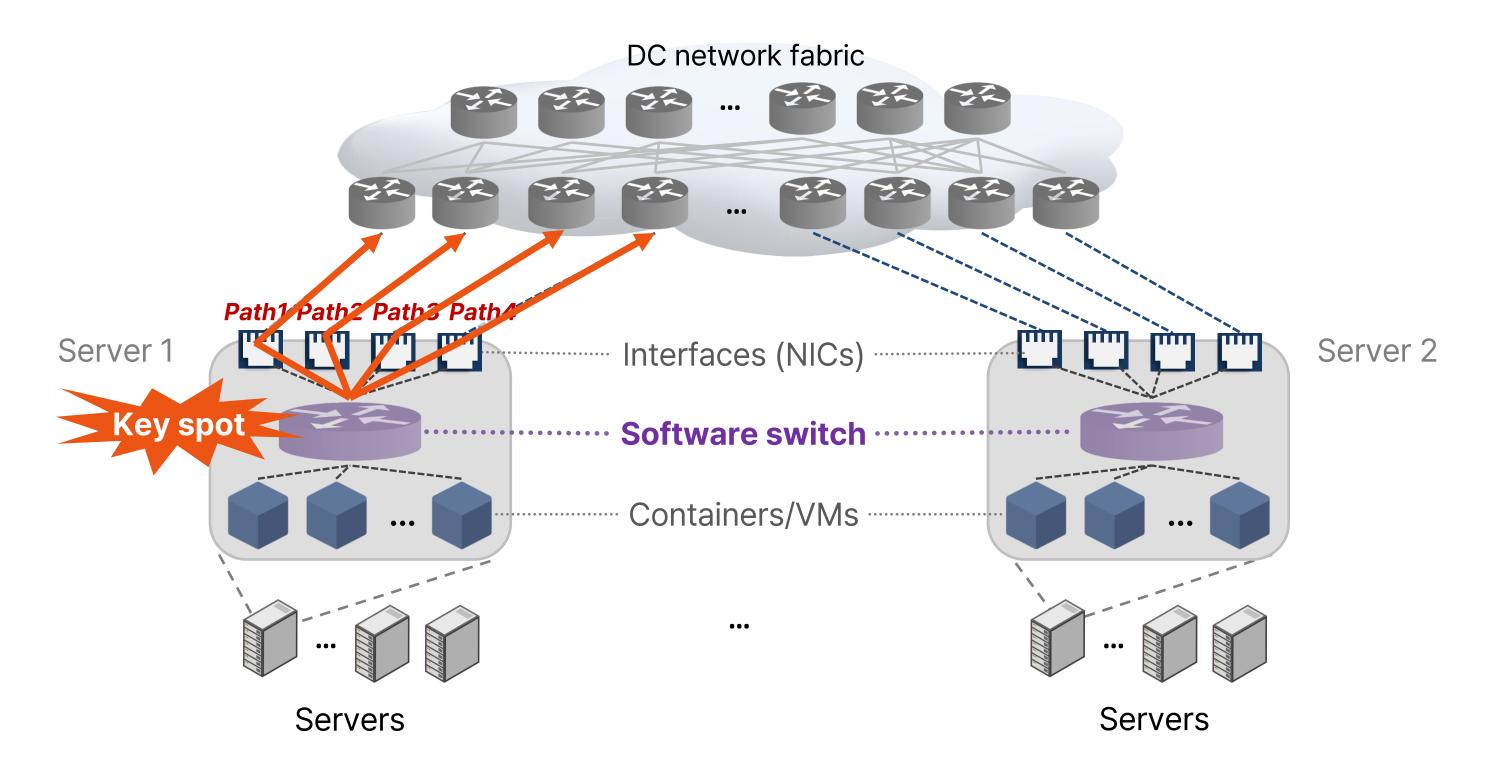
- Traffic in cloud datacenters (DCs) is increasing exponentially
  - Driven by web search, deep learning services, data mining, etc
- Multiple paths between servers for high throughput and reliability



Utilizing paths is critical

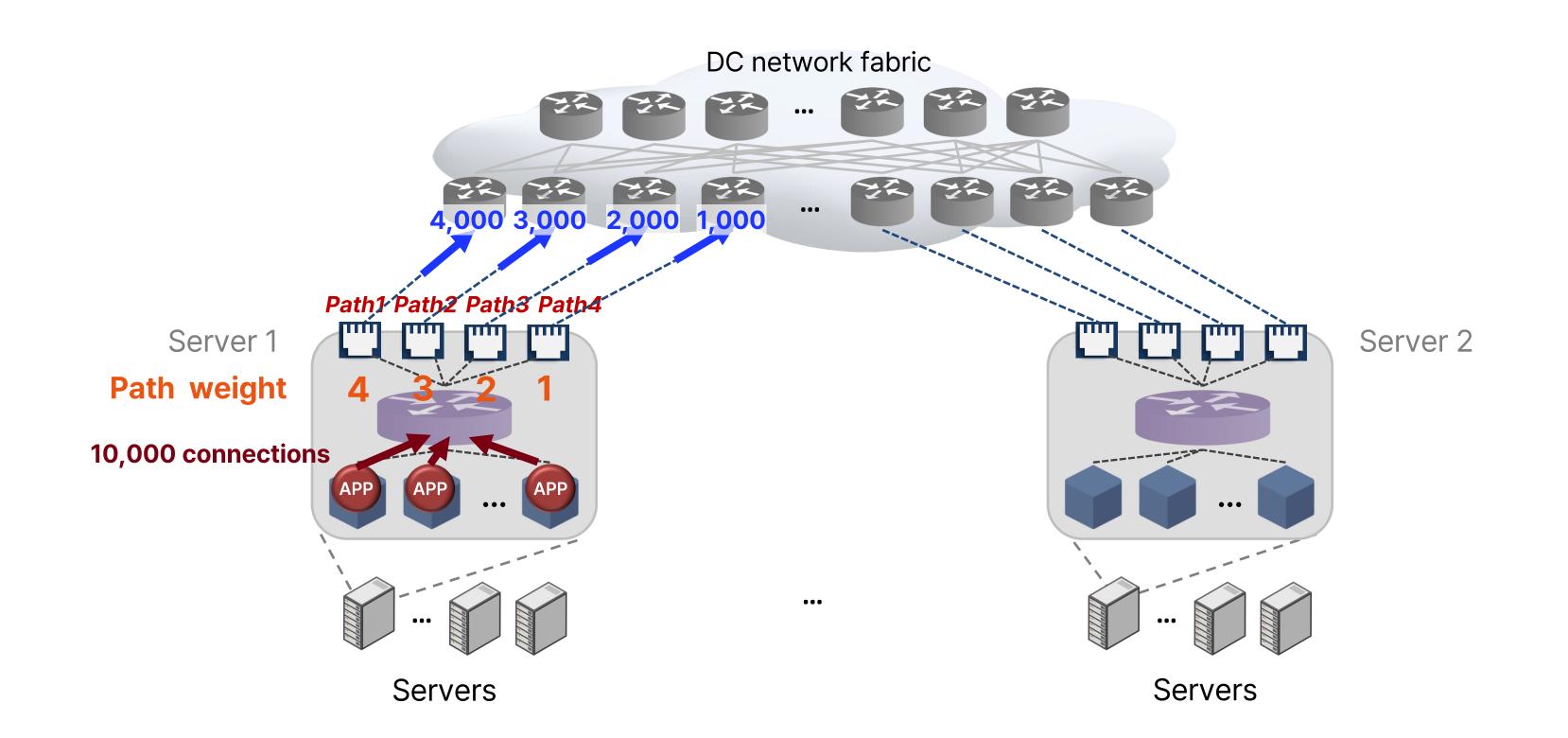
#### **Software Switch in DC**

- Multiple VMs and containers per server
- Software switch: bridge packets between 1) VMs/containers and 2) network interfaces
  - E.g., Open vSwitch (Linux Foundation), Apsara vSwitch (Alibaba), Hoverboard (Google)



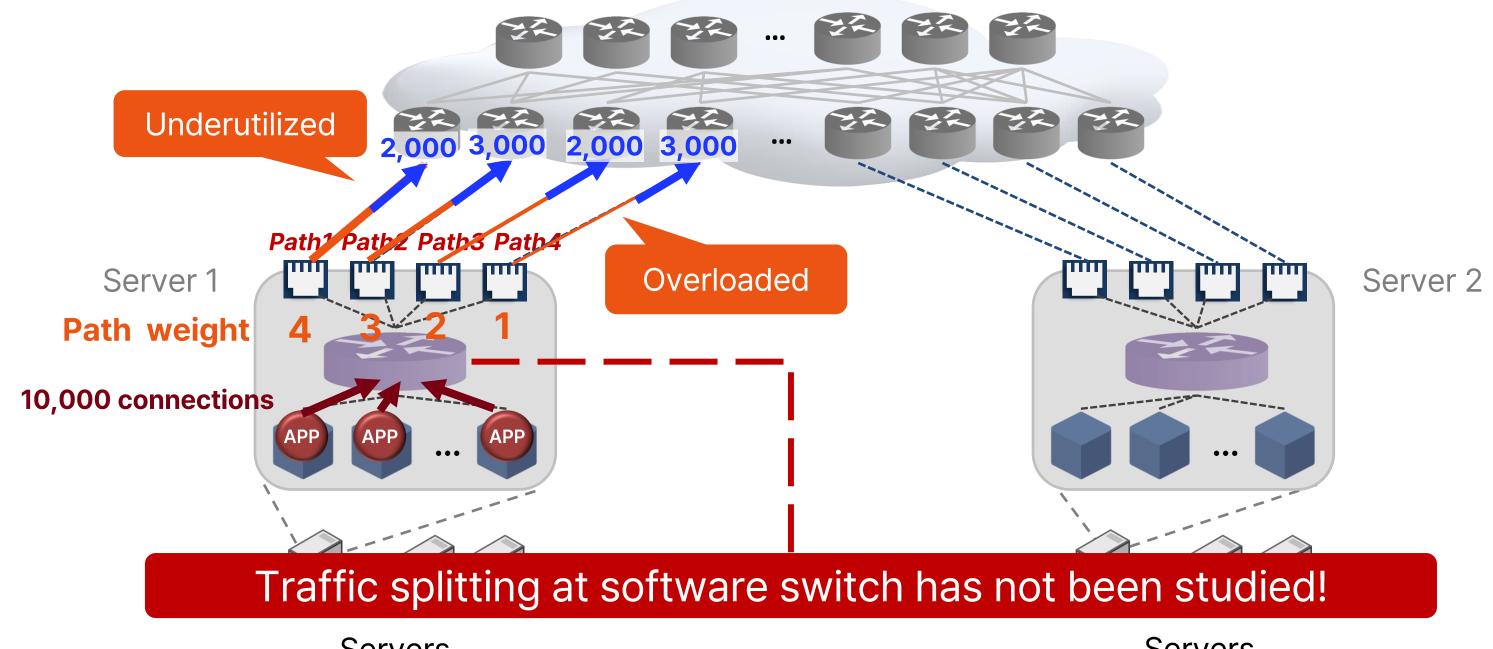
#### **Traffic Splitting at Software Switch**

- Thousands of network connections (TCP/UDP) from VMs/containers
- Software switch splits connections across multiple paths based on weights



#### **Previous Studies on Software Switches**

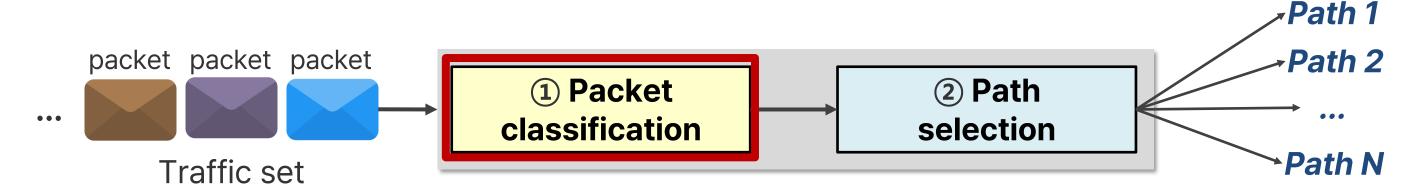
- Many studies have focused on how to determine path weights
  - CLOVE (CoNext`17): adjust weights based on congestion degree (e.g., # of ECN packets)
  - VMS (JSAC`20), TeaVisor (INFOCOM`21): path capacity (e.g., bandwidth, RTT)
- Accurate and efficient traffic splitting often overlooked



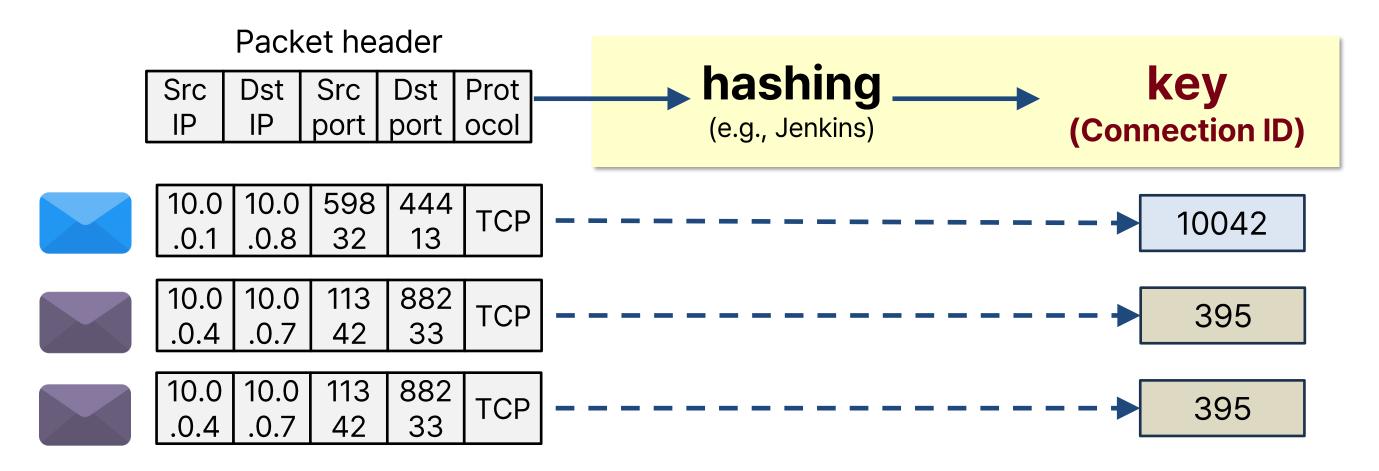
Servers

#### Background: Traffic Splitting Mechanism

Two-stage process: ① Packet classification → ② Path selection



- **① Packet classification**: Identify network connection to which incoming packets belong
- Use a hashing on the packet header's 5-tuple to determine "connection ID" (key)
- Ensure packets from the same connection follow the same path to prevent out-of-order packets



## Background: Traffic Splitting Mechanism

• Two-stage process: ① Packet classification → ② Path selection

Path 1

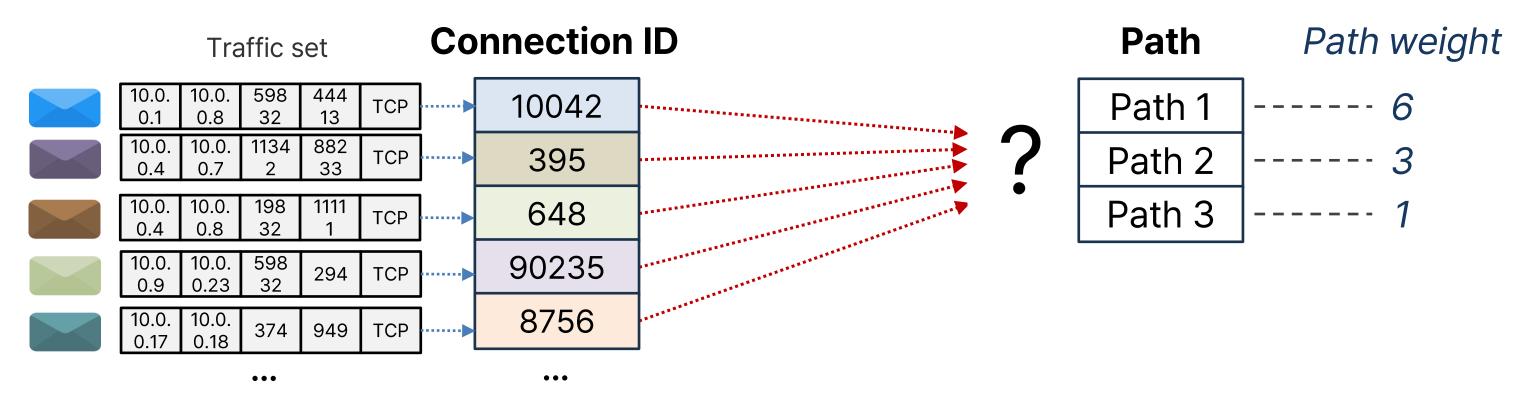
packet packet packet packet classification

Traffic set

Path 2

Path 2

• ② Path selection: assign a path to each connection considering path weights

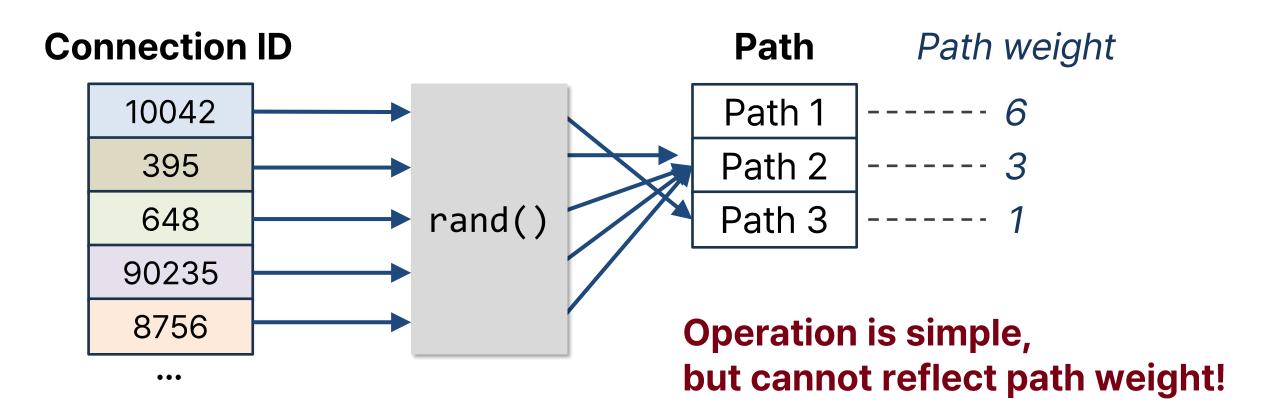


- Four techniques used in software switches:
  - 1) random, 2) weighted round-robin (WRR), 3) weighted cost multipath (WCMP), 4) scoring

Path N

#### Path Selection: (1) Random

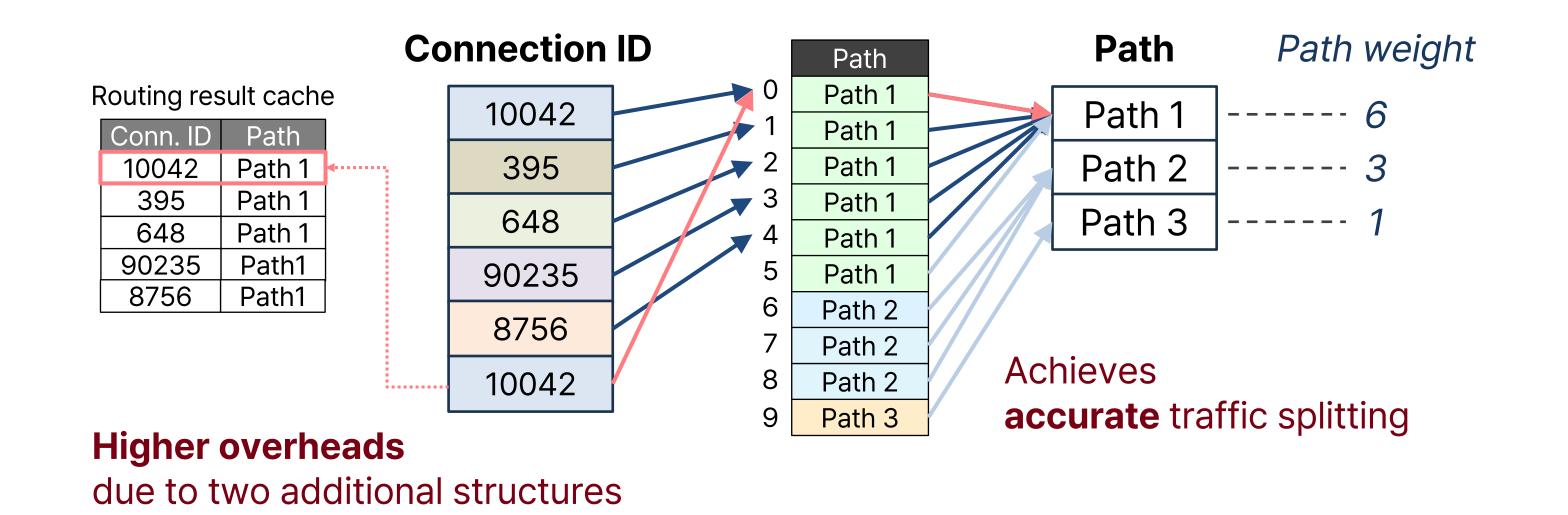
- Path is determined using a random distribution
- Connection ID serves as the random seed



E.g., ECMP generally uses random

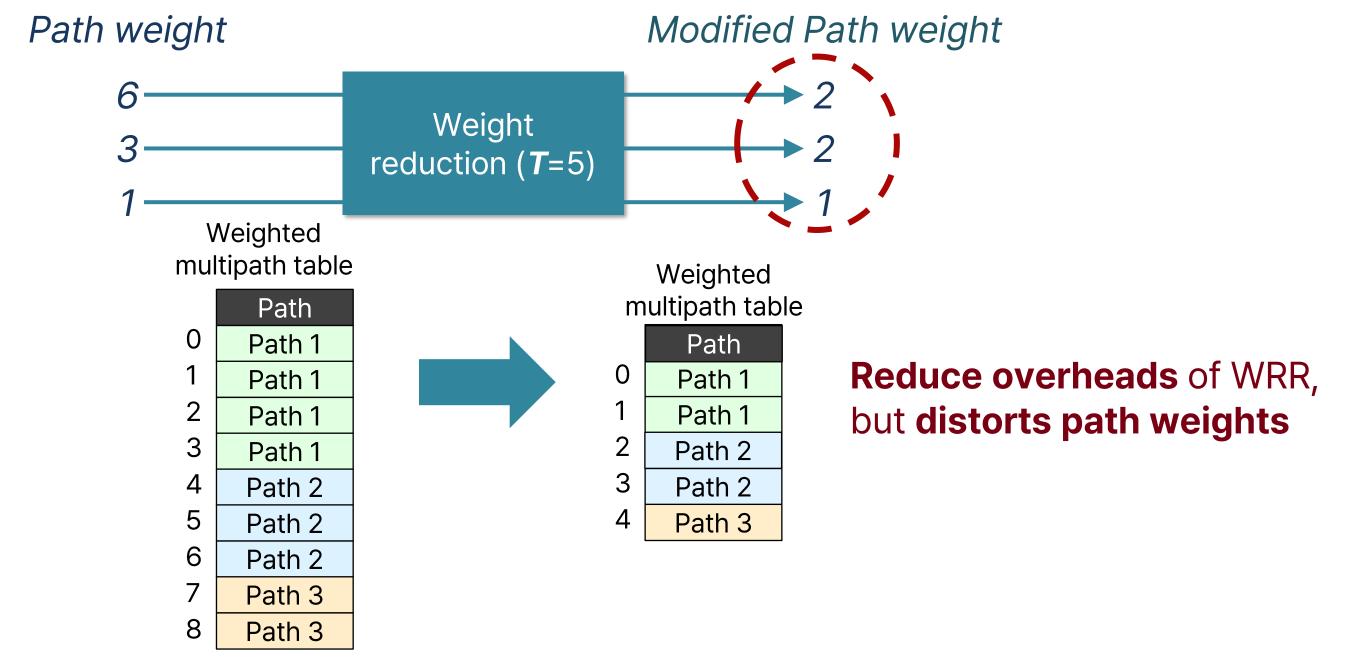
#### Path Selection: (2) WRR

- Select paths sequentially based on weights (round-robin manner)
  - Weighted multipath table: Contains multiple entries per path as much as weights (memory 个)
  - Routing result cache: Stores routing decisions to prevent packet reordering issues (time complexity ↑)



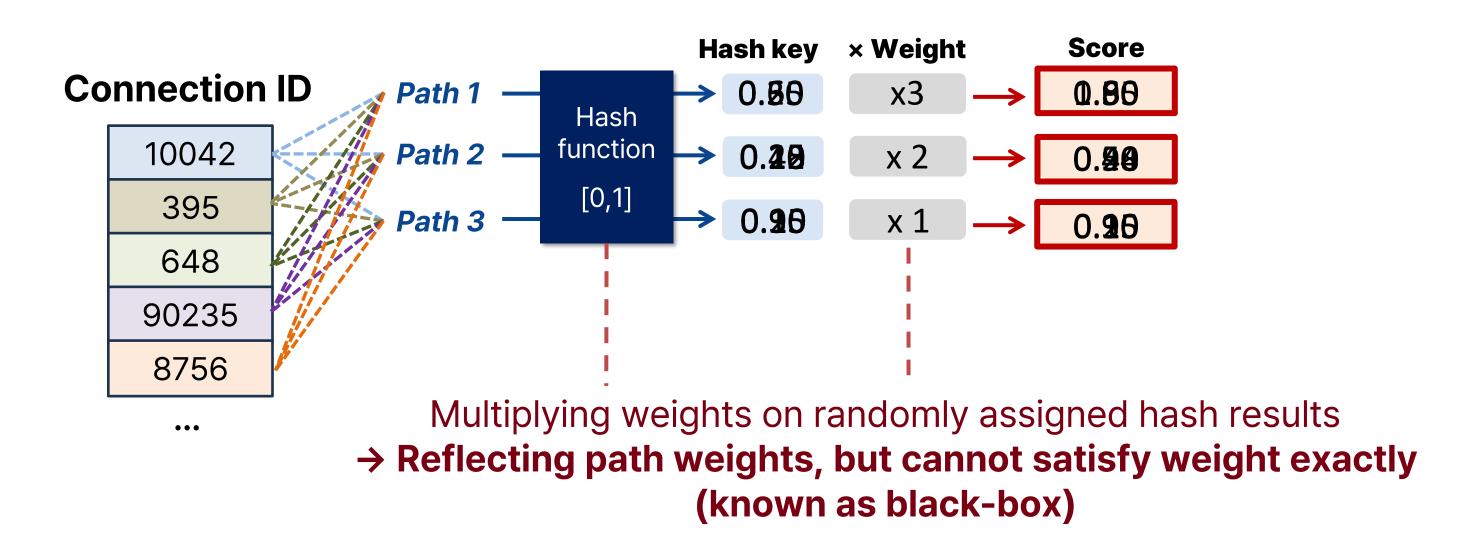
#### Path Selection: (3) WCMP

- Resource-efficient alternative to WRR
  - WRR becomes computationally heavier as number of path weights and connections increases
- Weight reduction\*: change the weights into smaller scales under the threshold
  - E.g., 6, 3,  $1 \rightarrow 2$ , 2, 1 for threshold 5



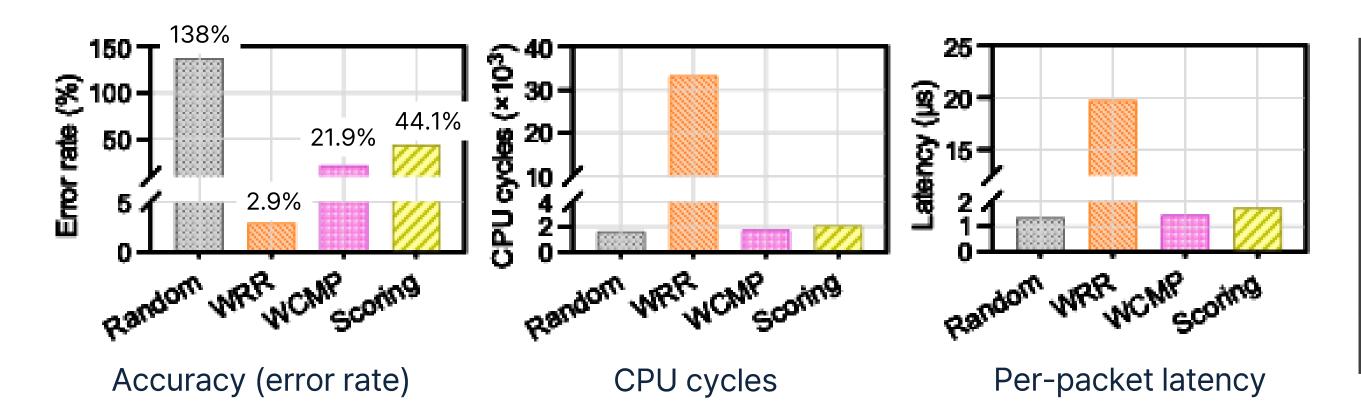
#### Path Selection: (4) Scoring

- De-facto technique
- For each connection, examine all paths`score → select the one with the highest score
  - Score calculation: hash key of (connection ID, path ID) × path weight



#### Problem: Inaccuracy and Resource-inefficiency

- Measure accuracy and resource-efficiency (CPU cycles and per-packet latency)
  - Dataset: Real-world DC traces from CAIDA\* and ClassBench\*\*
  - 200 trials with varying path weights
- Accuracy: Error rate (%) between actual and ideal connection counts per path
- Resource-efficiency: CPU cycles and per-packet latency

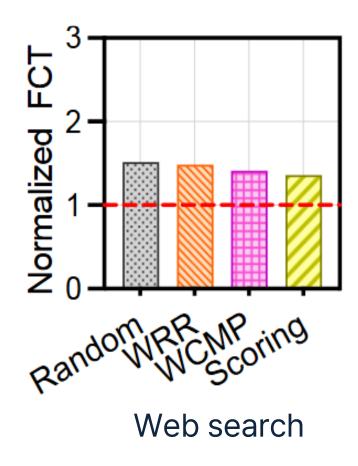


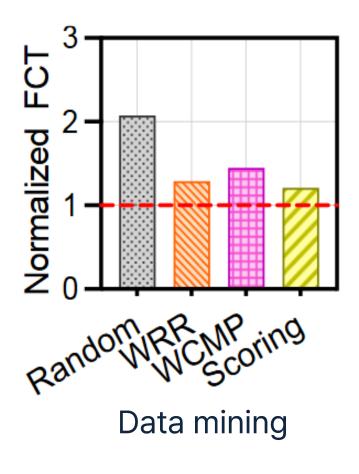
	Accuracy	Resource- efficiency
Random	X	0
WRR	0	X
WCMP	X	0
Scoring	X	0

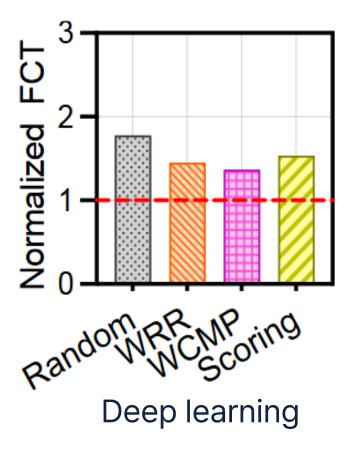
No technique achieves both high accuracy and resource-efficiency!

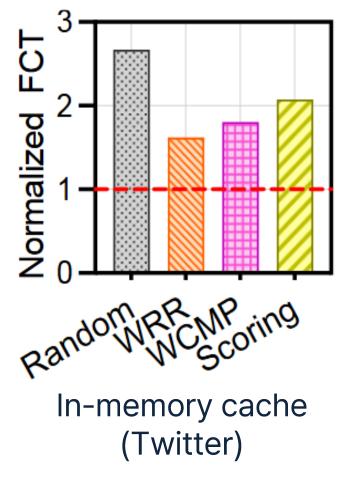
#### Lead to Poor DC Networking Services

- Inaccuracy  $\uparrow \rightarrow$  Leads to path overloading, resulting in slower speeds and packet retransmissions
- Per-packet overhead  $\uparrow \rightarrow$  Accumulates across thousands of packets, increasing total service latency
- Our experiment: Significantly delays DC services (~2.8× compared to ideal\*)





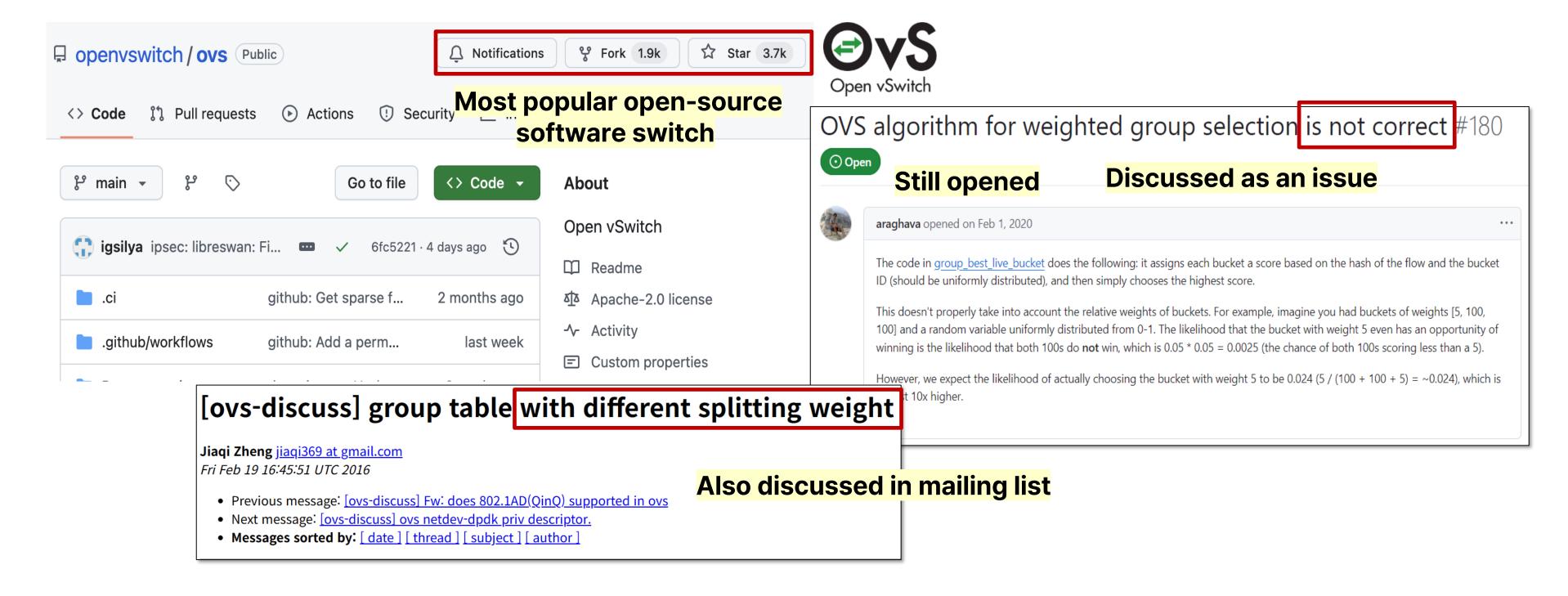




<sup>\*</sup> Ideal: factitious scenario where software switches perform perfectly and efficiently accurate traffic splitting

#### Widely Recognized Issue in Practice

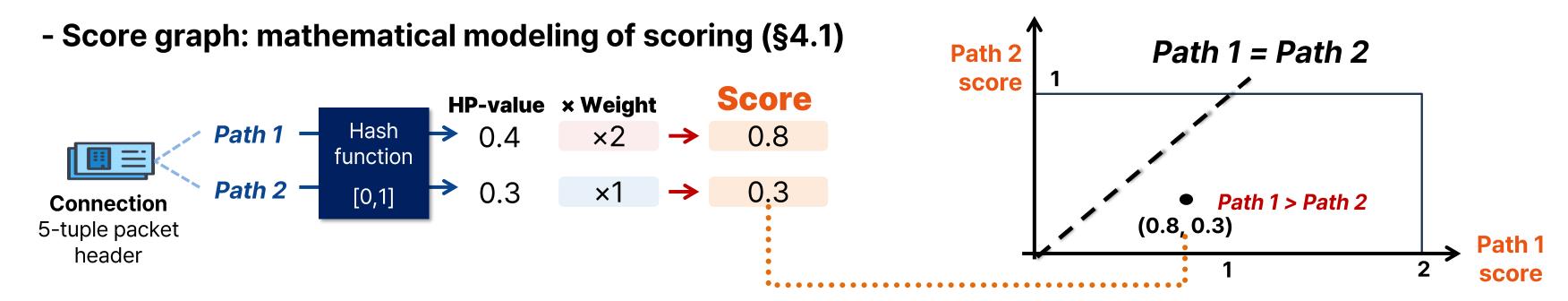
- The problem has been widely known in the open-source community, no solution exists yet
  - Inaccuracy of **scoring** is treated as black-box



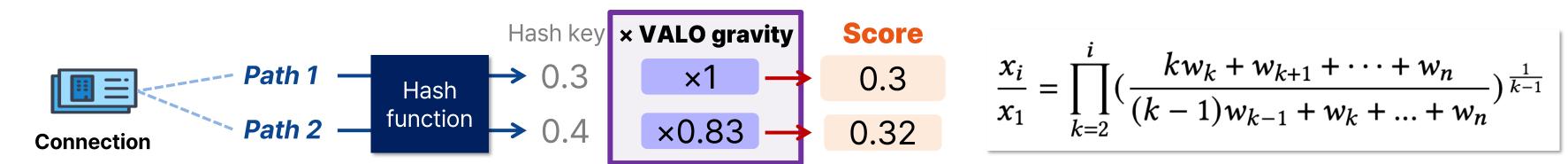
#### Propose VALO: New Traffic Splitting Mechanism

Our Goal: achieves both high accuracy and resource-efficiency (by improving scoring)

1. Modeling of scoring to identify root causes of its inaccuracy

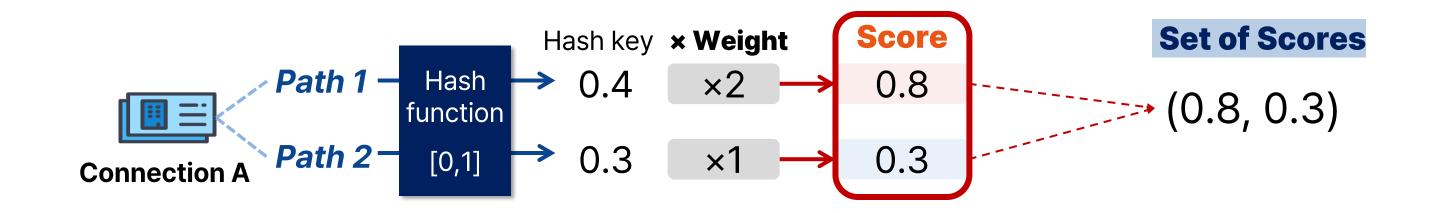


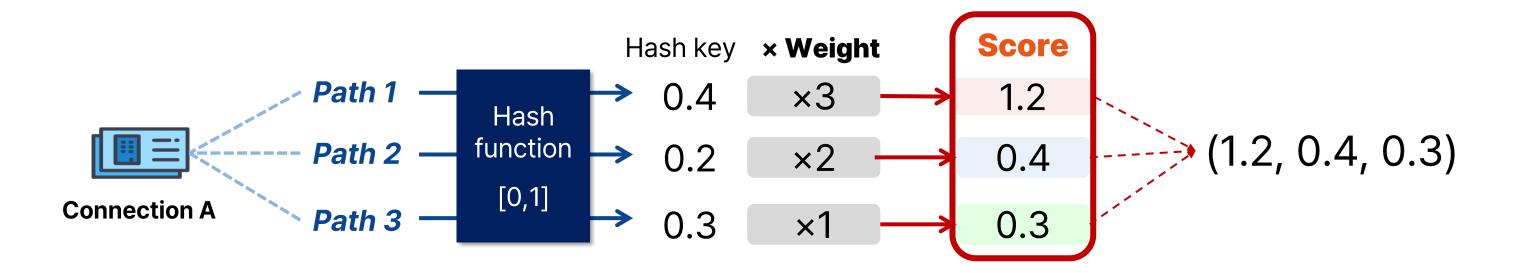
- 2. Devise VALO, incorporating its novel parameter, "VALO gravity"
  - Formulate resource-efficient method for calculating VALO gravity (§4.1)



#### **Mathematical Modeling of Scoring**

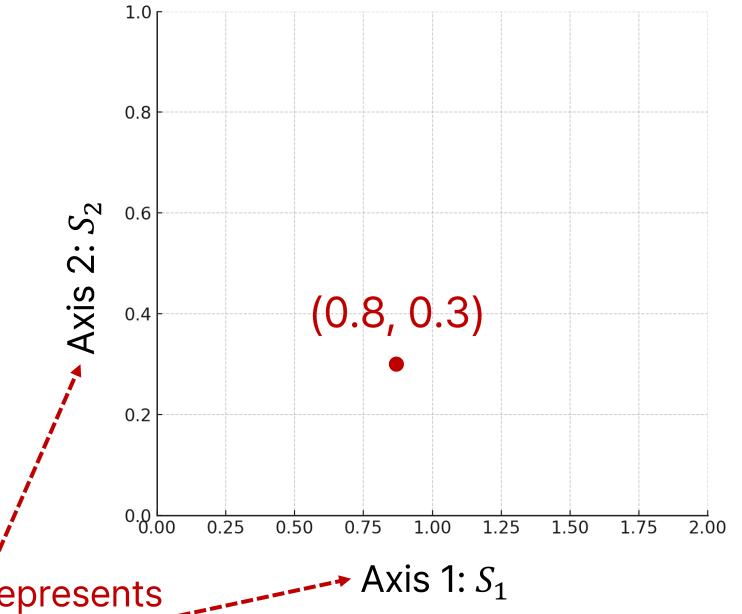
- Scoring calculates scores for all paths
- Each connection has N scores corresponding to N possible paths

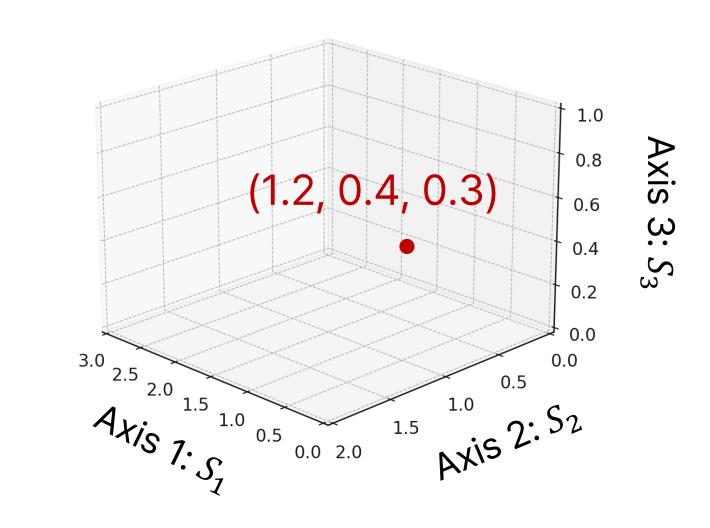




#### Each connection as a Coordinate

- Represent a connection by scores assigned to each path, denoted as  $s_i$  (score for path i):  $(s_1, s_2, ..., s_n)$ 
  - E.g., (0.8, 0.3) or (1.2, 0.4, 0.3)
- *N*-dimension Coordinate space:





Each axis represents scores of each path

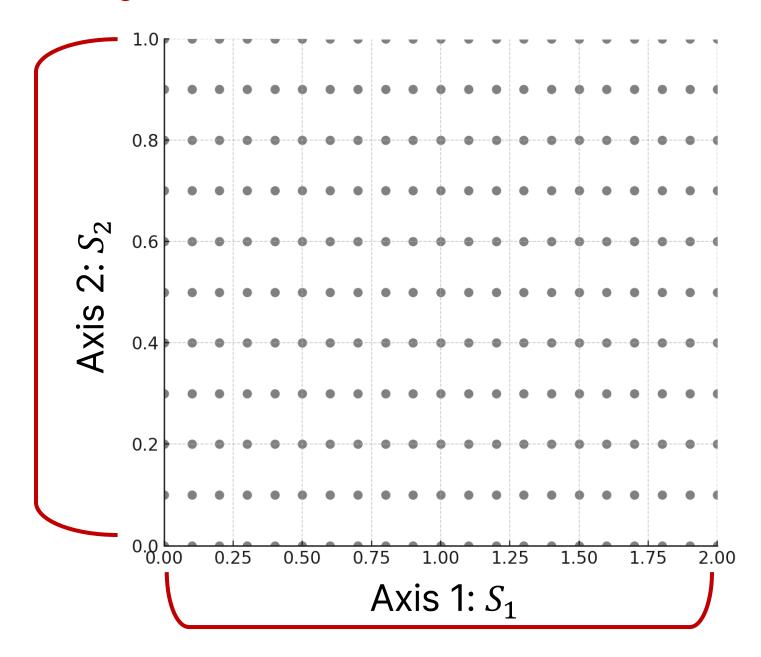
 $I_2$ : two paths exist (2-D)

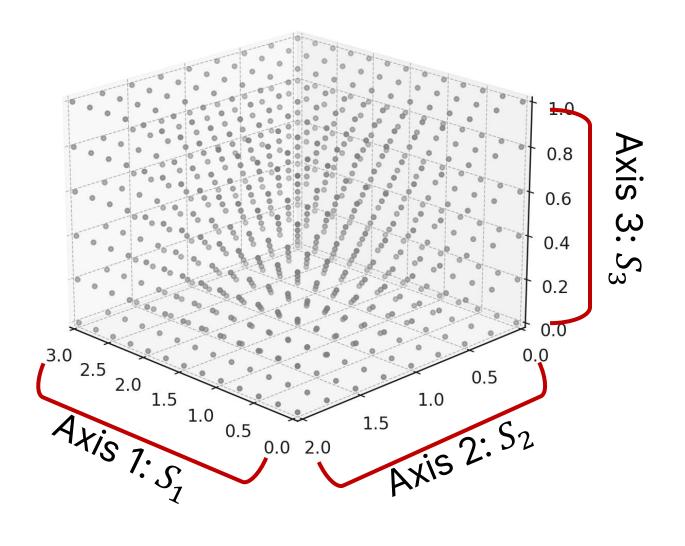
 $I_3$ : three paths exist (3-D)

#### Score Graph: for All Possible Connections

- Score graph can represents all possible network connections
  - Each axis represents the range of all possible scores for each path

#### All possible range of scores





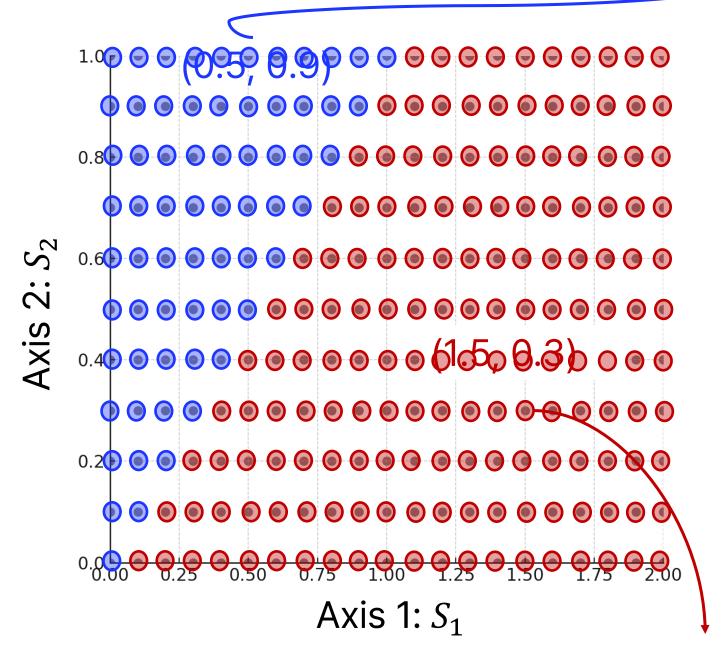
 $I_2$ : two paths exist (2-D)

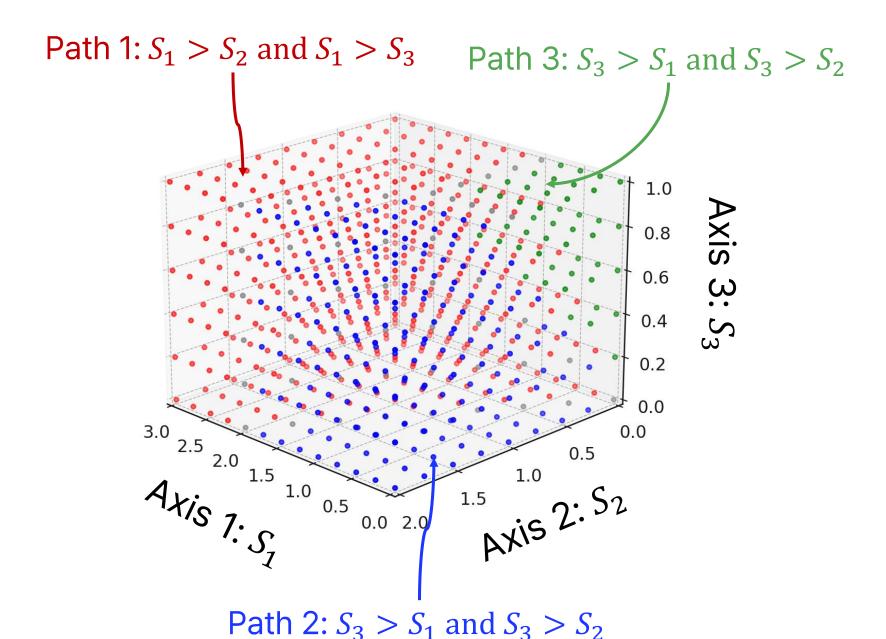
 $I_3$ : three paths exist (3-D)

#### Path Selection in Score Graph

• Connections (points) are distinguished by scoring: assigned to path with the highest score

score of path 1 (0.5) < score of path 2 (0.9)  $\rightarrow$  Select Path 2  $\rightarrow$ 





score of path 1 (1.5) > score of path 2 (0.3)  $\rightarrow$  Select Path 1

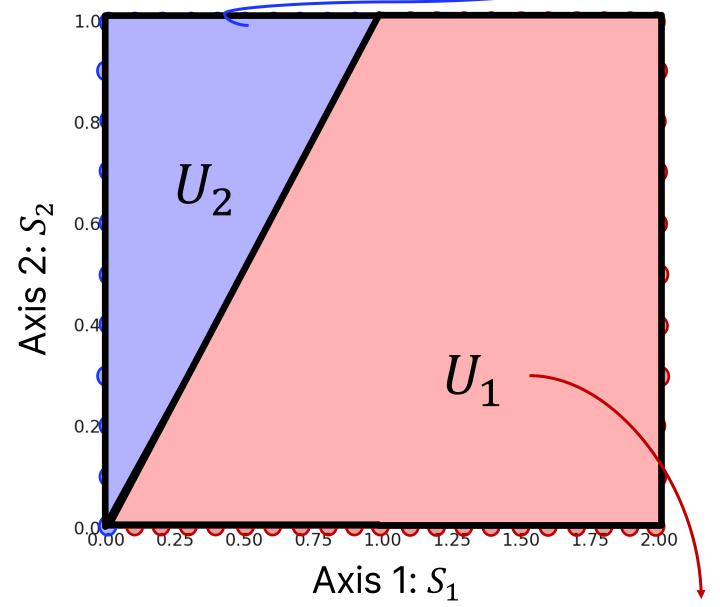
 $I_2$ :Two paths exist

 $I_3$ : Three paths exist

#### Volume: The Number of Connections per Path

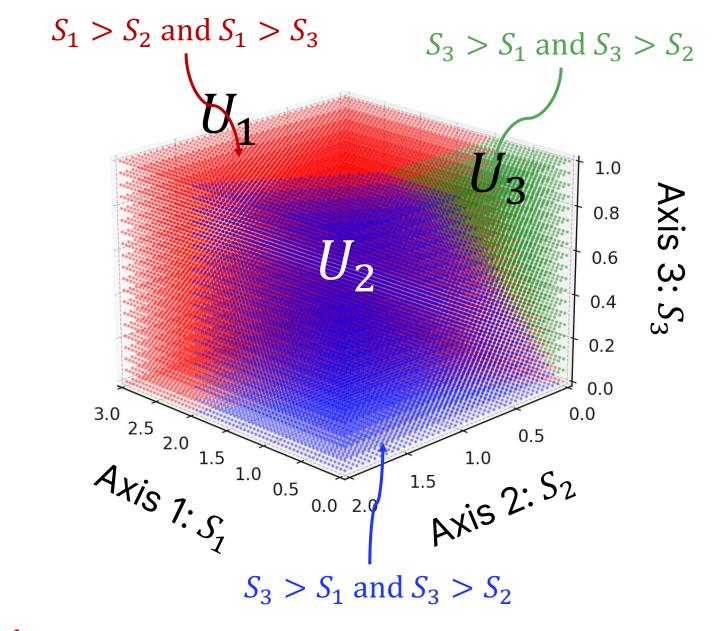
- Connections are divided into distinct **sub-area**  $(U_i)$ , each selected path i
- Volume: Total number of connections assigned to a path (e.g., calculated as the integral over area)

score of path 1 < score of path 2  $\rightarrow$  Path 2  $\rightarrow$ 





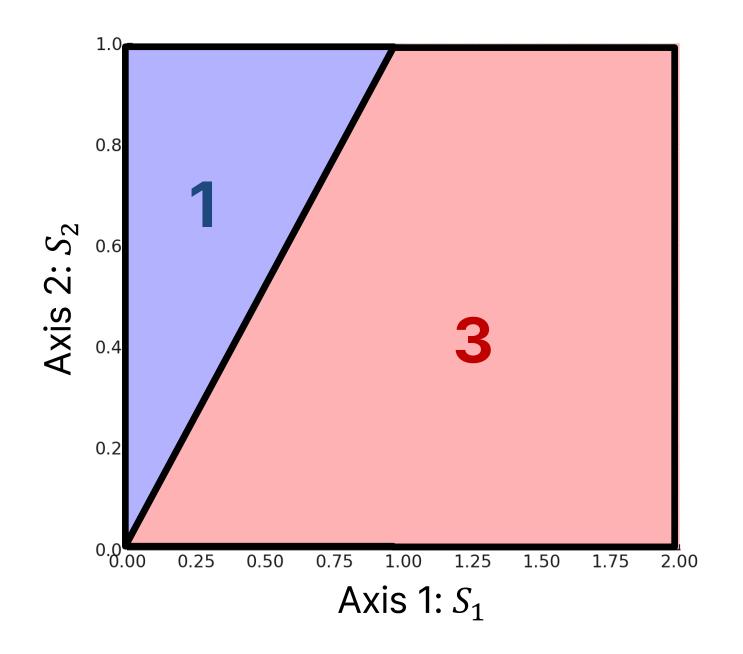
 $I_2$ : two paths exist

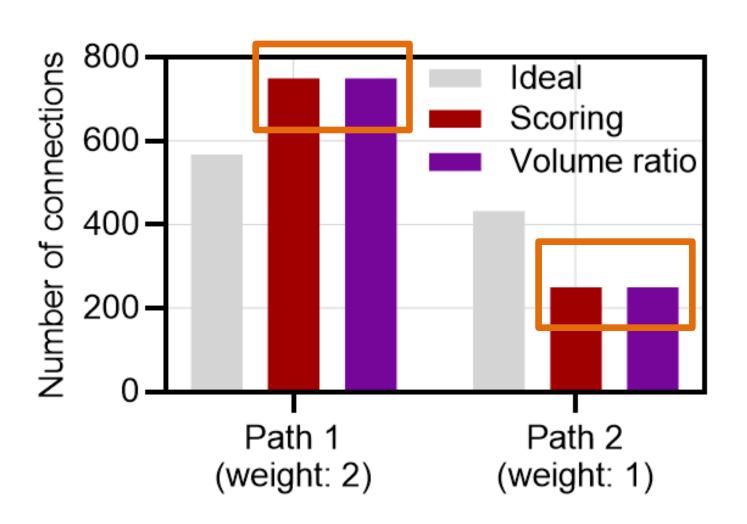


 $I_3$ : three paths exist

#### Mismatch between Volumes and Weights

- Given two paths with weights 2:1 → the volume 3:1 (mismatch)
- Actual connection split is 3:1 ratio: matching volume, not weights

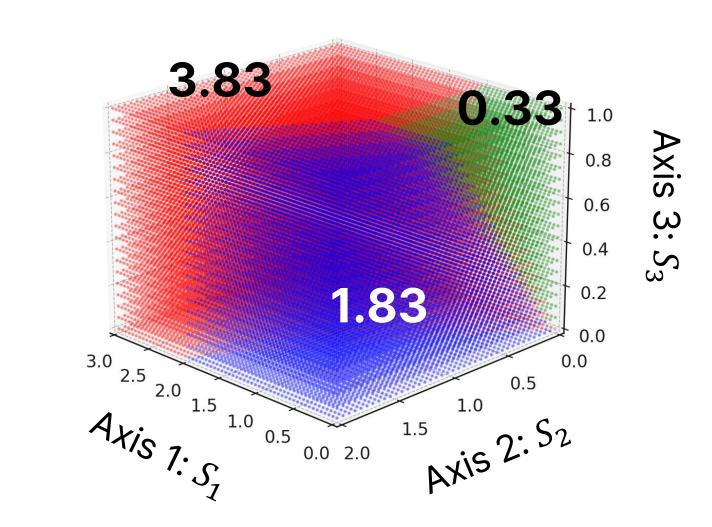


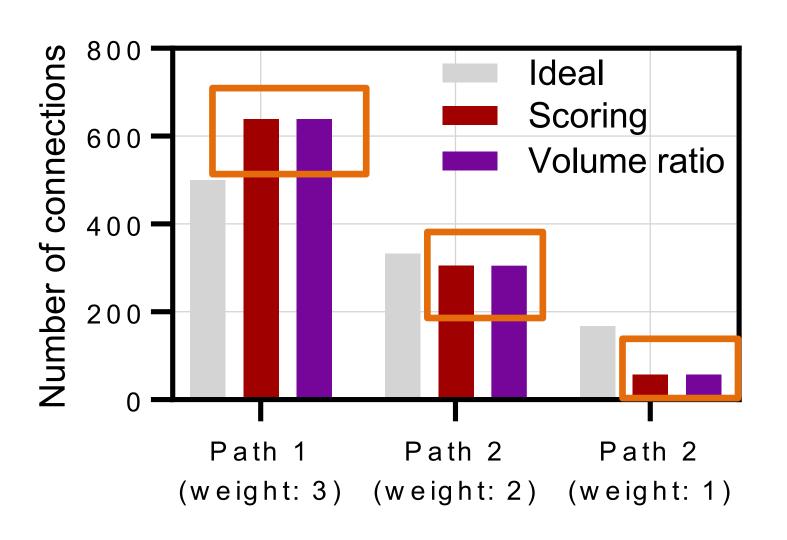


 $I_2$ : two paths exist

#### Mismatch between Volumes and Weights

- Similar mismatch for three paths:
  - For given weights 3:2:1, the volume ratios becomes 3.83:1.83:0.33
  - Actual connection split observed: 3.8:1.8:0.3



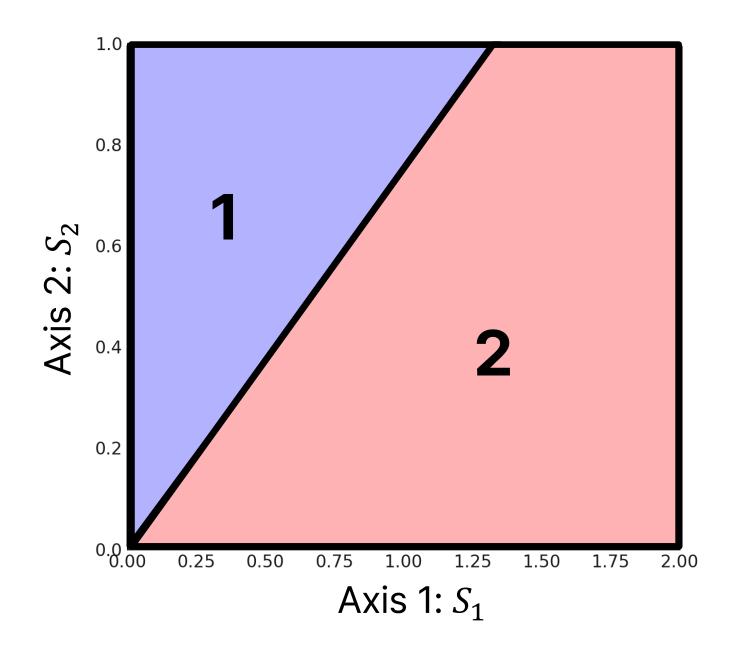


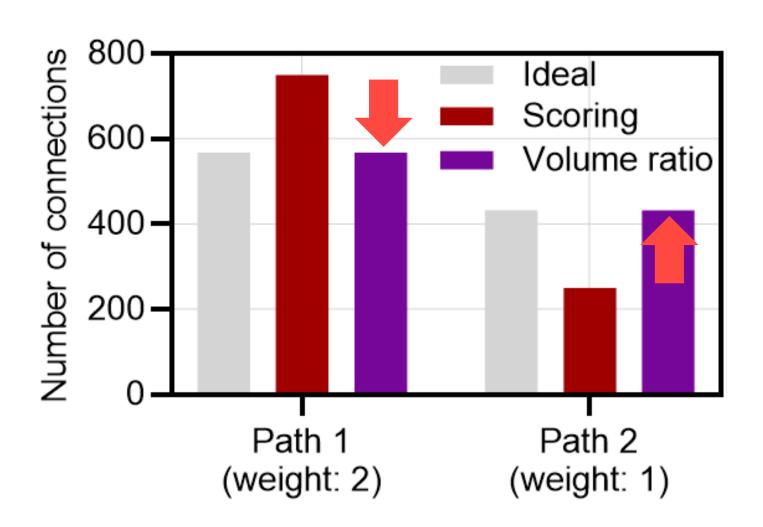
Actual splitting ratio matches volume, not weight!

: Cause of inaccuracy

## Our idea: Align Volumes with Path Weights

- Our observation: actual splitting ratio matches to "volume"
- Adjust sub-areas to ensure their volumes match path weights

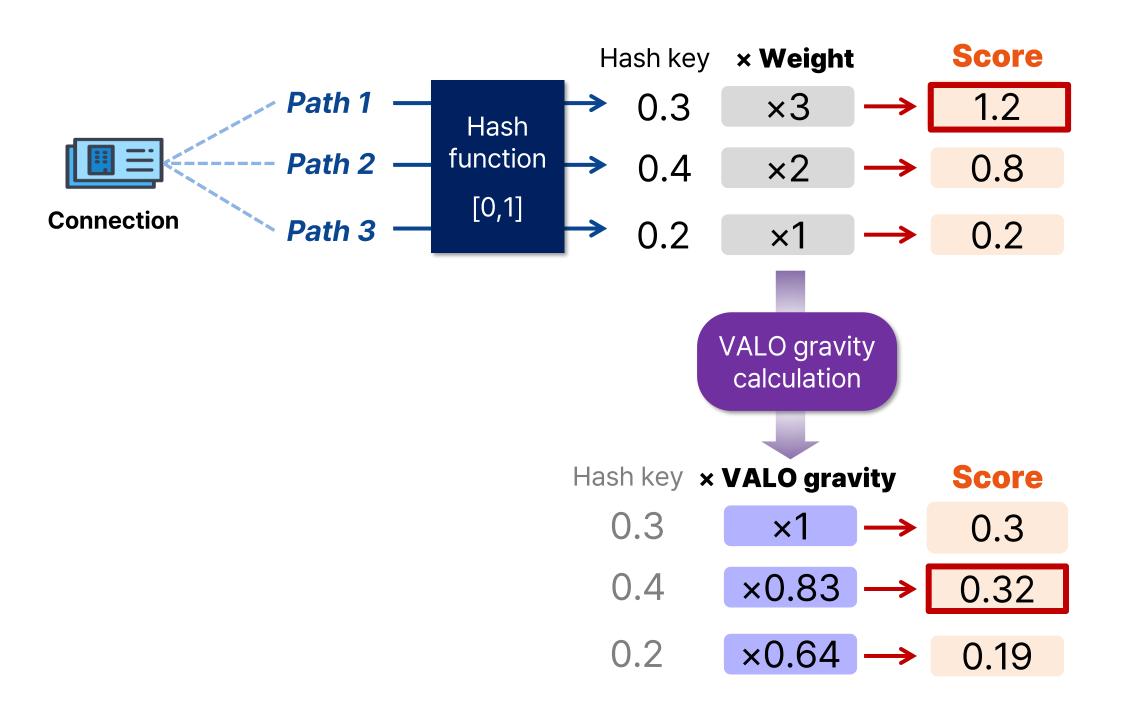




 $I_2$ : two paths exist

#### **VALO Gravity**

- Simple, but powerful parameter to align volumes accurately with given path weights
- Instead of directly multiplying path weights, multiply by "VALO gravity"



#### **VALO Gravity Calculation**

- VALO gravity: New weight to align volumes accurately with given path weights
- Optimized calculation: Efficient method with low complexity
  - Robust to higher dimension paths (e.g., 4D graph for four paths, 8D graph for eight paths)
- Derivation steps:
  - (1) Calculate path i's volume  $(vol(U_i))$  with variable path weight  $(x_i)$

$$vol(U_{n,i}) = \sum_{m=i}^{n} \frac{1}{m} (X_{n,m} - X_{n,m+1}), \quad X_{n,m} = \begin{cases} x_m^m x_{m+1} \dots x_n & n > m \\ x_n^n & n = m \\ 0 & n < m \end{cases}$$

• (2) Rearrange to above equation, express  $x_i$  in terms of the volume  $(vol(U_i))$ 

$$\frac{x_i}{x_1} = \prod_{k=2}^{i} \left( \frac{kvol(U_{n,k}) + vol(U_{n,k+1}) + \dots + vol(U_{n,n})}{(k-1)vol(U_{n,k-1}) + vol(U_{n,k}) + \dots + vol(U_{n,n})} \right)^{\frac{1}{k-1}}$$

• (3) Substituting volumes with path weights  $(w_i)$ , calculating  $x_i$  as VALO gravity

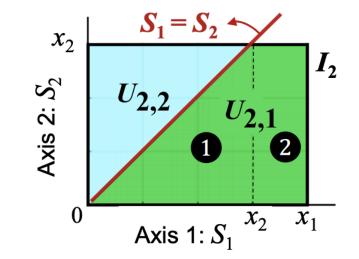
Optimize calculation by simple arithmetic operations

$$\frac{x_i}{x_1} = \prod_{k=2}^{i} \left( \frac{kw_k + w_{k+1} + \dots + w_n}{(k-1)w_{k-1} + w_k + \dots + w_n} \right)^{\frac{1}{k-1}}$$

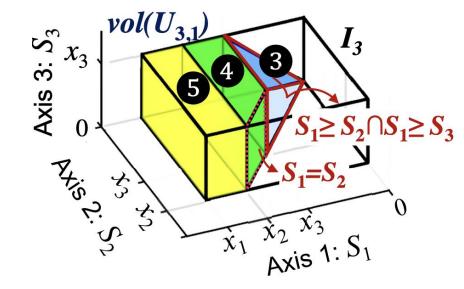
#### (1) Volume Calculation

- Derive path volume  $(vol(U_i))$  from path weight  $(x_i)$
- Use mathematical induction to generalize volume patterns as the number of paths increases

(2 Paths exist)



(3 Paths exist)



$$vol(U_{2,1}) = \frac{1}{2}x_2^2 + (x_1x_2 - x_2^2)$$

$$vol(U_{2,2}) = \frac{1}{2}x_2^2$$

$$vol(U_{2,2}) = \frac{1}{2} x_2^2$$

$$3 + 4 + 5$$

$$vol(U_{3,1}) = \frac{1}{3}x_3^3 + \frac{1}{2}(x_2^2x_3 - x_3^3) + (x_1x_2x_3 - x_2^2x_3)$$

$$vol(U_{3,2}) = \frac{1}{3}x_3^3 + \frac{1}{2}(x_2^2x_3 - x_3^3)$$

$$vol(U_{3,3}) = \frac{1}{3}x_3^3$$

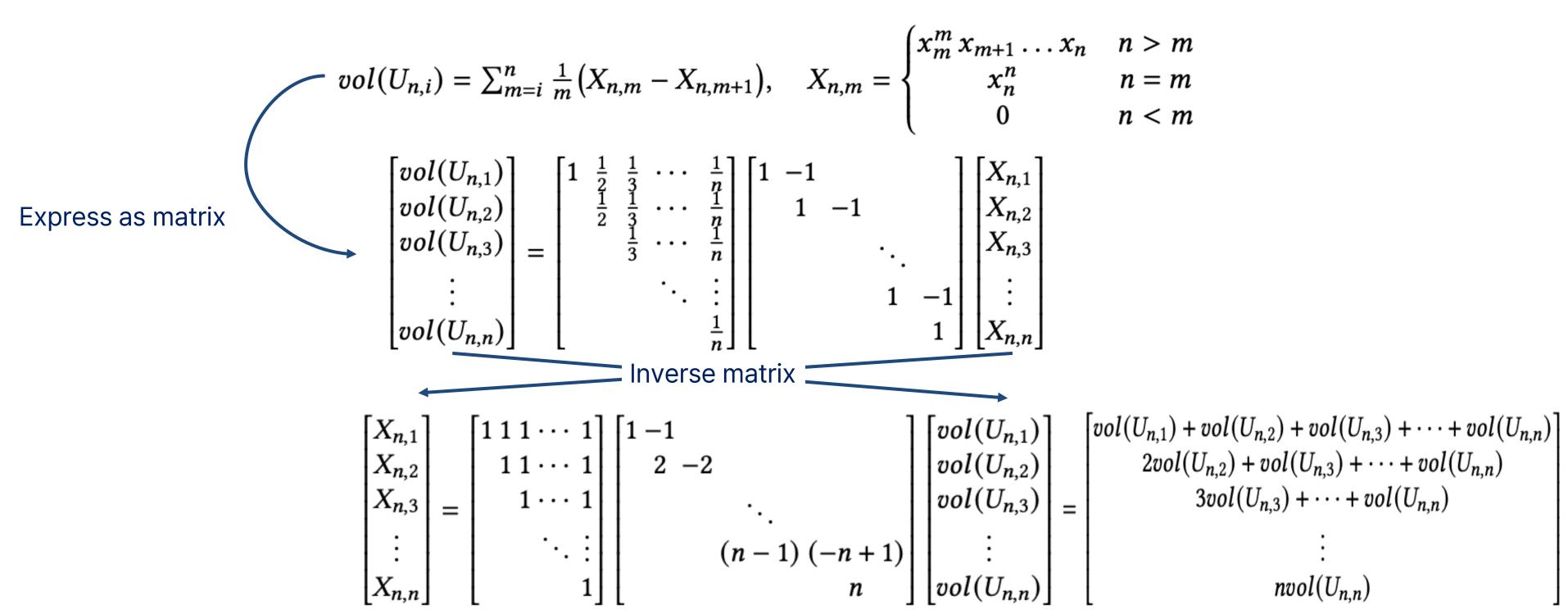
Mathematical induction (proof in §4.2.2)

(*n* Paths exist) 
$$vol(U_{n,i}) =$$

$$vol(U_{n,i}) = \sum_{m=i}^{n} \frac{1}{m} (X_{n,m} - X_{n,m+1}), \quad X_{n,m} = \begin{cases} x_m^m x_{m+1} \dots x_n & n > m \\ x_n^n & n = m \\ 0 & n < m \end{cases}$$

#### (2) Weight Calculation

- Reorganize equations to express  $x_i$  explicitly in terms of the volume  $(vol(U_i))$
- Using Inverse matrix operations to simplify complex calculations



## (2) Weight Calculation

Define weights as 
$$\frac{x_i}{x_1} = \begin{bmatrix} vol(U_{n,1}) + vol(U_{n,2}) + vol(U_{n,3}) + \cdots + vol(U_{n,n}) \\ 2vol(U_{n,2}) + vol(U_{n,3}) + \cdots + vol(U_{n,n}) \\ 3vol(U_{n,3}) + \cdots + vol(U_{n,n}) \\ \vdots \\ nvol(U_{n,n}) \end{bmatrix}$$

$$\frac{x_2}{x_1} = \frac{x_2^2 \times \cdots \times x_n}{x_1 \times x_2 \times \cdots \times x_n} = \frac{X_{n,2}}{X_{n,1}} = \frac{2vol(U_{n,2}) + \cdots + vol(U_{n,n})}{vol(U_{n,1}) + \cdots + vol(U_{n,n})}$$

$$\frac{x_3}{x_1} = \frac{X_{n,2}}{X_{n,1}} \times (\frac{X_{n,3}}{X_{n,2}})^{\frac{1}{2}} = \frac{x_2}{x_1} \times (\frac{3vol(U_{n,3}) + \cdots + vol(U_{n,n})}{2vol(U_{n,2}) + \cdots + vol(U_{n,n})})^{\frac{1}{2}}$$

$$\frac{x_4}{x_1} = \frac{X_{n,2}}{X_{n,1}} \times (\frac{X_{n,3}}{X_{n,2}})^{\frac{1}{2}} \times (\frac{X_{n,4}}{X_{n,3}})^{\frac{1}{3}} = \frac{x_3}{x_1} \times (\frac{4vol(U_{n,4}) + \cdots + vol(U_{n,n})}{3vol(U_{n,3}) + \cdots + vol(U_{n,n})})^{\frac{1}{3}}$$

Generalize to n paths

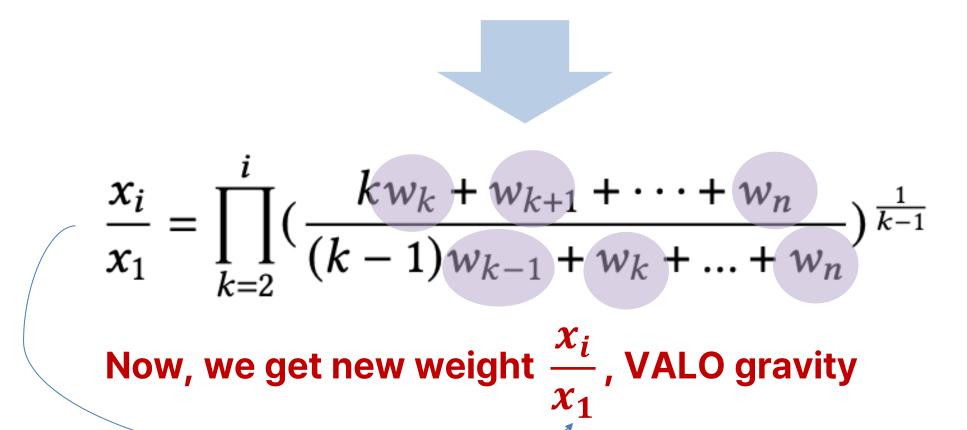
$$\frac{x_i}{x_1} = \prod_{k=2}^{i} \left( \frac{kvol(U_{n,k}) + vol(U_{n,k+1}) + \dots + vol(U_{n,n})}{(k-1)vol(U_{n,k-1}) + vol(U_{n,k}) + \dots + vol(U_{n,n})} \right)^{\frac{1}{k-1}}$$

#### (3) VALO Gravity Calculation

Finally, weights are expressed in terms of the volume of each path

$$\frac{x_i}{x_1} = \prod_{k=2}^{i} \left( \frac{kvol(U_{n,k}) + vol(U_{n,k+1}) + \dots + vol(U_{n,n})}{(k-1)vol(U_{n,k-1}) + vol(U_{n,k}) + \dots + vol(U_{n,n})} \right)^{\frac{1}{k-1}}$$

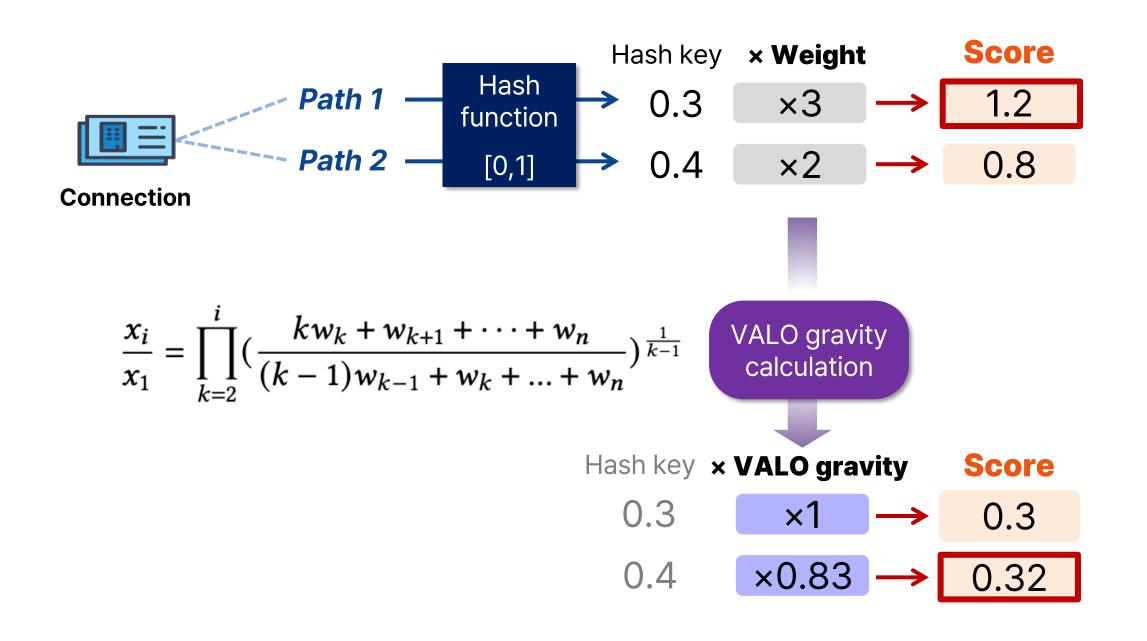
Instead of volume  $(vol(U_{n,i}))$ , substitute given path weights  $(w_i)$ 

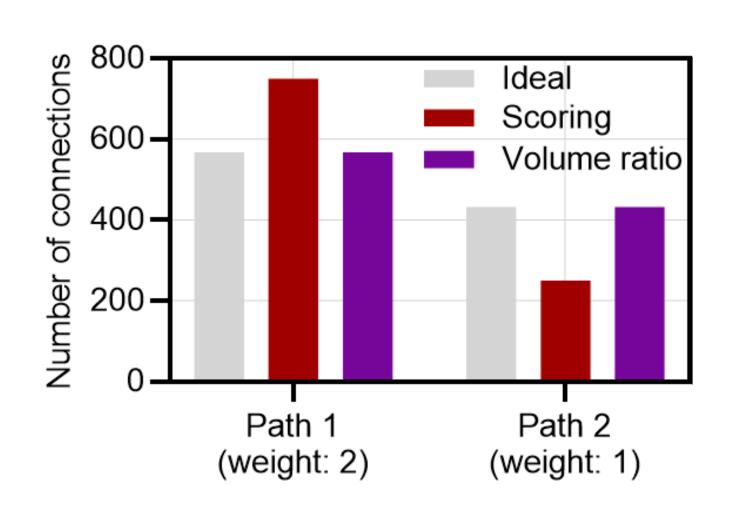


Calculated by only simple arithmetic operations

#### VALO Workflow Summary

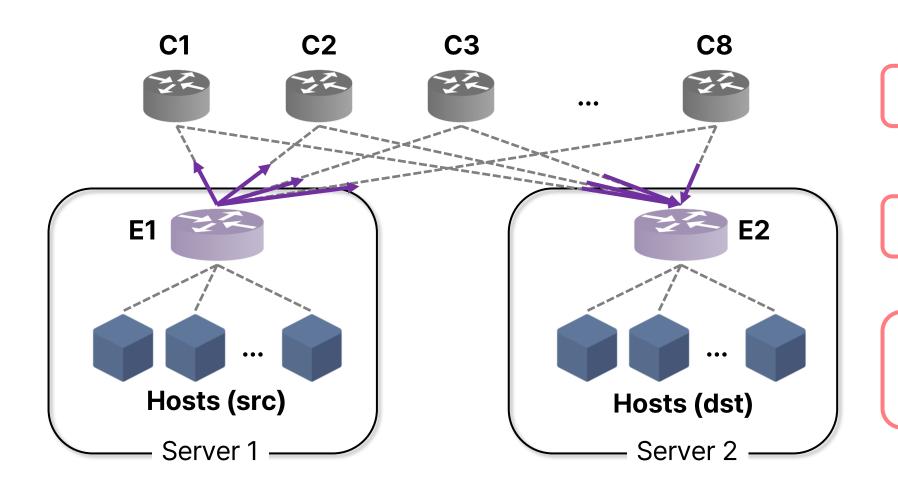
Through VALO gravity, the number of connections matches with the path weights





#### **Evaluation**

- Topology: Two-tier DC topology (core switches increased 2–8), using ~32 containers
- Compare five techniques: random, WRR, WCMP, scoring, and VALO (implemented on OVS)
- Workloads: 1) DC traffic traces (CAIDA, ClassBench) and 2) Real-world DC workloads
- Measurement: 1) accuracy, resource-efficiency and 2) end-to-end latency of DC services



- 2–8 paths exist

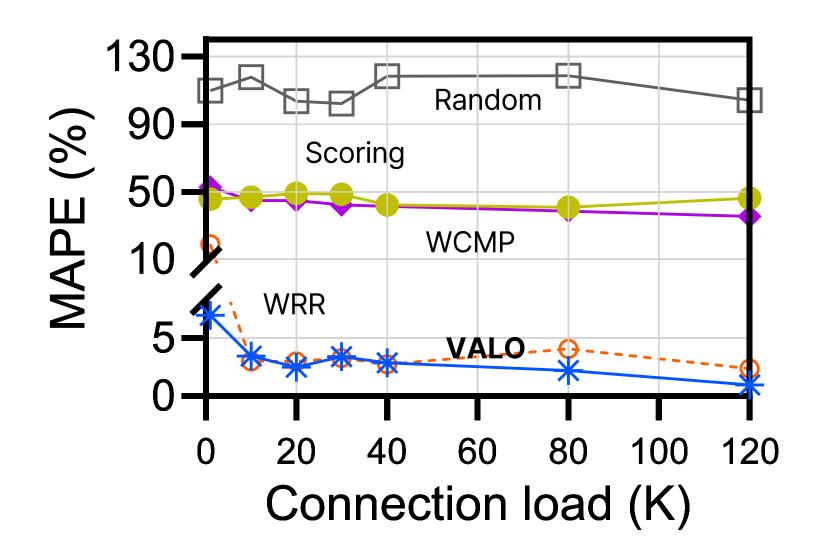
- Random, WRR, WCMP, scoring, VALO

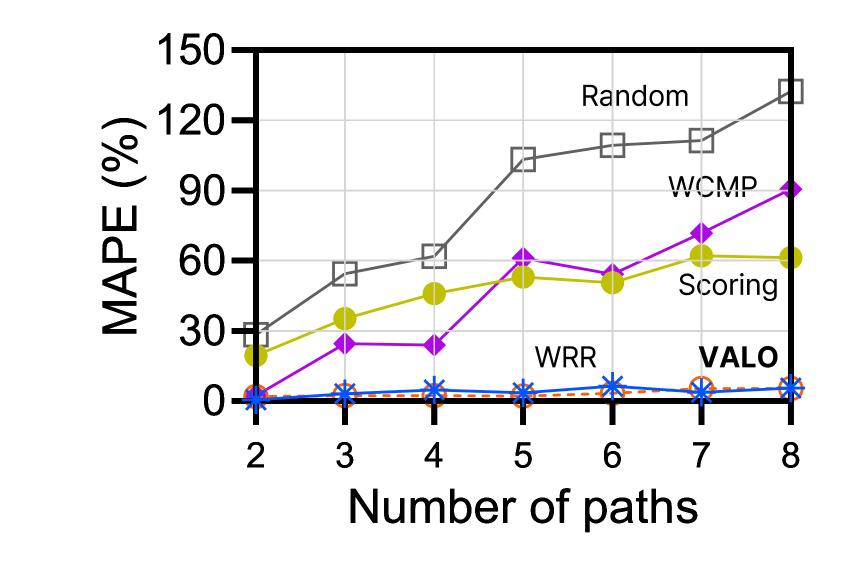
- CAIDA, ClassBench
- Web search, data mining, deep learning, in-memory cache

## **Traffic Splitting Accuracy**

• Measure error rate (MAPE) between ideal  $(C_i)$  and actual  $(\widehat{C}_i)$  connection distributions

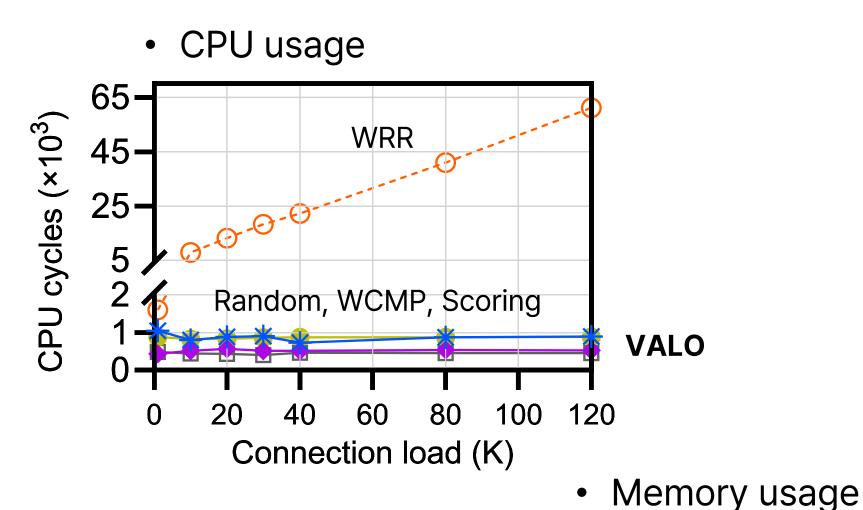
$$MAPE = \frac{100}{n} \times \sum_{i=1}^{n} \frac{|C_i - \hat{C}_i|}{C_i}$$



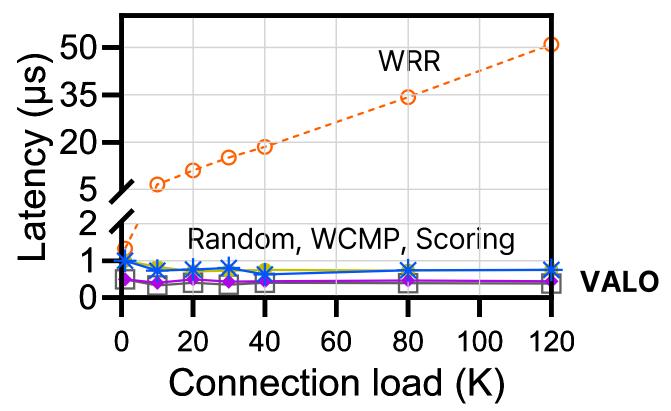


VALO keeps low accuracy (3.1% on average) in all cases (~46.3× improvement)

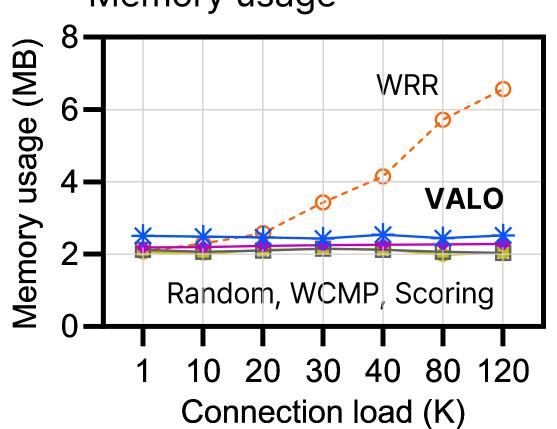
#### Resource-efficiency



Per-packet latency



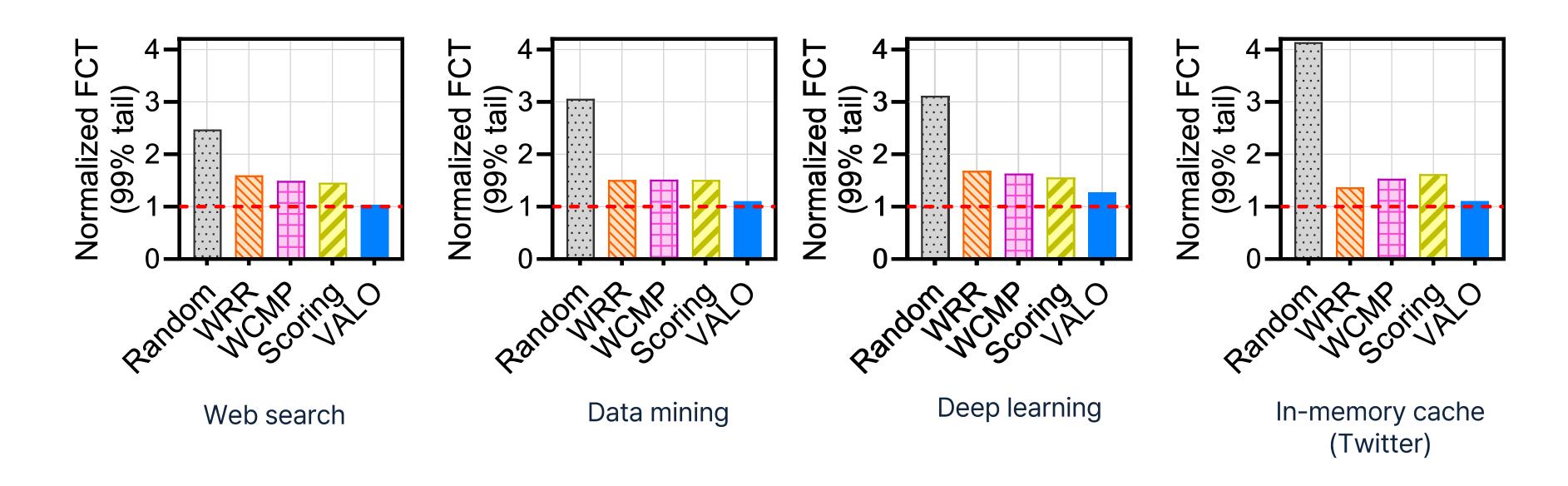
VALO shows a similar level of resource-efficiency as random, WCMP, and scoring



VALO does not sacrifice efficiency to achieve high accuracy!

#### Flow Completion Time of DC Services

- VALO achieves the shortest flow completion time (FCT) on real-world DC service workloads
  - VALO improves 99% tail latency by ~2.8×



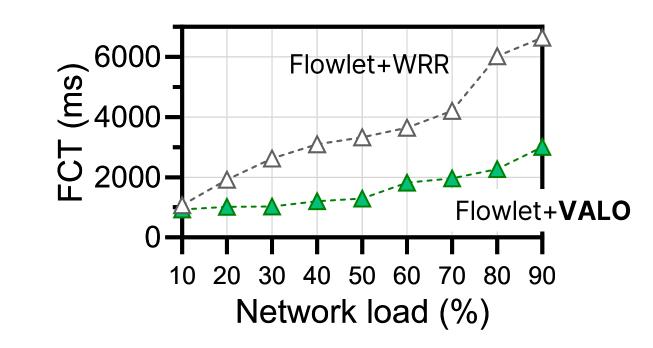
#### Conclusion

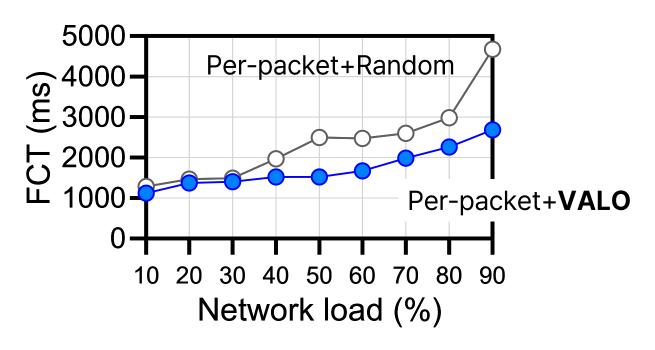
#### Summary

- Major result: Achieves **accuracy** by  $\sim 46.3 \times$  and **CPU usage** by  $\sim 10.7 \times \rightarrow$  accelerate DC services ( $\sim 2.8 \times$ )
- Approach 1: Analyze root-cause of inaccuracy of scoring with score graph
- Approach 2: Find new internal weight, VALO gravity that align volumes to path weights

#### VALO can integrate with other load balancing techniques (§5.5)

Not only per-connection, VALO can work at <u>finer granularity</u> (e,g., flowlet, per-packet)





#### Artifact

- VALO implemented and evaluated on de-facto software switch (Open vSwitch of Linux foundation)
- Our codes are available at GitHub!
  - https://github.com/yeonhooy/VALO-OVS-SIGMETRICS25.git



# Thank you

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