

# Resource Provisioning in Optical Networks

Invited Talk @IIT Indore, India

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Knowledge for Tomorrow

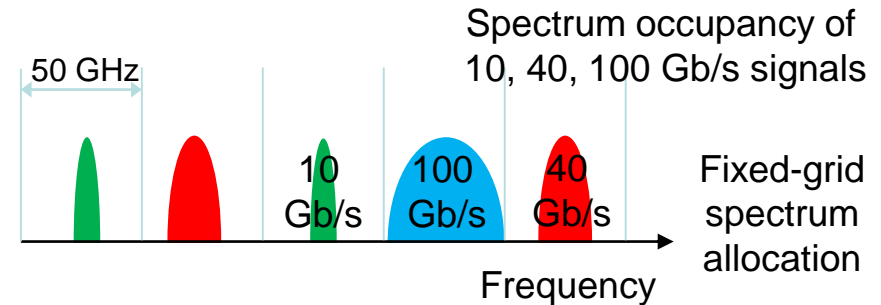


# A Brief Overview of Optical Network Evolution

- ❑ Internet traffic growth, Big-data, Cloud computing and research networks driving deployment of Terabit/s

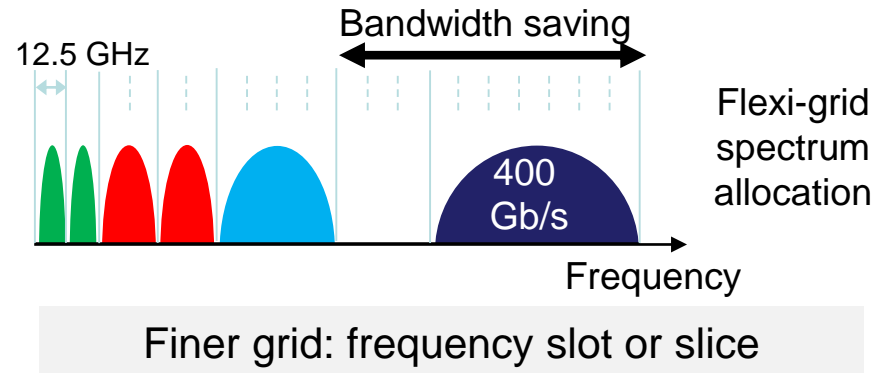
- ❑ Wavelength-division multiplexing

- ❖ Used in backbone network (Gbit/s)
- ❖ Tbit/s spectrum too broad to fit
- ❖ 50/100 GHz fixed grid
- ❖ Bandwidth wastage



- ❑ Elastic optical network (EON)

- ❖ Finer and flexi-grid (6.25/12.5 GHz)
- ❖ BVT (OFDM-based), switches
- ❖ Better spectrum utilization
- ❖ Superchannel support (Tbit/s)
- ❖ Distance-adaptive modulation



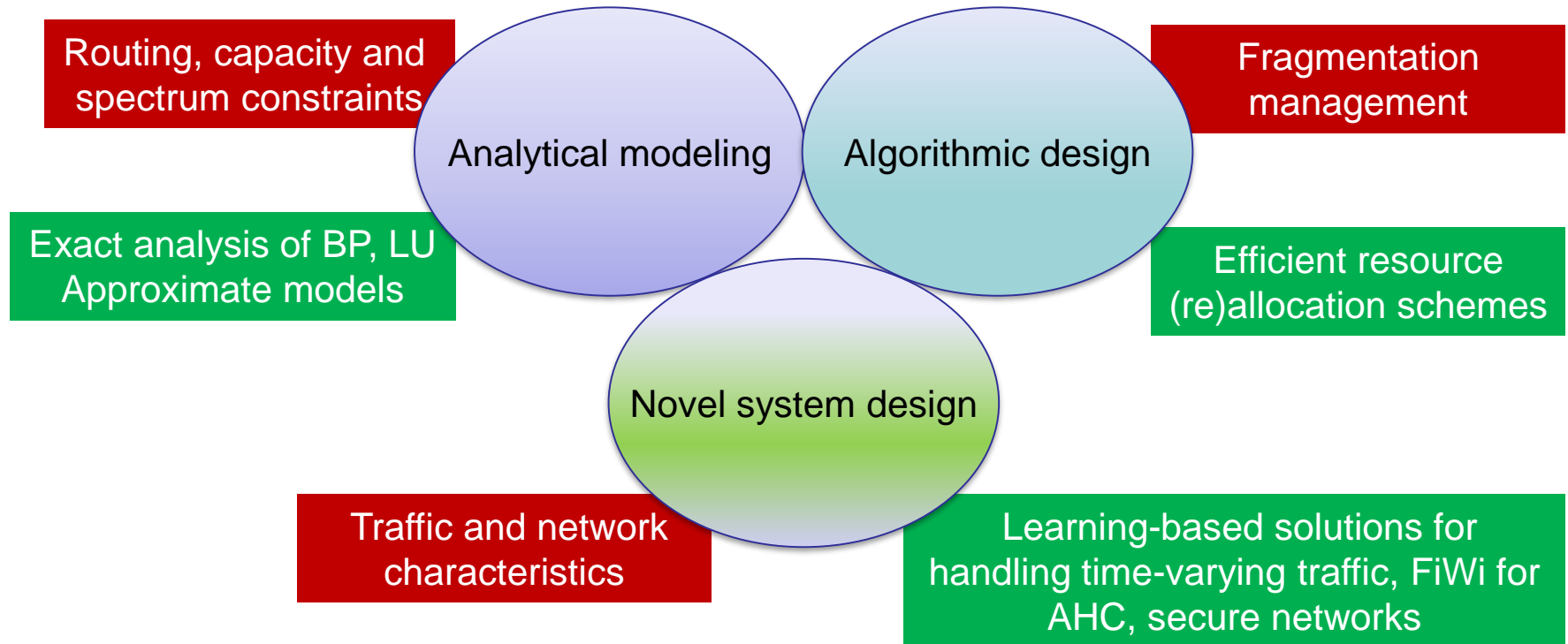
## Elastic Optical Networking is Promising

- P. Winzer, "Optical Networking Beyond WDM," *IEEE Photonics Journal*, vol. 4, no. 2, pp. 647–651, 2012.
- O. Gerstel, M. Jinno, A. Lord, and S. B. Yoo, "Elastic optical networking: A new dawn for the optical layer?" *Communications Magazine*, IEEE, vol. 50, no. 2, pp. s12{s20, 2012.

# Resource Provisioning is a key Challenge in EONs

**Resources:** fibers, spectrum, transponders, etc. ➡ Lightpath (optical channel) establishment

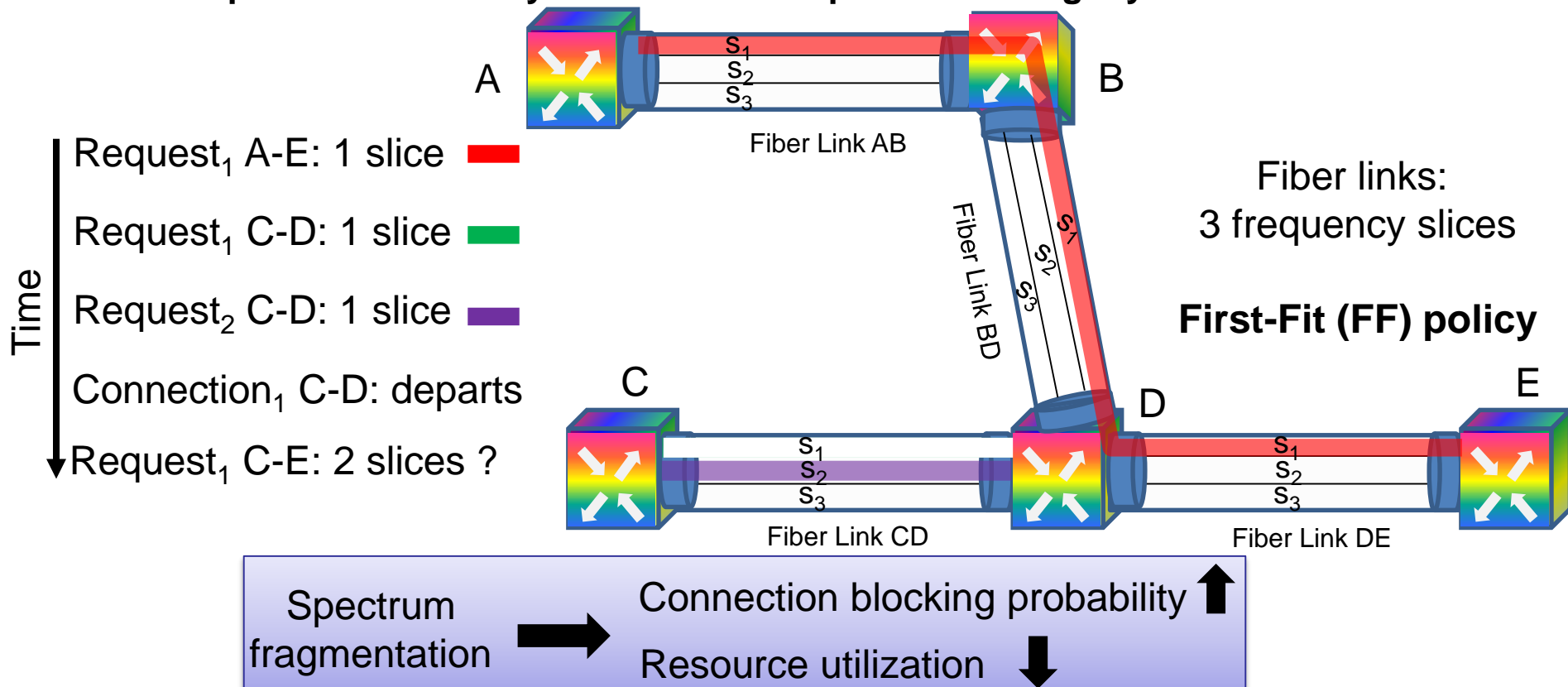
**Challenges:** latency, impairments, routing, spectrum allocation, faults, security, etc.



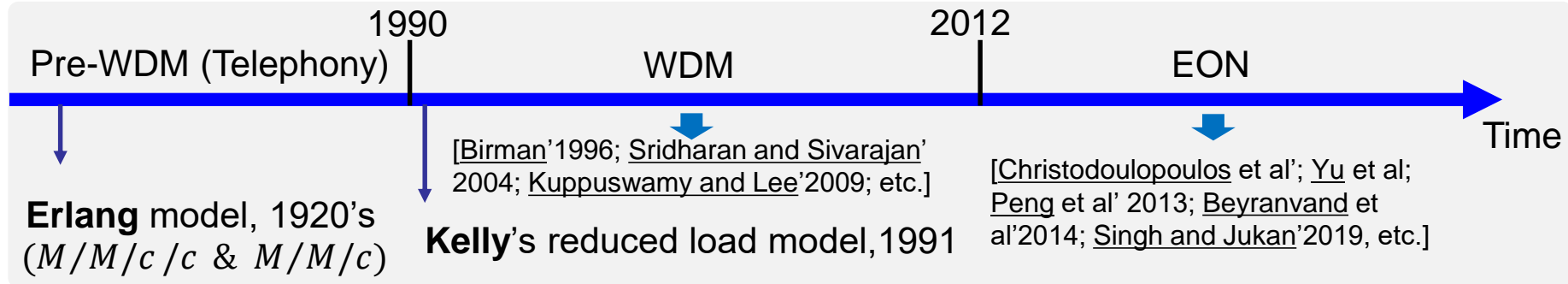
# Challenges with Routing & Spectrum Allocation in EONs

## ❑ Routing and Spectrum Allocation (RSA) for lightpath establishment

- ❖ **Routing:** Fixed, Fixed-alternate, Dynamic
- ❖ **Spectrum allocation:** Random-Fit (RF), First-Fit (FF), Fragmentation-aware
- ❖ **Spectrum continuity constraint** and **Spectrum contiguity constraint**



# RSA Modeling in Elastic Optical Networks (Stochastic)



## Exact models: accurate and useful for small networks

\*H. Beyranvand et al, IEEE TOC 2014 => a so-called (inaccurate) exact network Markov model

\*\*S. K. Singh and A. Jukan, INFOCOM 2019 => Exact network Markov model

## Approximate Models: less accurate, but scalable

❖ Erlang-based models: Connection setup rate on a link  $j$  is  $\alpha_k^j(x) = \sum_{o:j \in r(o)} \lambda_k^o$

❖ Reduced load independence model  $\alpha_k^j(x) = \sum_{o:j \in r(o)} \lambda_k^o \times \underbrace{\Pr[Z_{r(o)} \geq d_k | X_j = x_j]}_{\text{Reduced load contribution}}$

❖ Reduced load correlation model: open problem

Reduced load contribution

original external offered rate thinned by capability of demand acceptance

\* H. Beyranvand et al, "An analytical framework for the performance evaluation of in EONs," IEEE Trans. on Comm., 2014. 5

\*\* S. K. Singh, and A. Jukan, "Computing Blocking Probabilities in EONs with Spectrum Defragmentation," INFOCOM, 2019.

# RSA Modeling in EONs: Our Approach

- ❑ Existing stochastic RSA models ignore key issues\*
- ❑ Our work: Exact Network Markov model for small scale EONs, and Reduced Load Approximation models for large scale EONs
- ❑ Assumptions and limitations
  - ❖ Poisson arrival ( $\lambda_k^o$ ) & exponential departure ( $\mu_k$ )
  - ❖ Fixed-alternate routing; Spectrum allocation: Random-Fit (RF) or First-Fit (FF)
  - ❖ A class-k connection request is blocked if there are insufficient ( $\leq d_k$ ) contiguous and continuous free slices on its routes.

## ❑ Notations and parameters

$C$	Capacity of each fiber link (in slices)
$K$	Number of bandwidth classes, $k=1,2,\dots, K$
$d_k$	Bandwidth demand of class-k connection (in slices)
$r(o)$	Routes between an origin-destination (OD) node-pair $o = \{1,2, \dots,  O \}$
$\lambda_k^o$	Class-k connection request arrival rate on OD pair $o$
$\mu_k$	Class-k connection departure rate

\* 1.H. Beyranvand, M. Maier, and J. Salehi, "An analytical framework for...", IEEE Trans. on Comm., vol. 62, no. 5, 2014.

2. L. Peng, C.-H. Yoon, and C. Qiao, "Theoretical analyses of lightpath blocking performance in co-ofdm optical networks with/without spectrum conversion," IEEE Communications Letters, vol. 17, no. 4, pp. 789-792, 2013.

# Exact Network Markov Model

## Exact Network Markov Model (Complexity: Non-polynomial)

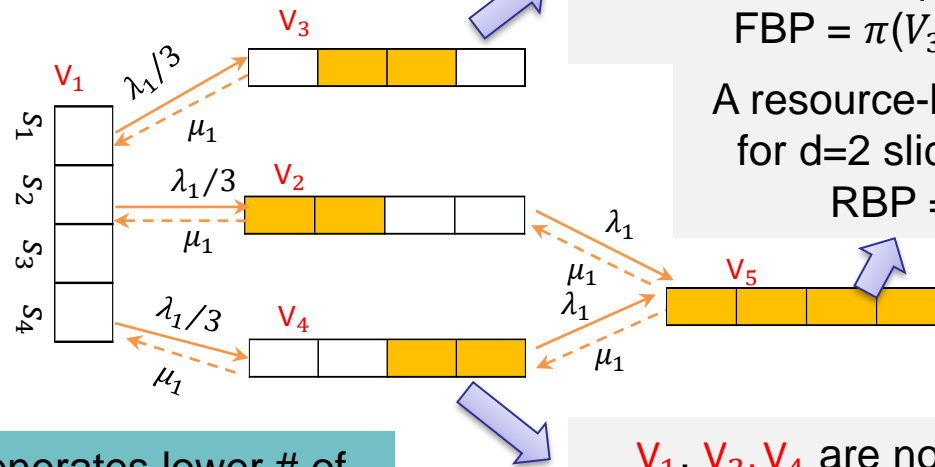
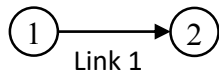
❖ Define network states:  $V_i \equiv (S^1, S^2, \dots, S^{|J|})$ , where a link state  $S^j \equiv (s_1^j, s_2^j, \dots, s_C^j)$

$s_i^j$  = free/busy slice

❖ Find transition rates, obtain stationary network state probabilities

❖ Compute blocking probability

**Single-link,  
C=4 slices,  $d_1=2$  slices,  
Random-Fit (RF) policy**



A fragmentation state for  
 $d=2$  slices requests  
 $FBP = \pi(V_3)$

A resource-blocking state  
for  $d=2$  slices requests  
 $RBP = \pi(V_5)$

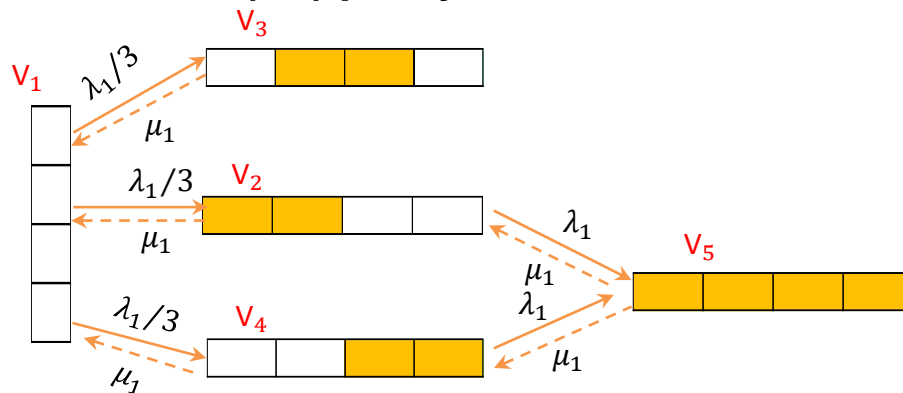
First-Fit (FF) generally generates lower # of  
fragmentation states

$V_1, V_2, V_4$  are non-blocking  
states for  $d=2$  slices requests

# Approximate Network Models: Our Approach

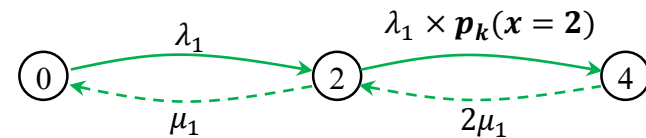
- ❑ A Link state ( $X_j$ ) is represented by # of occupied slices ( $x_j$ )
- ❑ Compute connection setup rates and departure rates for each link  $j$
- ❑ Apply fixed-point iteration algorithm to compute state probabilities & BP

**Single-link,  
C=4 slices,  $d_1=2$  slices,  
Random-Fit (RF) policy**



Exact model

Approximation model



**Probability of Acceptance:**

$$p_k(x) = \Pr[Z_{r(o)} \geq d_k | X_j = x_j]$$



# Approximate Probability of Acceptance Computation

## □ Ours

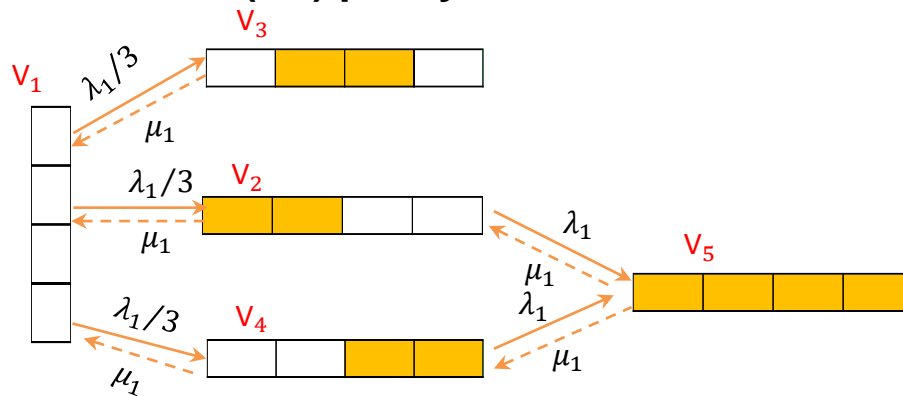
1. Equiprobable Exact States (EES)
2. Spectrum Occupancy Correlation (SOC)
3. Uniform approximation

## □ Others\*

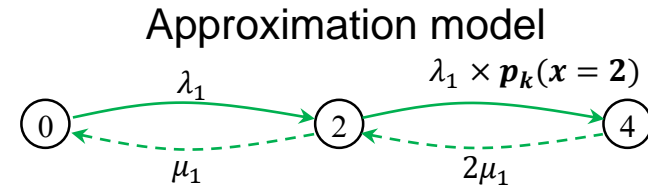
1. Kaufman-based
2. Binomial approach

Simpler, not Reduced load model

**Single-link,  
C=4 slices,  $d_1=2$  slices,  
Random-Fit (RF) policy**



Ignores valid  
spectrum patterns



Approximation methods	Prob. of Accept. $p_k(x=2)$
EES	2/3
SOC	Load-dependent
Uniform	3/6
Kaufman	1
Binomial	Load-dependent

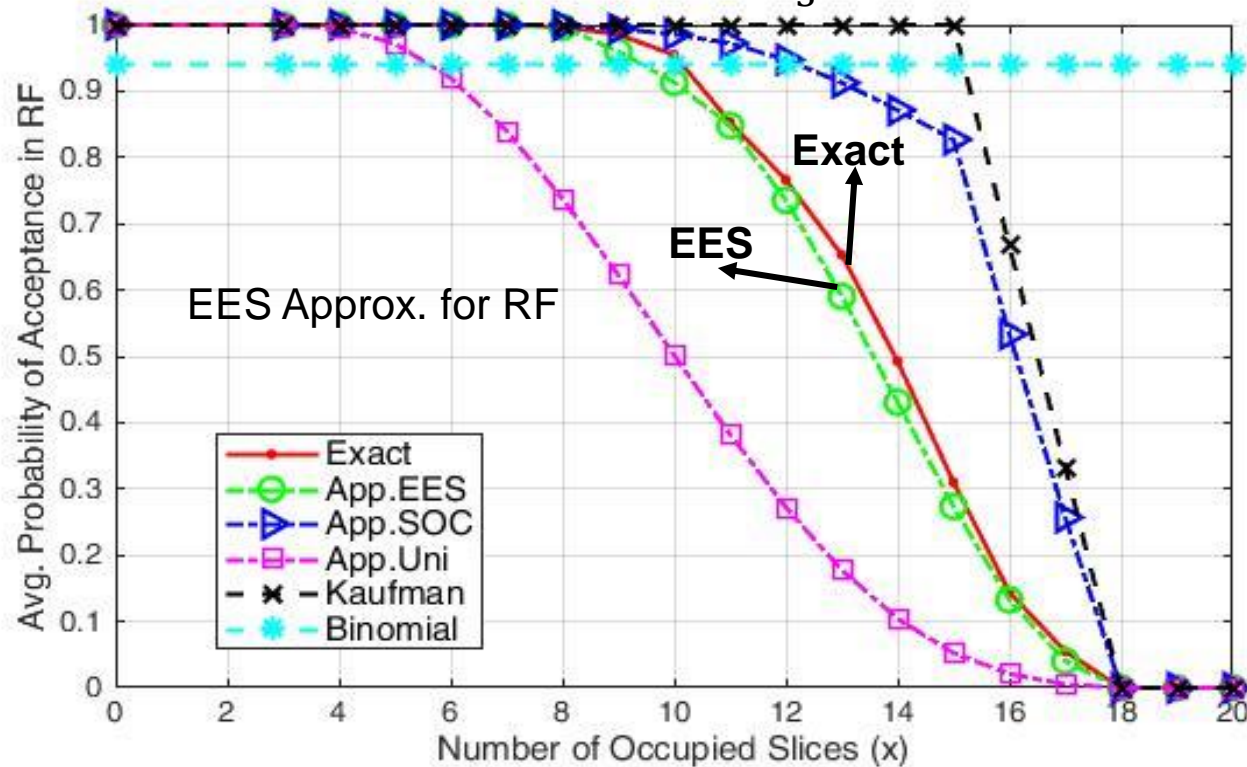
\*[1] H. Beyranvand et al, "An analytical framework for the performance evaluation in EONs," IEEE Trans. on Comm., 2014. 9

[2] L. Peng et al, "Theoretical analyses of lightpath blocking performance ...in EONs," IEEE Comm. Lett, 2013.

# Comparison of Various Approximations under RF

□ Link capacity  $C = 20$  slices; Demands  $(d_k) = \{3, 4, 5\}$  slices, offered load = 1.2 Erlang;

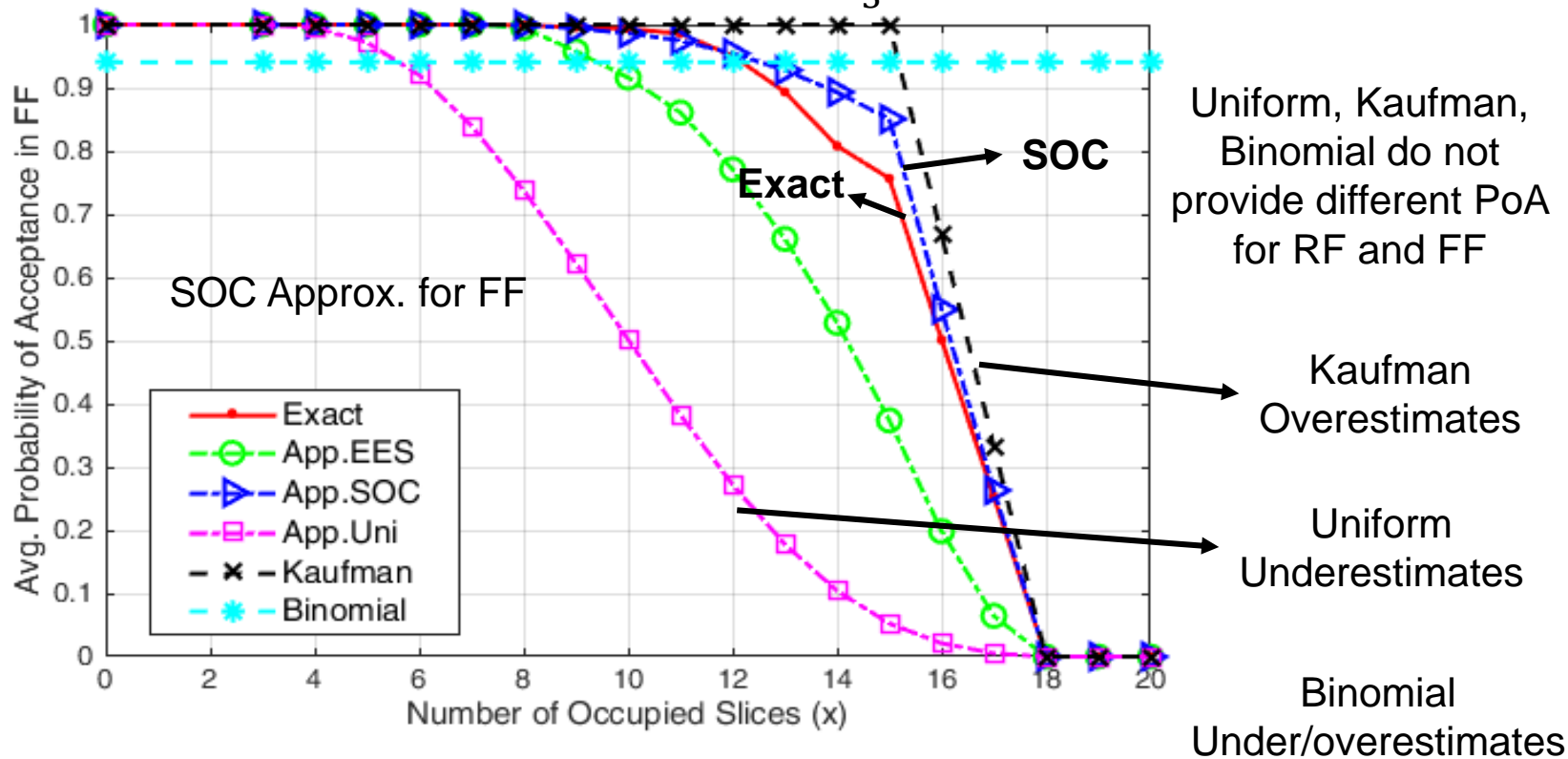
□ Avg. probability of acceptance  $p(x) = \frac{1}{3} \sum_{k=1}^3 p_k(x)$



# Comparison of Various Approximations under FF

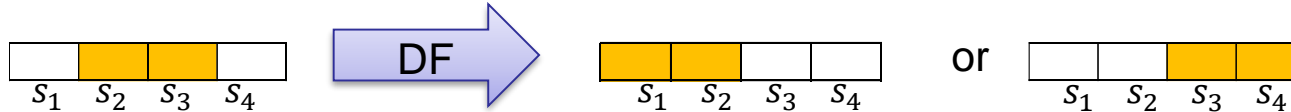
□ Link capacity  $C = 20$  slices; Demands  $(d_k) = \{3, 4, 5\}$  slices, offered load = 1.2 Erlang;

□ Avg. probability of acceptance PoA  $p(x) = \frac{1}{3} \sum_{k=1}^3 p_k(x)$

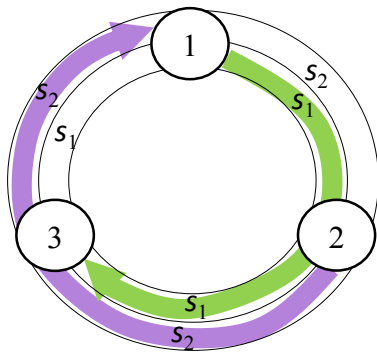


# Spectrum Fragmentation is Inevitable in EONs

- ❑ Spectrum defragmentation is a solution
- ❑ **Defragmentation (DF):** A resource reallocation scheme to reconfigure connections in EONs to consolidate a large block of free slices



- ❑ Assumptions\*: reactive trigger, exponential reconfiguration time
- ❑ **Exact DF model:** include DF states, network transits to least-fragmentation state(s) after reconfiguration and accepts of waiting connection(s)
- ❑ Some fragmented states cannot be defragmented in ring/mesh networks



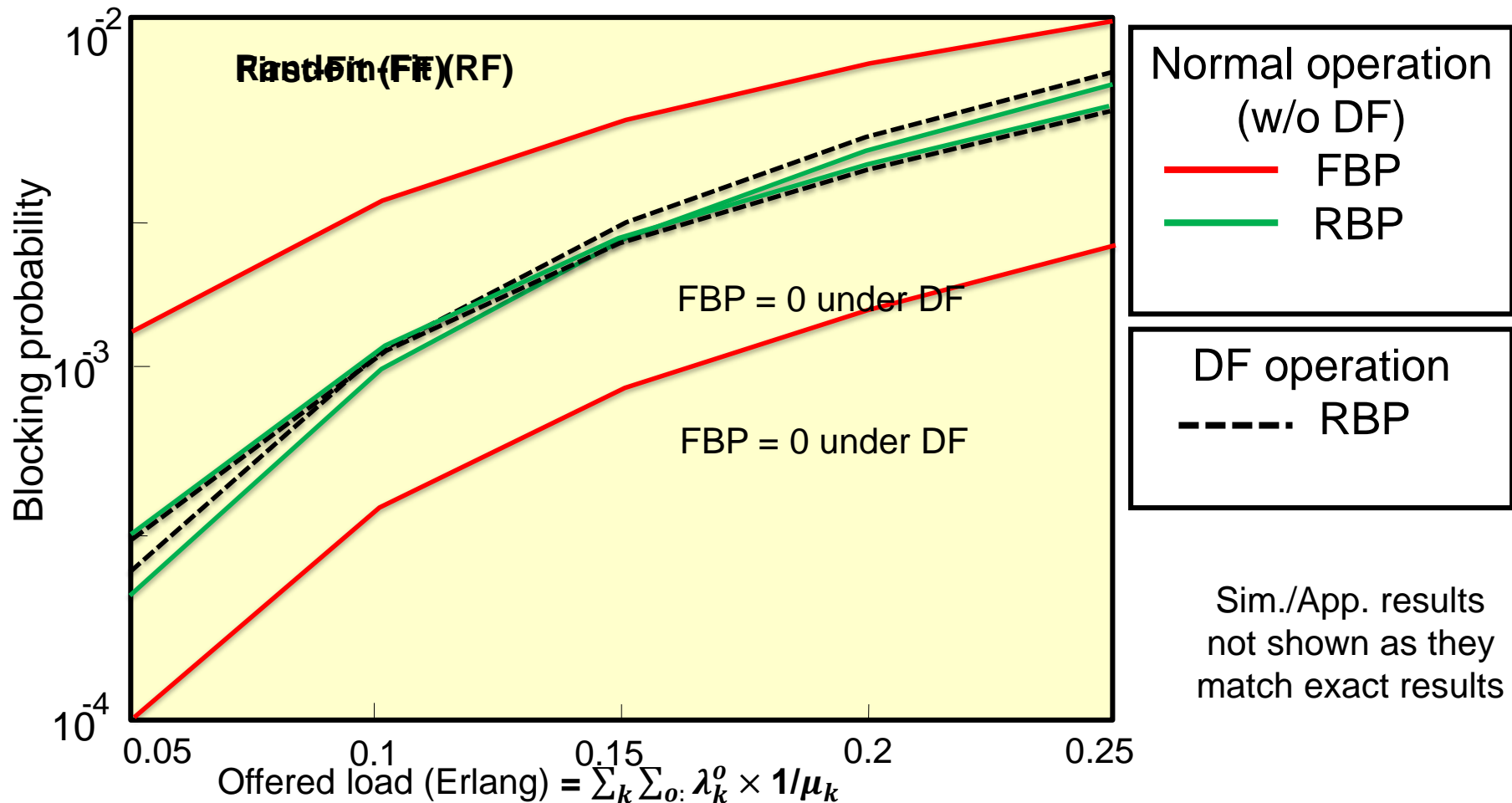
**Approx. Probability of Acceptance under DF:**

$$p_{k,DF}^{App.EES}(x_r) = \prod_{i=1}^h \frac{|\mathbb{NB}(x_{j_i}, k)|}{|\Omega_S(x_{j_i})|} + \prod_{j_i \in r: \mathbb{FI}(x_{j_i}, k) \neq 0} \frac{|\mathbb{FI}(x_{j_i}, k)|}{|\Omega_S(x_{j_i})|}$$

# Impact of DF in Networks with Bus Topology

## 2-link network (Exact results)

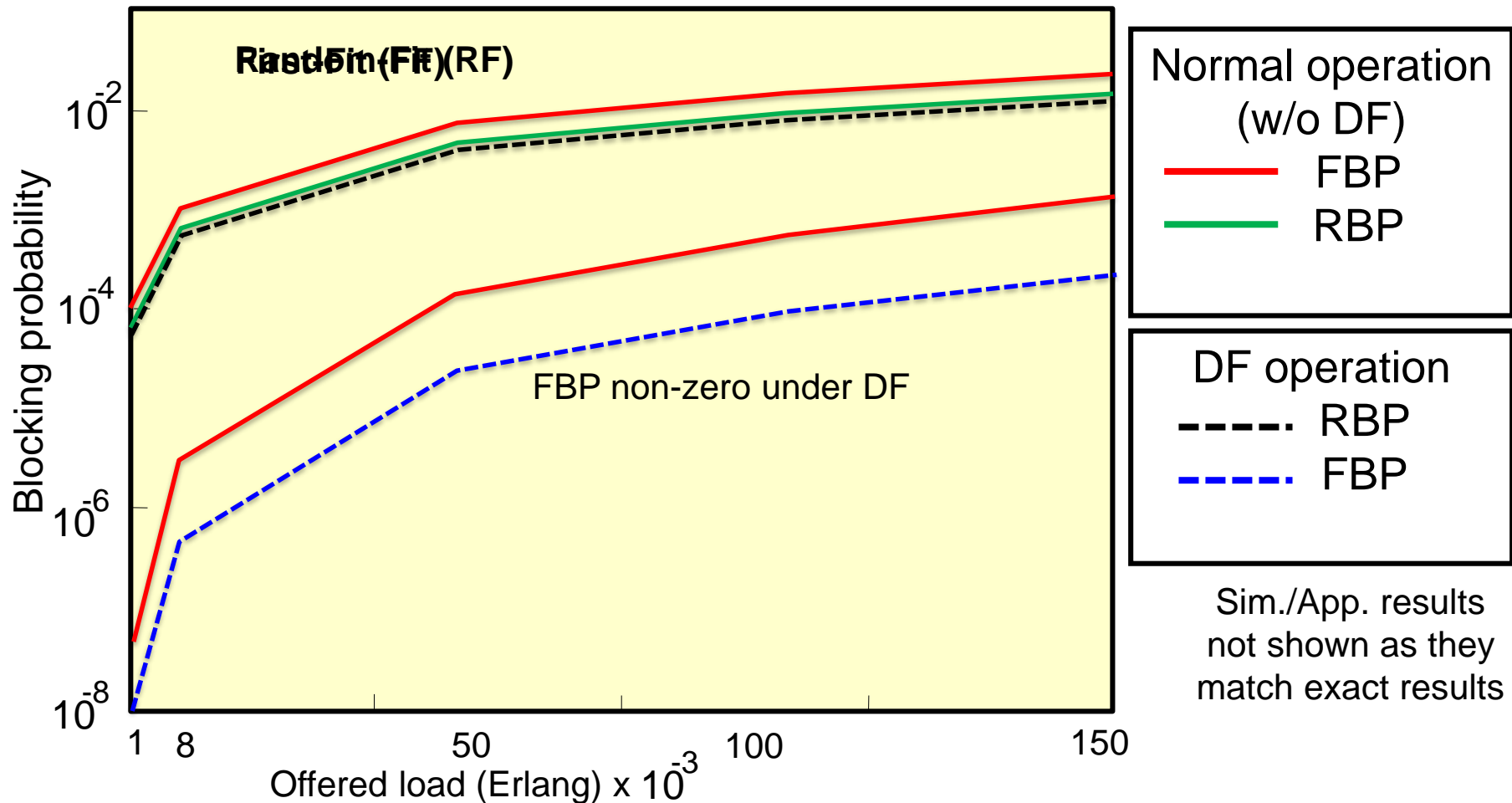
Link capacity  $C=10$  slices, demands  $(d) = (3,4)$  slices,  $\mu_k=1$ , mean reconfig time=0.001



# Impact of DF in Networks in Ring/Mesh Topology

## 3-node ring network (Exact results)

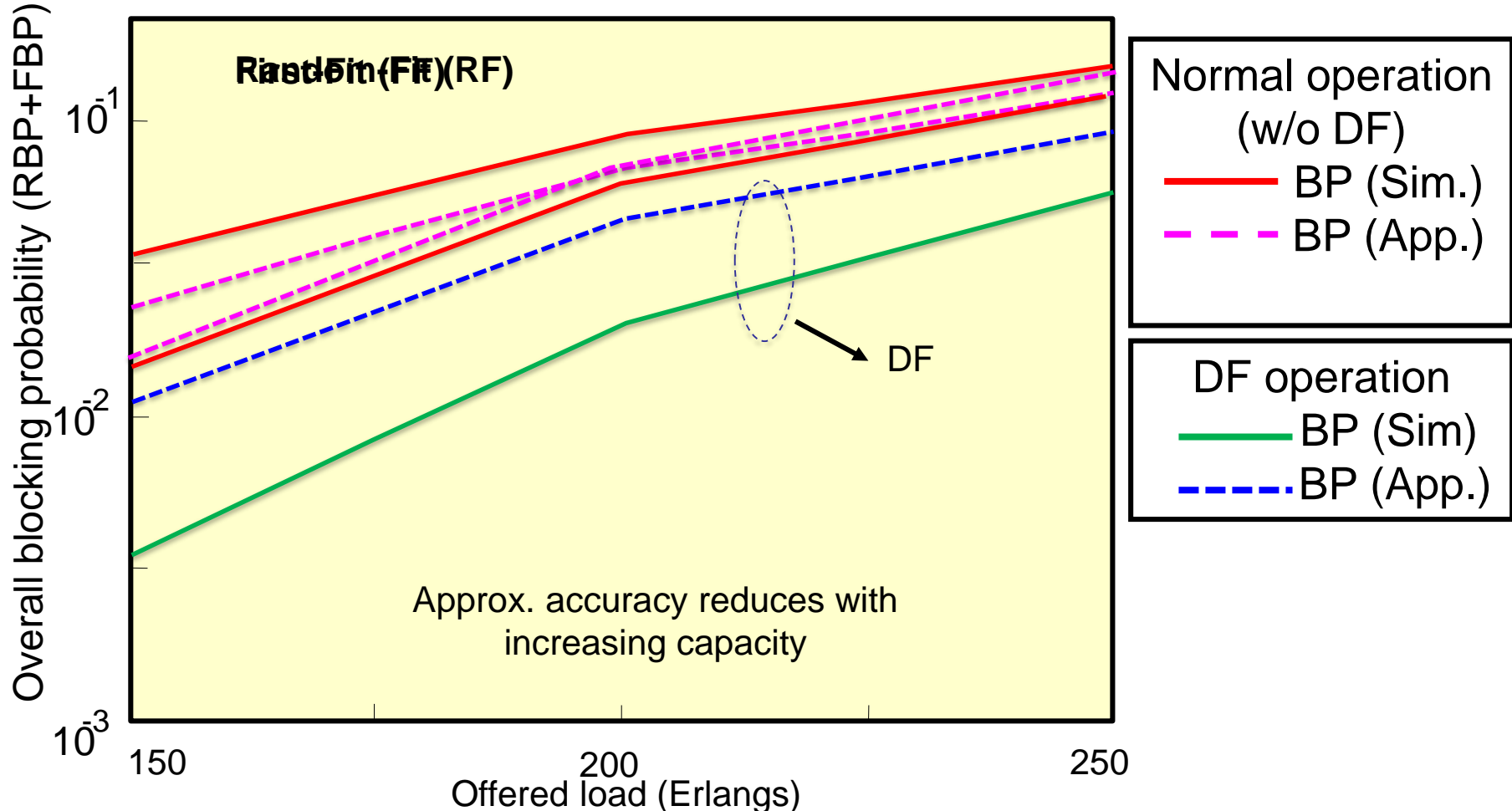
Link capacity  $C=7$  slices, demands  $(d) = (3,4)$  slices,  $\mu_k=1$ , mean reconfig time=0.001



# Defragmentation Impact in a Large Scale Network

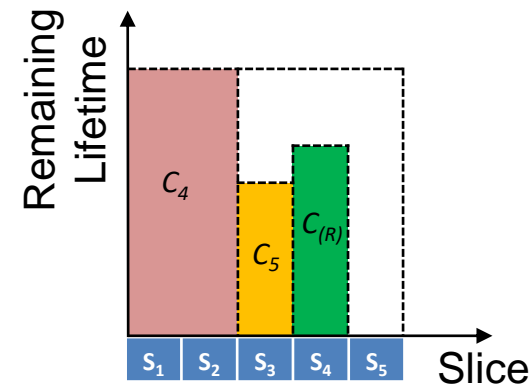
## 14-node NSF network with 182 OD pairs (Simulation and Approx.)

Link capacity  $C=100$  slices, demands  $(d) = (3,4,6)$  slices,  $\mu_k=1$ , mean reconfig time=0



# Algorithmic Design of Efficient Resource Allocation

- ❑ Fragmentation a multidimensional problem
  - ❖ spectral, time and spatial
- ❑ Holding-Time-Aware Resource Allocation (HTA-RA) scheme\*
  - ❖ Use residual lifetime of connections

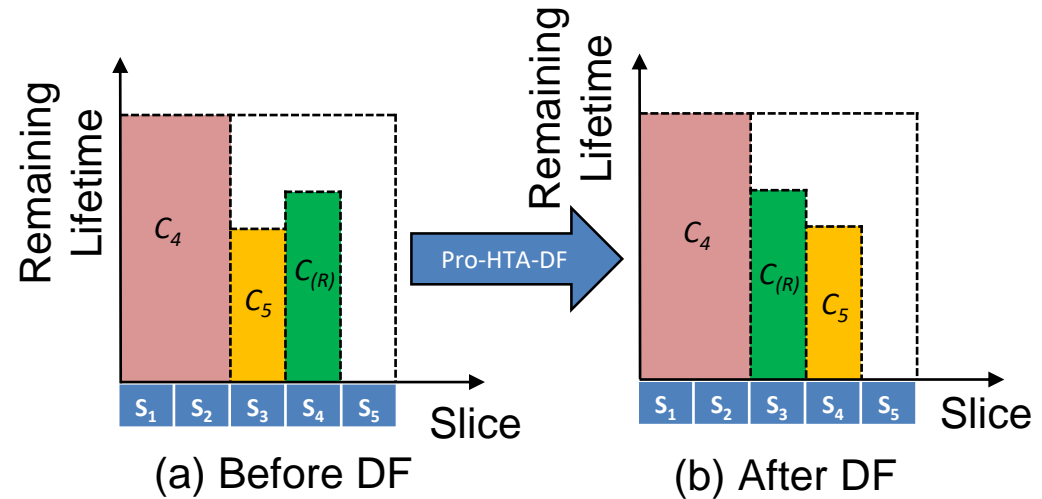


- ❑ In dynamic environment, Holding-Time-Aware Defragmentation (HTA-DF) is essential

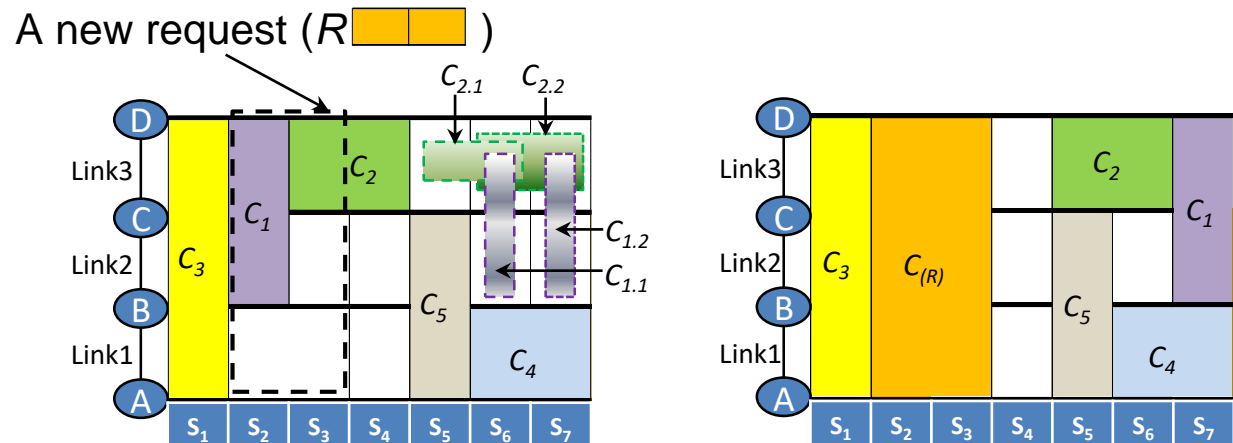


# Algorithmic Design of Efficient Defragmentation

- Proactive approach  
(Pro-HTA-DF)



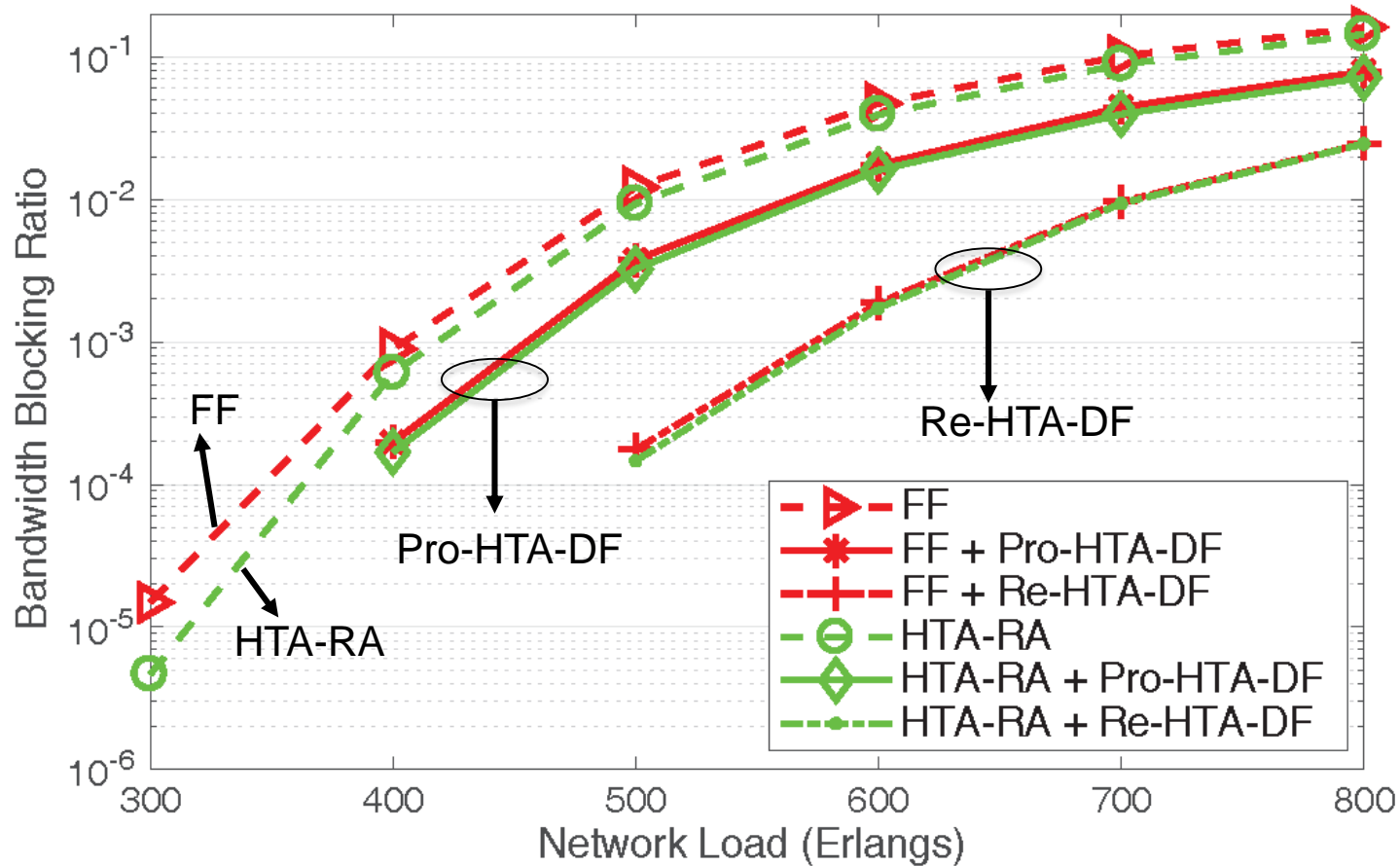
- Reactive approach  
(Re-HTA-DF)



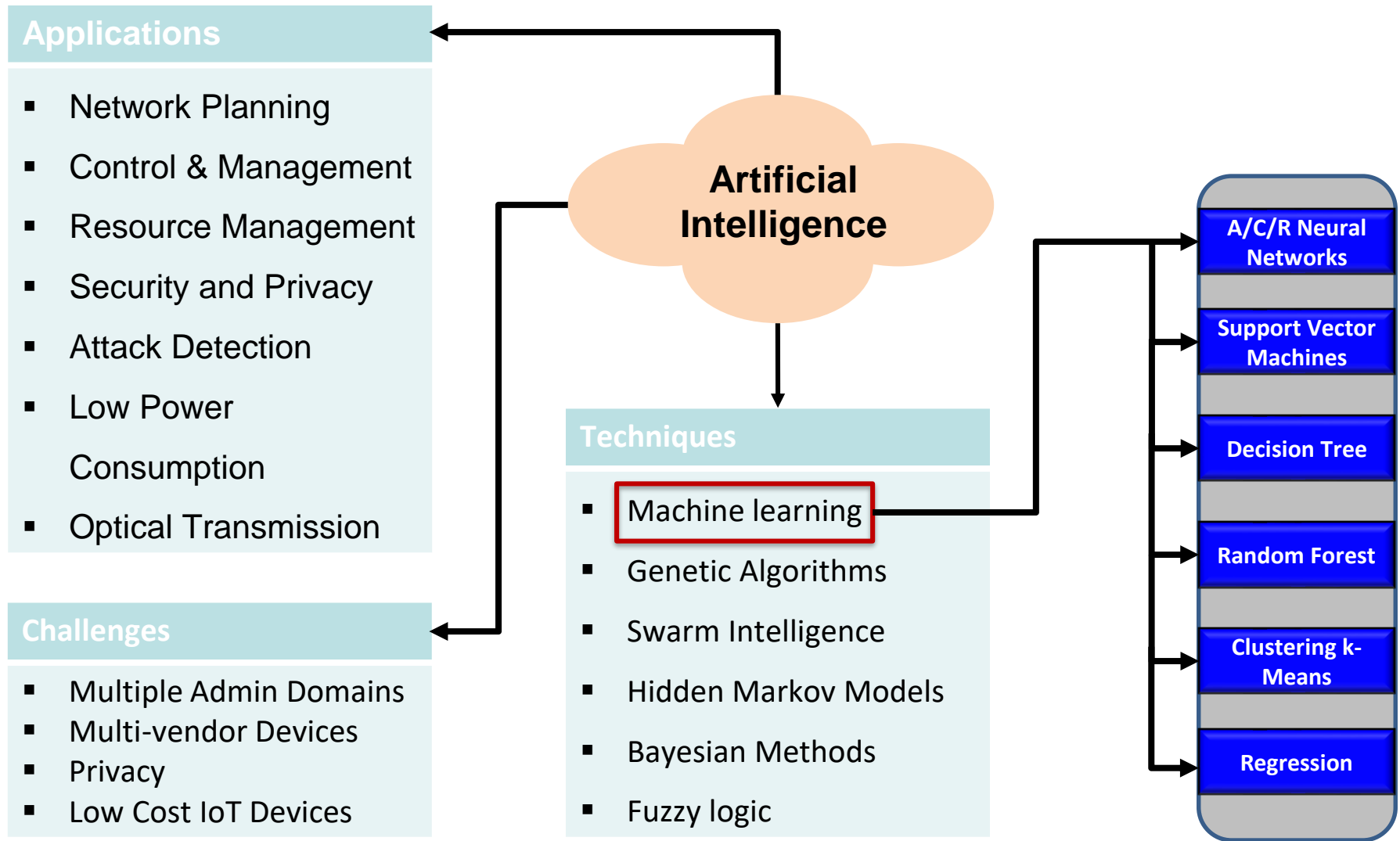
# Resource Allocation with Proactive/Reactive DF?

## 14-node NSF network with 21 bidirectional links

Link capacity  $C=200$  slices, demands  $(d) = (10,40,100)$  Gbit/s, DP-QPSK



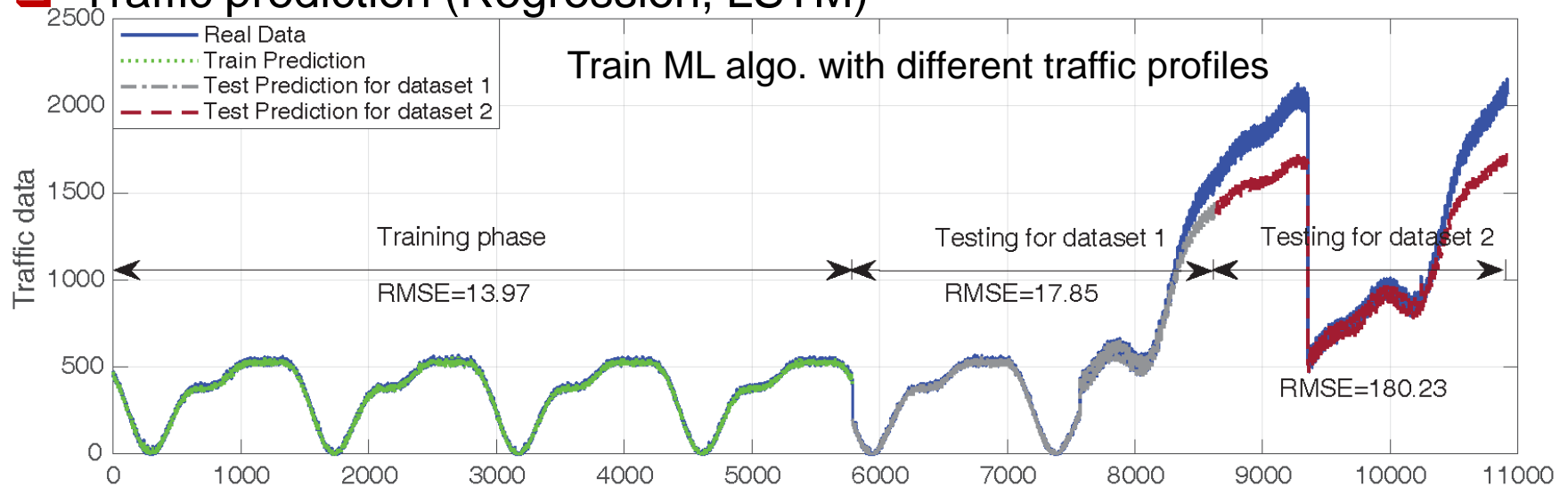
# Resource Management & Novel System Design with AI\*



# Handling Time-Varying Traffic in ODCN with ML

- ❑ DC traffic are diverse, bursty (ON/OFF), service time heavy-tailed distr.
- ❑ ML-assisted efficient resource (re)allocation algorithms\*
  - ❖ Route a new request along a lightpath whose residual life is comparable to the request's estimated mean service time.

## ❑ Traffic prediction (Regression, LSTM)



Time-varying traffic

$$m(t) = \alpha + \sum_{k=1}^n \beta_k \times \sin(\omega_k t + \phi_k)$$

# Performance of ML-Assisted Resource Allocation

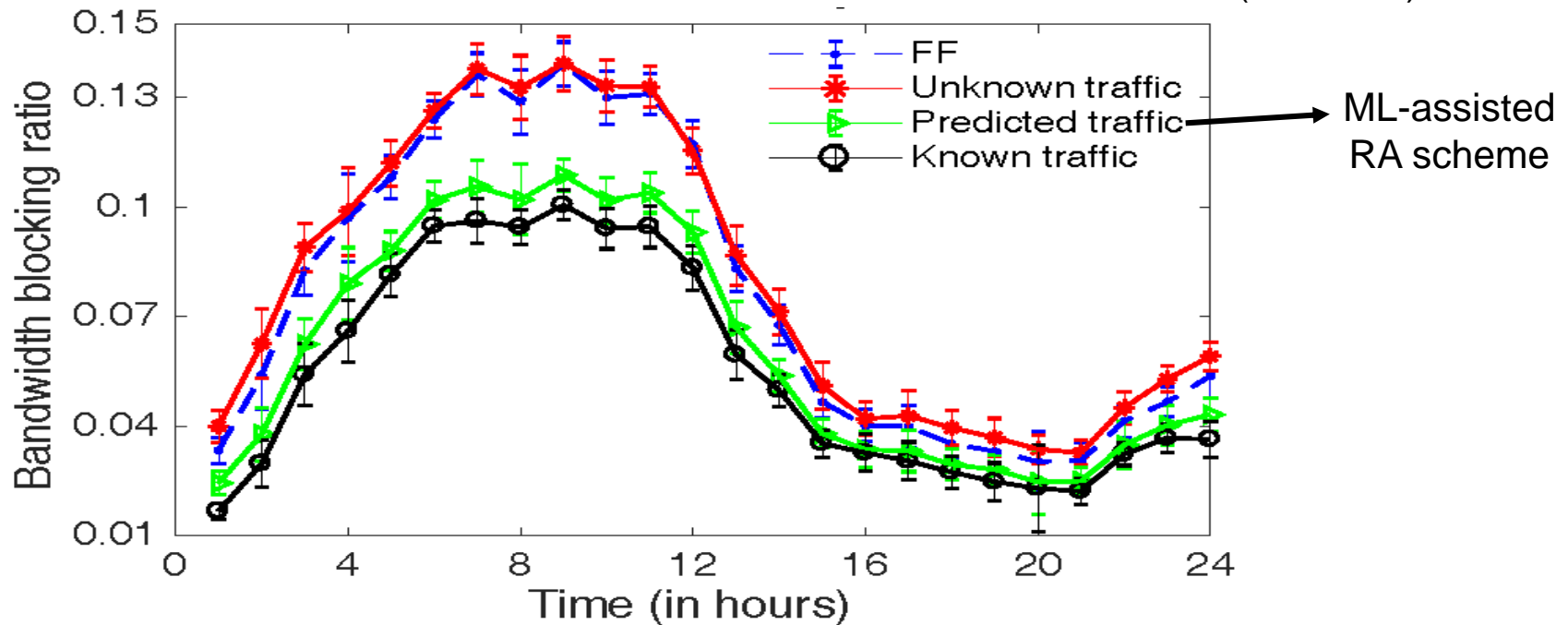
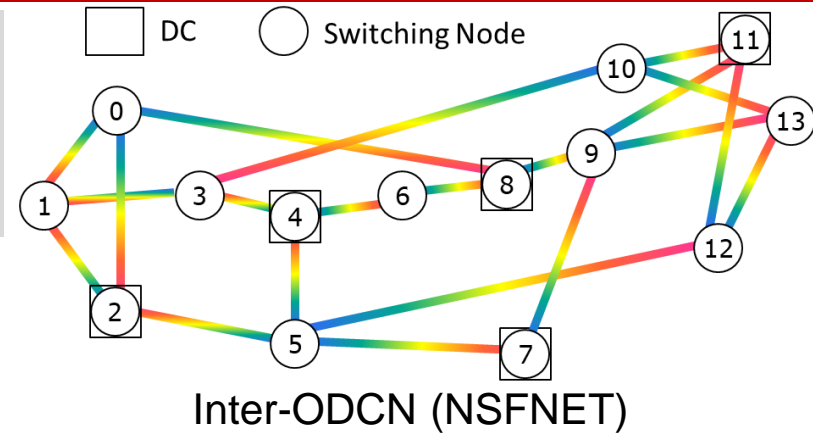
Traffic Assumptions:

Demand: {10, 40, 100} Gbit/s

Time-varying **exponential** arrival rates:  $\lambda_k$

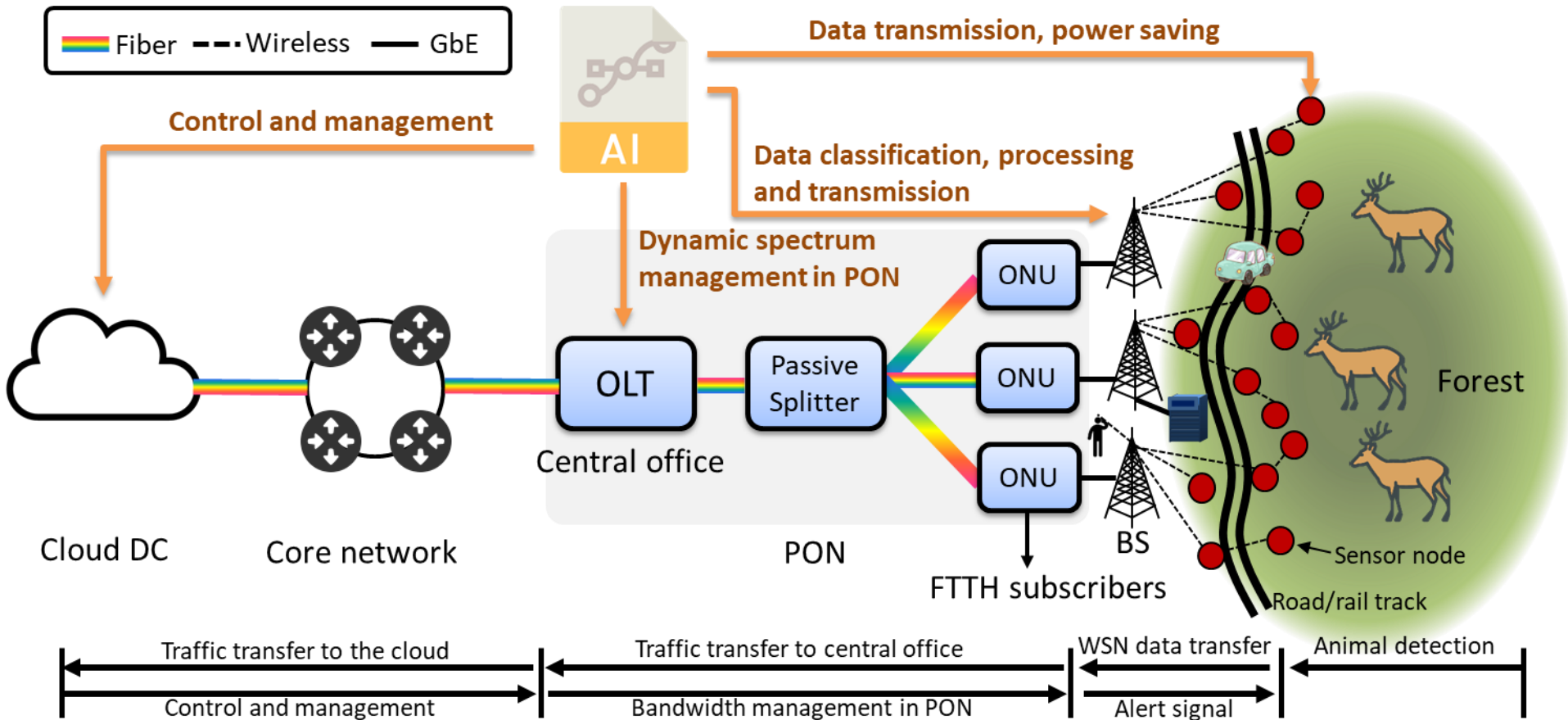
Time-varying **Lognormal** holding (service) times:  $\mu_k$

Fiber link with 3 cores  
Core capacity C=200 slices



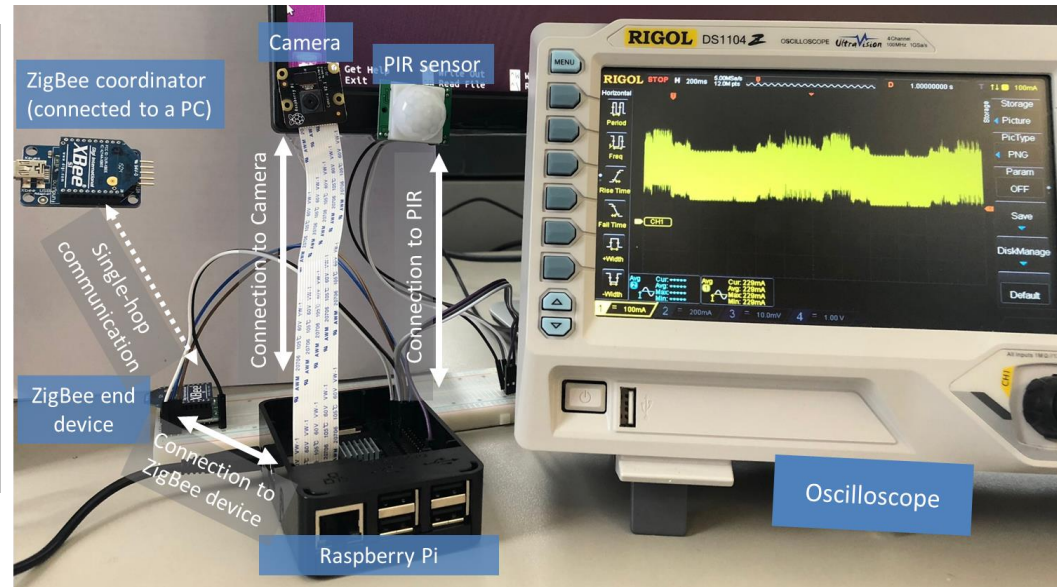
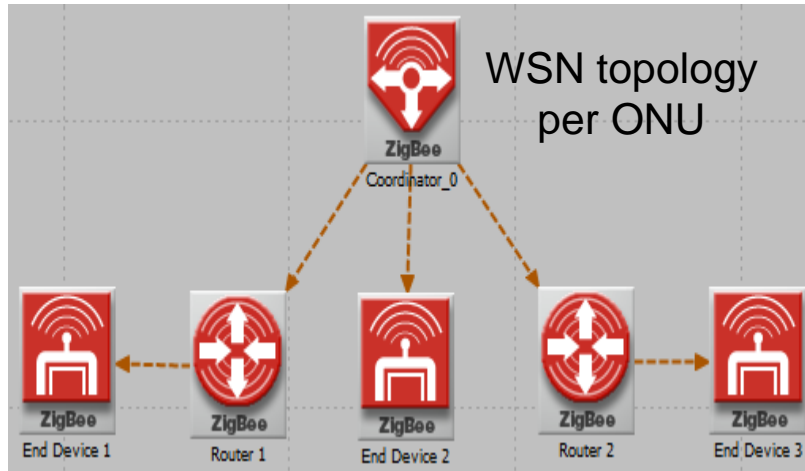
# ML-Assisted Early Warning System\*

- Every year there are hundreds of thousands of Animal-vehicle collision worldwide

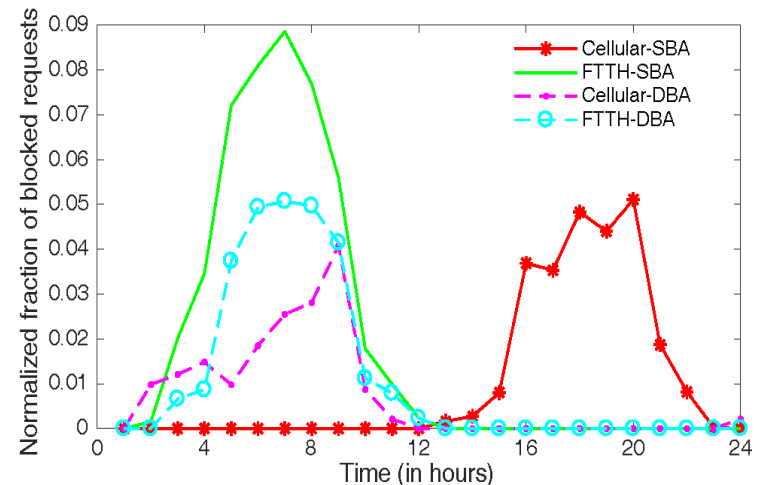


\* S. K. Singh, F. Carpio, and A. Jukan, "Improving Animal-Human Cohabitation with Machine Learning in Fiber-Wireless Networks," *MDPI Journal of Sensor and Actuator Networks*, vol. 7, no. 3, 2018.

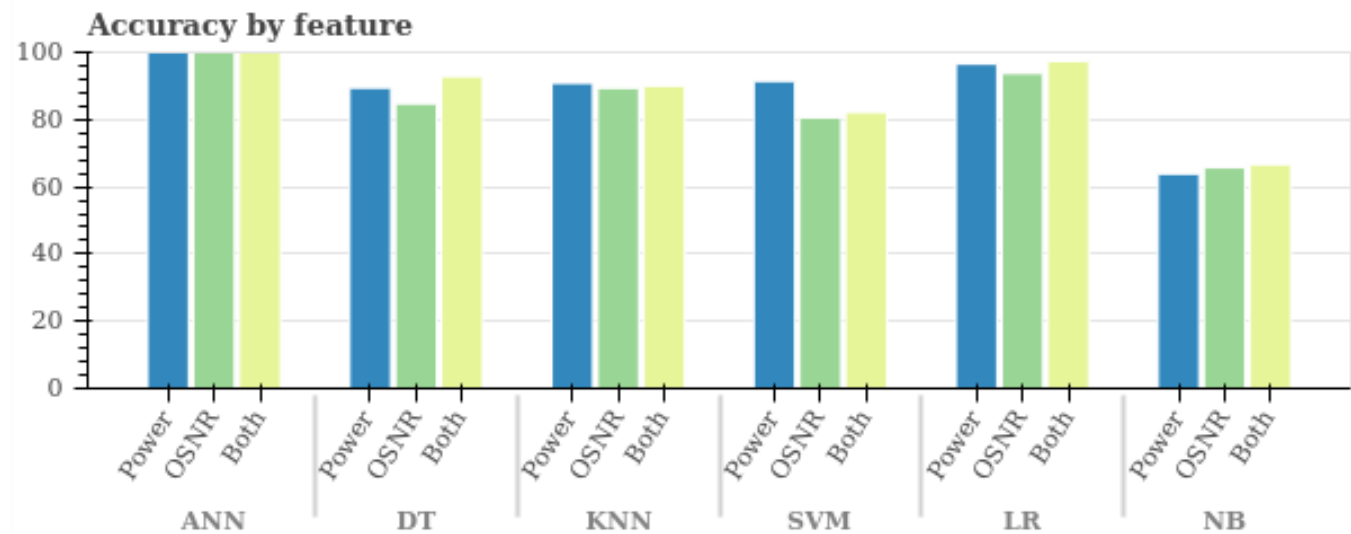
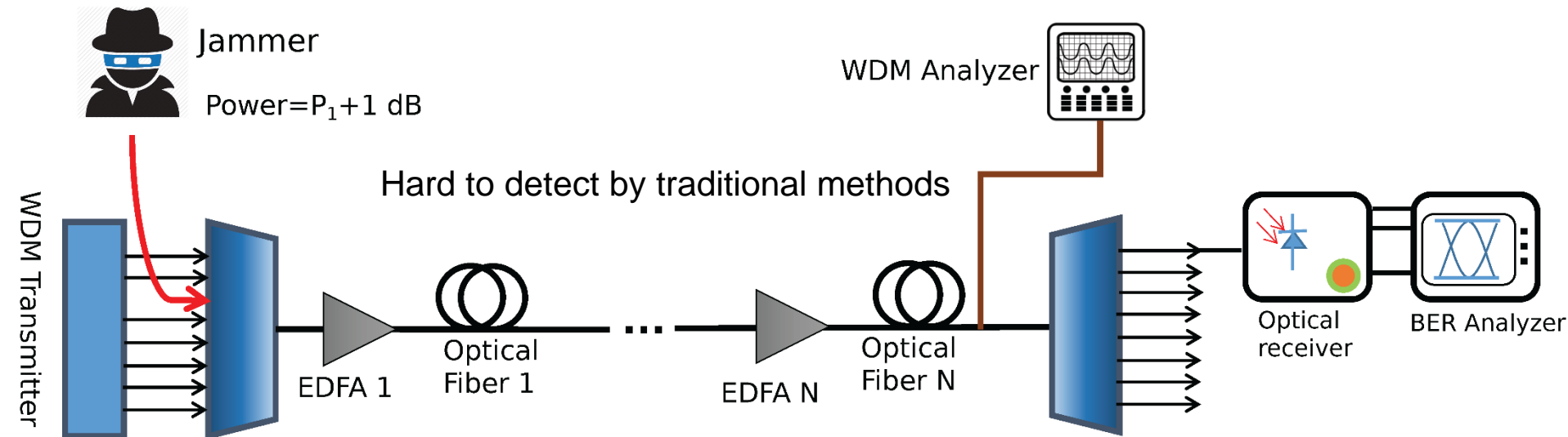
# Simulation and Experimental Setup and Results



- ❑ Processing @edge=> reduces alert notification time, and energy
- ❑ Dynamic bandwidth management in PON for time-varying traffic classes (Cellular, FTTH, others) by traffic prediction



# ML-Assisted Power Jamming Attack Detection and Prevention





# Conclusions

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## ❑ Elastic optical networking is promising

- ❖ Support for heterogeneous demands
- ❖ Able to provision beyond 100 Gbit/s channel

## ❑ Requirements and challenges

- ❖ Flexible bandwidth-variable transponders, and switches
- ❖ Resource provisioning is complex
- ❖ Spectrum fragmentation is inevitable, defragmentation is required

## ❑ Thesis Contribution

- ❖ Analytical modeling of routing and spectrum allocation schemes with/without DF
- ❖ Efficient algorithms for resource provisioning (Holding-Time-Aware-based)
- ❖ ML-assisted intelligent resource provisioning and novel system design

# References

1. A. Birman, "Computing approximate blocking probabilities for a class of all-optical networks," IEEE Journal on Selected areas in Communications, vol. 14, no. 5, pp. 852–857, 1996.
2. K. Kuppuswamy and D. C. Lee, "An analytic approach to efficiently computing call blocking probabilities for multiclass wdm networks," IEEE/ACM Transactions on Networking (TON), vol. 17, no. 2, pp. 658–670, 2009.
3. H. Beyranvand, M. Maier, and J. Salehi, "An analytical framework for the performance evaluation of node-and network-wise operation scenarios in elastic optical networks," IEEE Trans. on Commun., vol. 62, no. 5, pp. 1621–1633, 2014.
4. L. Peng, C.-H. Youn, and C. Qiao, "Theoretical analyses of lightpath blocking performance in co-ofdm optical networks with/without spectrum conversion," IEEE Communications Letters, vol. 17, no. 4, pp. 789–792, 2013.
5. S. K. Singh and A. Jukan, "Computing Blocking Probabilities in Elastic Optical Networks with Spectrum Defragmentation," in proc. of IEEE Conference on Computer communications (**INFOCOM**), Paris, April 2019.
6. S. K. Singh and A. Jukan, "Machine Learning-based Prediction for Resource (Re)allocation in Optical Data Center Networks," IEEE/OSA Journal of Optical Communications and Networking (JOCN), vol. 10, issue 10, pp. D12–D28, 2018.
7. S. K. Singh and A. Jukan, "Efficient Spectrum Defragmentation with Holding Time Awareness in EONs," IEEE/OSA JOCN, vol. 9, no. 3, pp. B78–B89, 2017.
8. S. K. Singh, F. Carpio, and A. Jukan, "Improving Animal-Human Cohabitation with Machine Learning in Fiber-Wireless Networks," *MDPI Journal of Sensor and Actuator Networks*, vol. 7, no. 3, 2018.
9. M. Bensalem, S. K. Singh, and A. Jukan, "Machine Learning Techniques to Detecting and Preventing Jamming Attacks in Optical Networks," under submission in IEEE Globecom, 2019.

**Questions?**

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**Thanks**

# Approximation Models: Our Approach

- ❑ Reduced load approximation; statistical independence link assumption
- ❑ For each link  $j$ , find connection setup rate per class per state

$$\alpha_k^j(x_j) = \sum_{m=1}^{\kappa_o} \sum_{o: j \in r(o,m)} \lambda_k^{o,m} \Pr(Z_{r(o,m)} \geq d_k | X_j = x_j; \bar{x}_j)$$

$$\Pr(Z_r \geq d_k | X_j = x_j; \bar{x}_j) = \sum_{x_{j_2}=0}^{C-d_k} \cdots \sum_{x_{j_h}=0}^{C-d_k} \pi_{j_2}(x_{j_2}) \cdots \pi_{j_h}(x_{j_h}) \times p_k(\mathbf{x}_r; \bar{\mathbf{x}}_r)$$

Probability of Acceptance  $p_k(\mathbf{x}_r; \bar{\mathbf{x}}_r) \equiv \Pr[Z_r \geq d_k | X_{j_1} = x_{j_1}, X_{j_2} = x_{j_2}, \dots, X_{j_h} = x_{j_h}; \bar{\mathbf{x}}_r]$

- ❑ For each link  $j$ , find connection departure rates per class
- ❑ Apply fixed-point iteration algorithm to compute **fragmentation blocking probability (FBP)** and **resource blocking probability (RBP)**

$BP_k^o =$

Poisson assumption: Under this assumption, calls arrive at a link as a Poisson process and the corresponding arrival rate is the original external offered rate thinned by blocking on other links, thus known as the reduced load.

# Approximation Models: Existing Approaches

□ Link state (X) is represented by # of occupied slices (x);  $\pi(x) \equiv \Pr[X = x]$

H. Beyranvand, M. Maier, and J. Salehi, "An analytical framework for the performance evaluation of node-and network-wise operation scenarios in elastic optical networks," IEEE Transaction on Communications, vol. 62, no. 5, pp. 1621–1633, 2014.

□ **Kaufman's Approach** (Ignoring fragmentation and RSA constraints)

$$p_k(x) = 1, 0 \leq x \leq C - d_k \\ 0, \text{ otherwise}$$

$$x * \pi(x) = \sum_{k=1}^K d_k \left( \frac{\lambda_k}{\mu_k} \right) \pi(x - d_k)$$

$$\sum_{x=0}^C \pi(x) = 1$$

$$\text{Link } BP_k^{\text{Kaufman}} = \sum_{x=C-d_k+1}^C \pi(x)$$

$$\text{OD Route } BP = 1 - (1 - BP_{\text{link}})^h$$

□ **Binomial Approach** based on [\*]

$p$ : probability that a slice is free on a link;

$$p = 1 - \frac{1}{C} \sum_{i=0}^C x \pi(x)$$

$$f(C, d_k) \\ = \sum_{i=1}^{d_k} (f(C - i, d_k) (1 - p) p^{i-1}) + p^{d_k}$$

Replace  $p$  with  $p^h$  for h-hop route

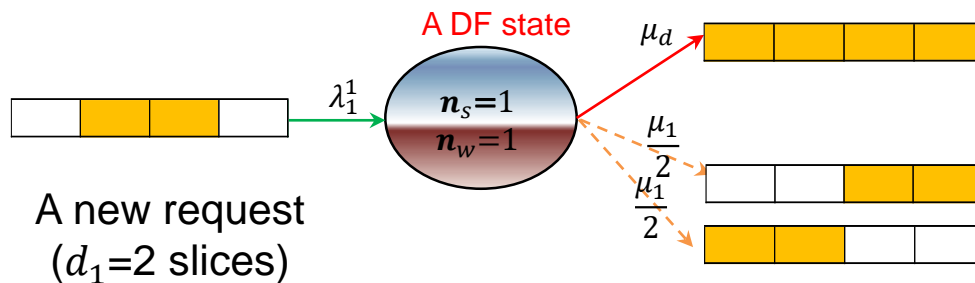
$$\text{Link/Route } BP_k^{\text{Binomial}} = 1 - f(C, d_k)$$

# Exact and Approx. Models of Reactive Defragmentation\*

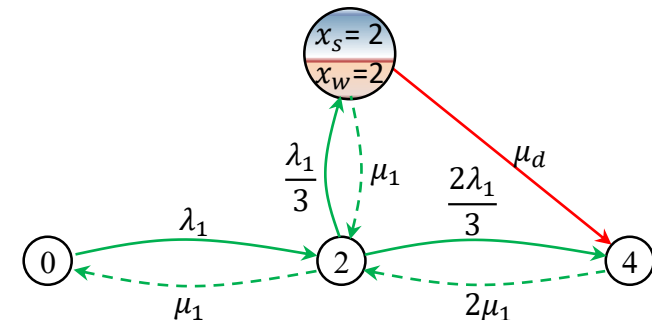
## Assumptions:

- ❖ Put on hold (delayed) to triggering request, and serve it after DF
- ❖ Reconfiguration time  $\sim \exp(1/\mu_d)$
- ❖ DF also completes when a serving connection departs (since  $1/\mu_k \gg 1/\mu_d$ )

## Example: Single-link, $C=4$ slices, $d=2$ slices, RF policy



Exact model



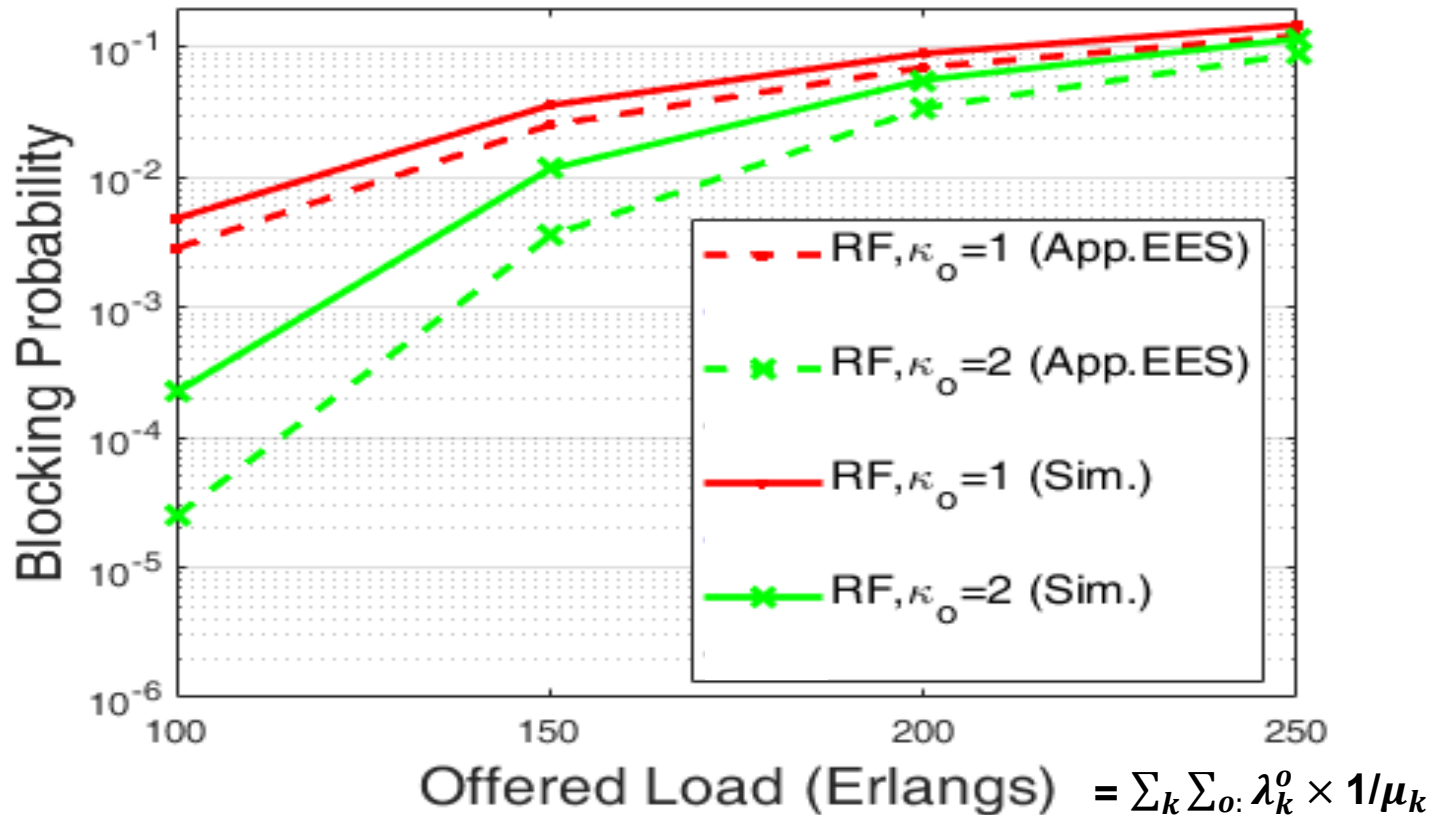
Approximation model (small scale)

# Accuracy Analysis under RF policy

## 14-node NSF network with 182 OD pairs, RF policy

Link capacity  $C=100$  slices, demands  $(d) = (3,4,6)$  slices,

Mean service time  $(1/\mu_k) = 1$  unit

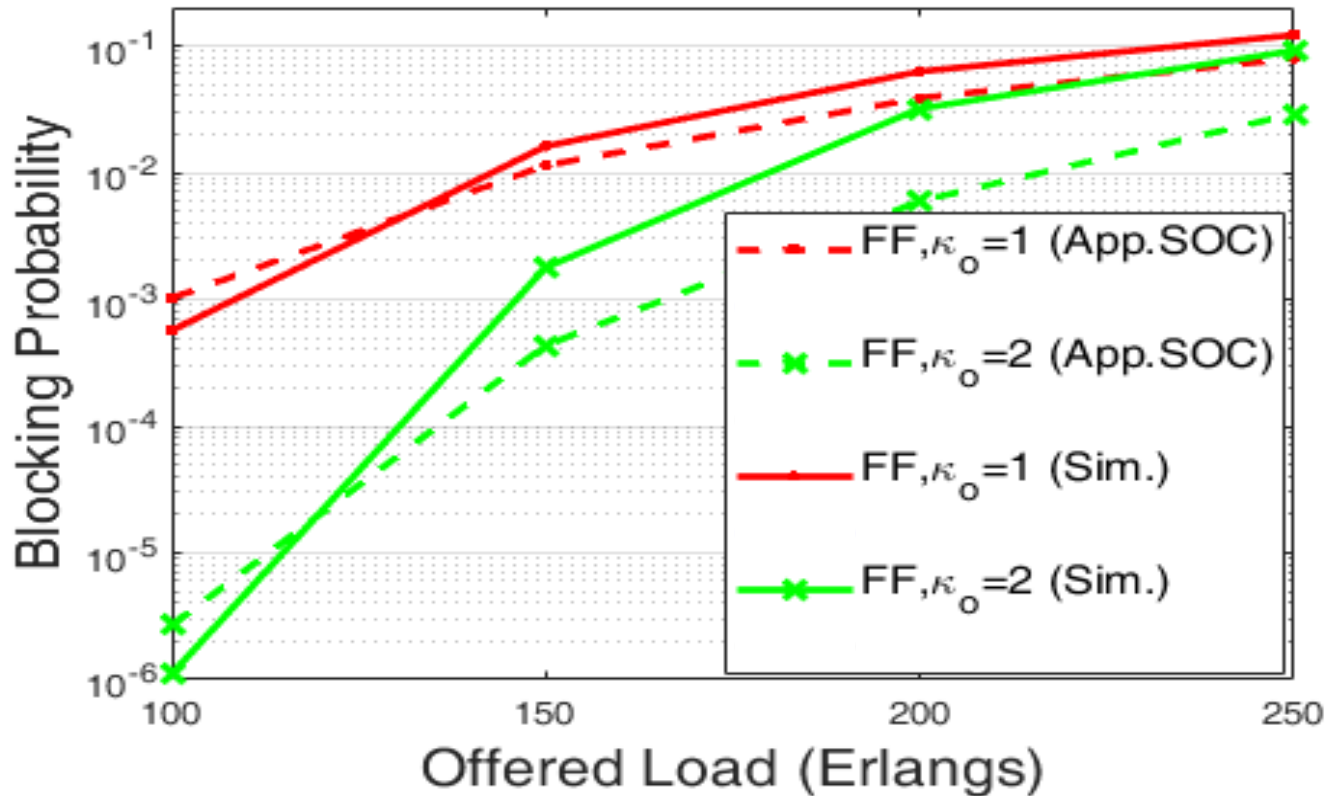


# Accuracy Analysis under FF policy

## 14-node NSF network with 182 OD pairs, FF policy

Link capacity  $C=100$  slices, demands  $(d) = (3,4,6)$  slices,

Mean service time  $(1/\mu_k) = 1$  unit





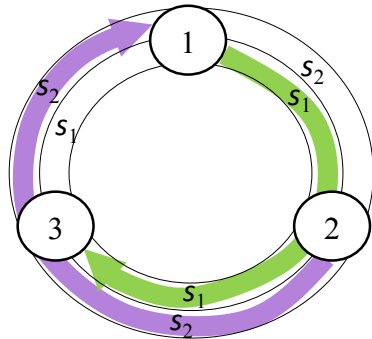
# Exact and Approx. Defragmentation Modeling

## □ The First Exact Defragmentation (DF) Network (Markov) Model

- ❖ Include DF states: # of serving and waiting connections (o,k)
- ❖ Assumptions:  $DF \sim \exp(\mu_d)$  and DF also completes when a serving connection departs (since  $1/\mu_k \gg 1/\mu_d$ )

## □ Approximation Model:

- ❖ All fragmented states can't be defragmented

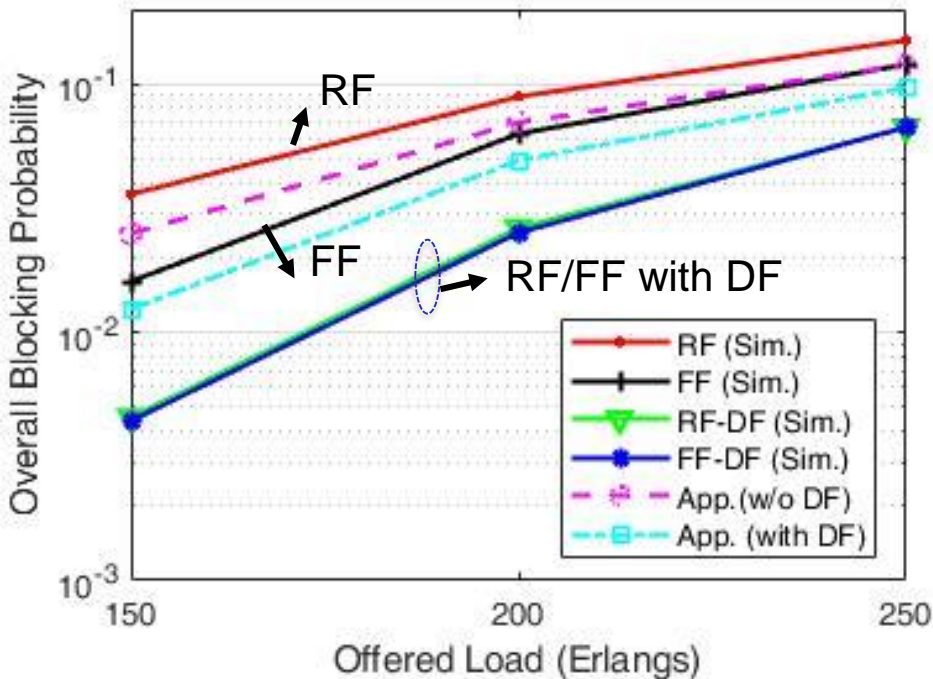


$$p_{k,DF}^{App.}(\mathbf{x}_r; \bar{\mathbf{x}}_r) = \prod_{j_i \in r} p_k^{App.}(x_{j_i}; \bar{x}_{j_i}) + \prod_{j_i \in r: |FB(x_{j_i}, k)| \neq 0} g_k^{App.}(x_{j_i}; \bar{x}_{j_i})$$

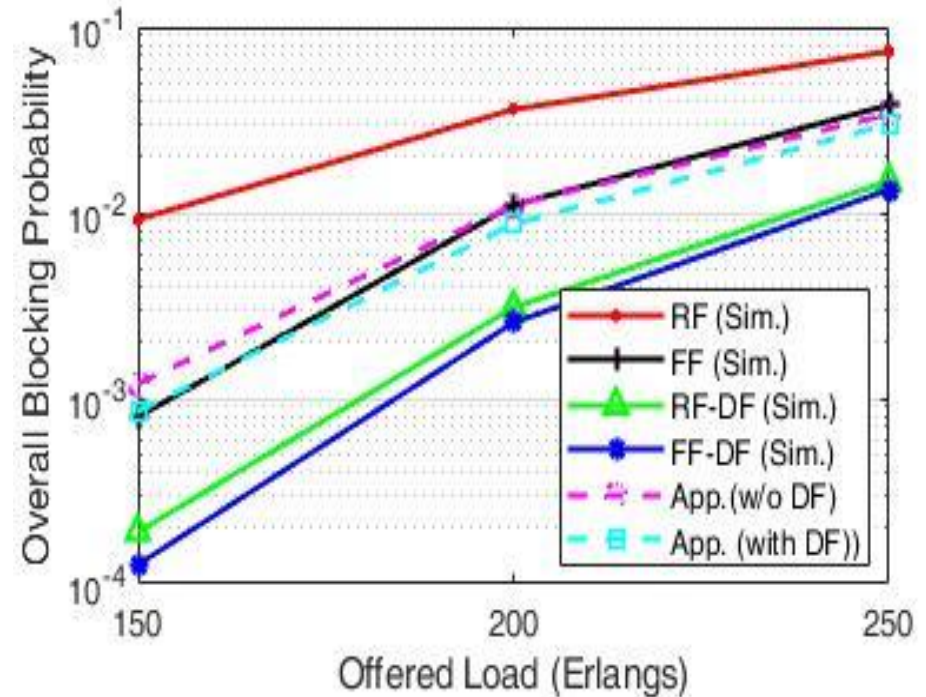
# Defragmentation Impact in a Large Scale Network

## 14-node NSF network with 182 OD pairs (Simulation and Approx.)

$C=100$  slices,  $d=(3,4,6)$ , reconfig. time=0



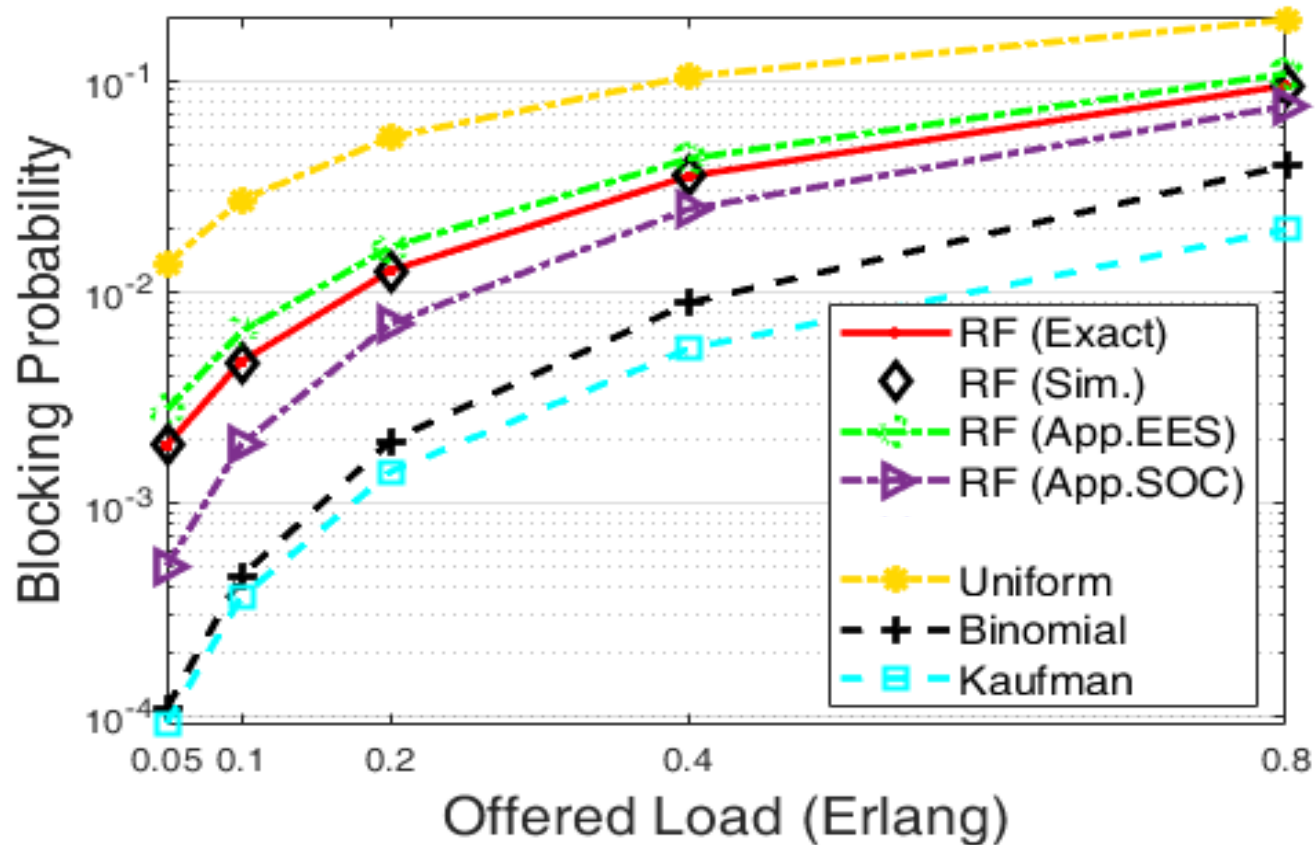
$C=200$  slices,  $d=(4,6,10)$ , reconfig. time=0



# Accuracy of Exact and App. Approaches under RF policy

- A 2-Hop network with 3 OD pair routes (RF policy)

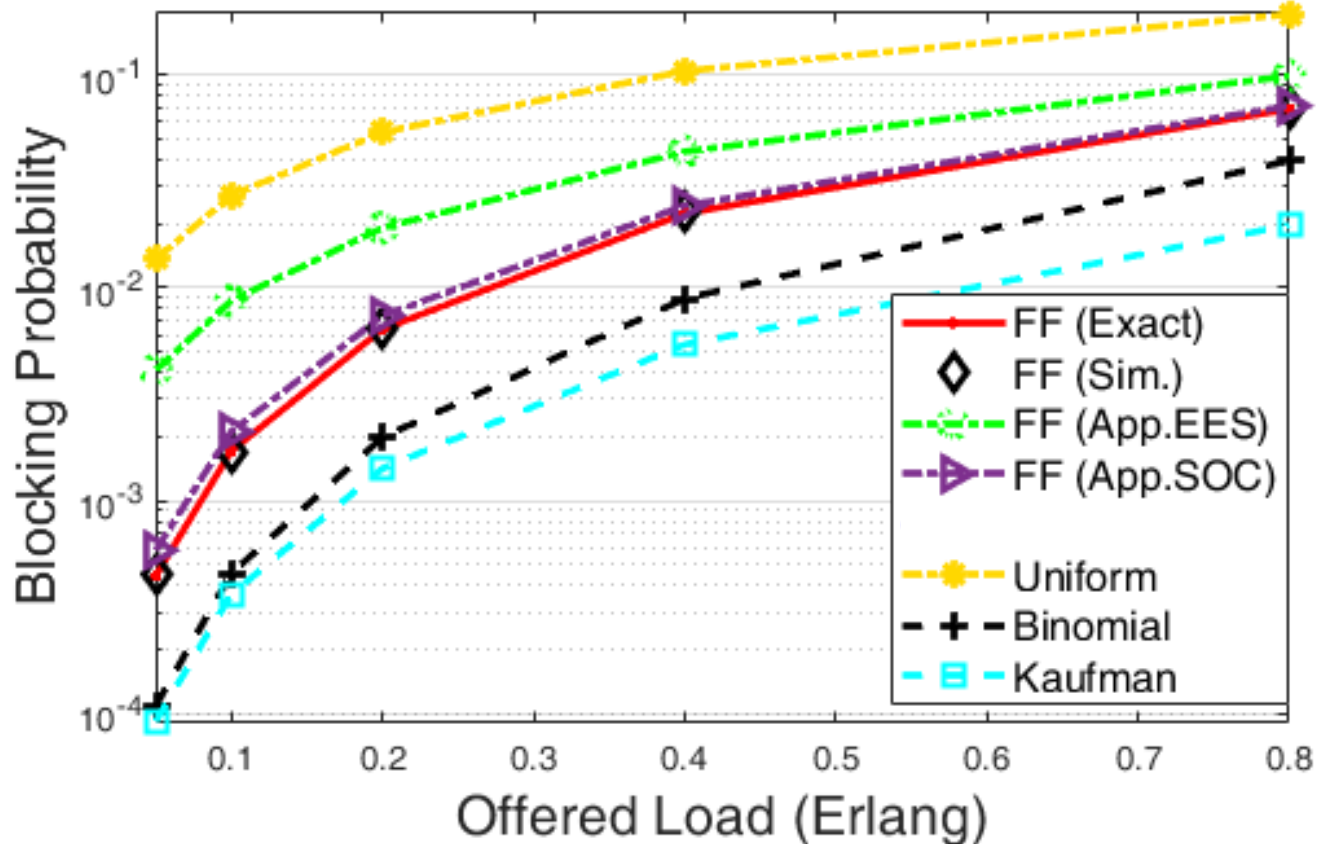
Link capacity  $C=10$  slices, demands  $(d) = (3,4)$  slices



# Accuracy of Exact and App. Approaches under FF policy

- A 2-Hop network with 3 OD pair routes (RF policy)

Link capacity  $C=10$  slices, demands  $(d) = (3,4)$  slices



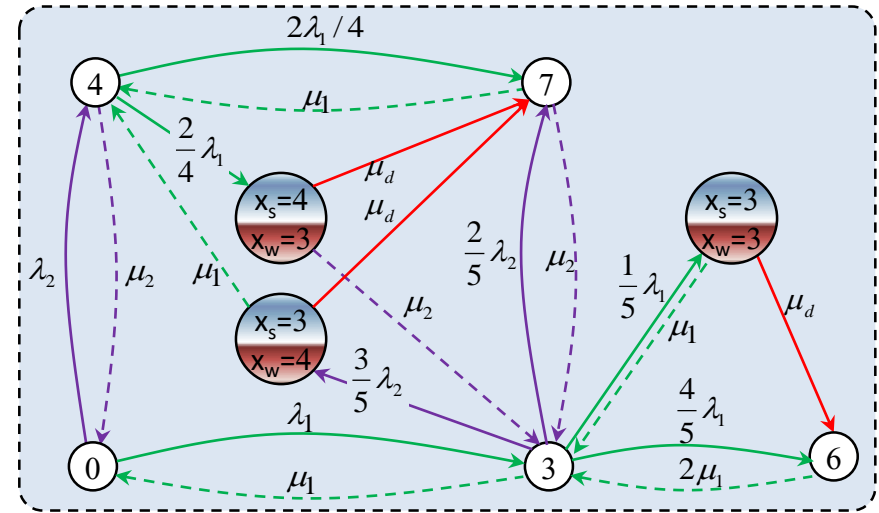
# Defragmentation Modeling: Approximation Approach

## □ DF Reduced state (Approximation) link model

$$g_k^{App.EES}(x) = \frac{|\mathbb{FI}(x, k)|}{|\Omega_S(x)|}$$

$$g_k^{App.SOC}(x; \bar{x}) = \frac{|\mathbb{FI}(x, k)|}{|\Omega_S(x)|} \times \left( 1 - \exp \left( -\frac{\bar{x}}{C} \times |\log(\frac{x}{\bar{x}})| \right) \right)$$

$$p_{k,DF}^{App.}(\mathbf{x}_r; \bar{\mathbf{x}}_r) = \prod_{j_i \in r} p_k^{App.}(x_{j_i}; \bar{x}_{j_i}) + \prod_{j_i \in r: |FB(x_{j_i}, k)| \neq 0} g_k^{App.}(x_{j_i}; \bar{x}_{j_i})$$

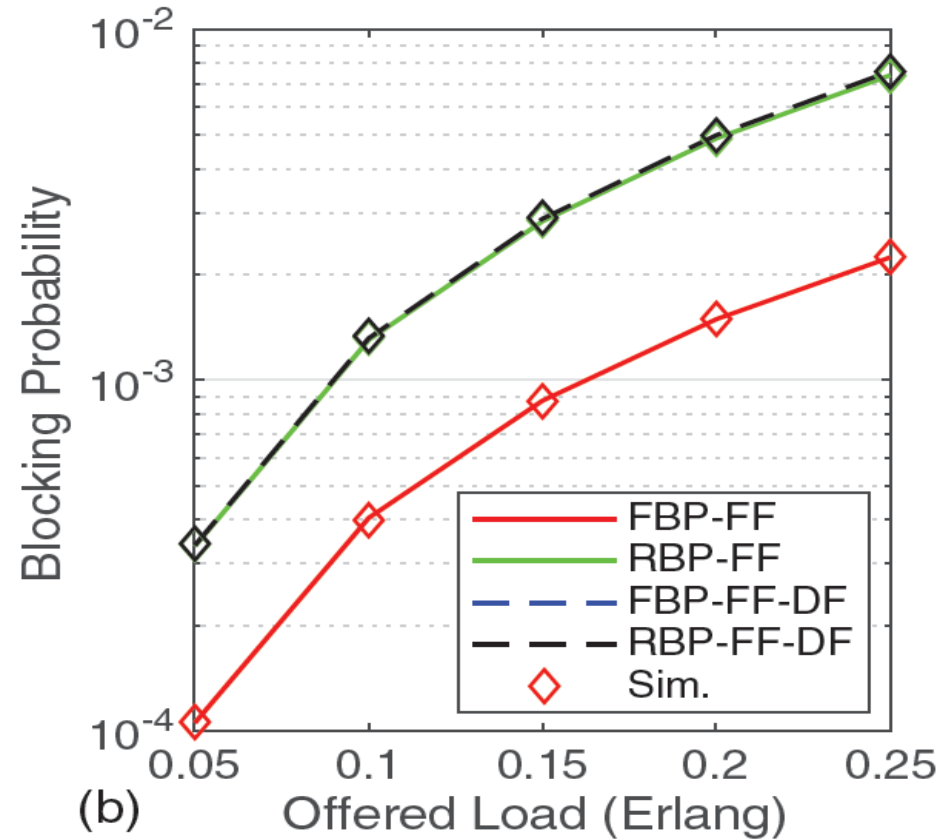
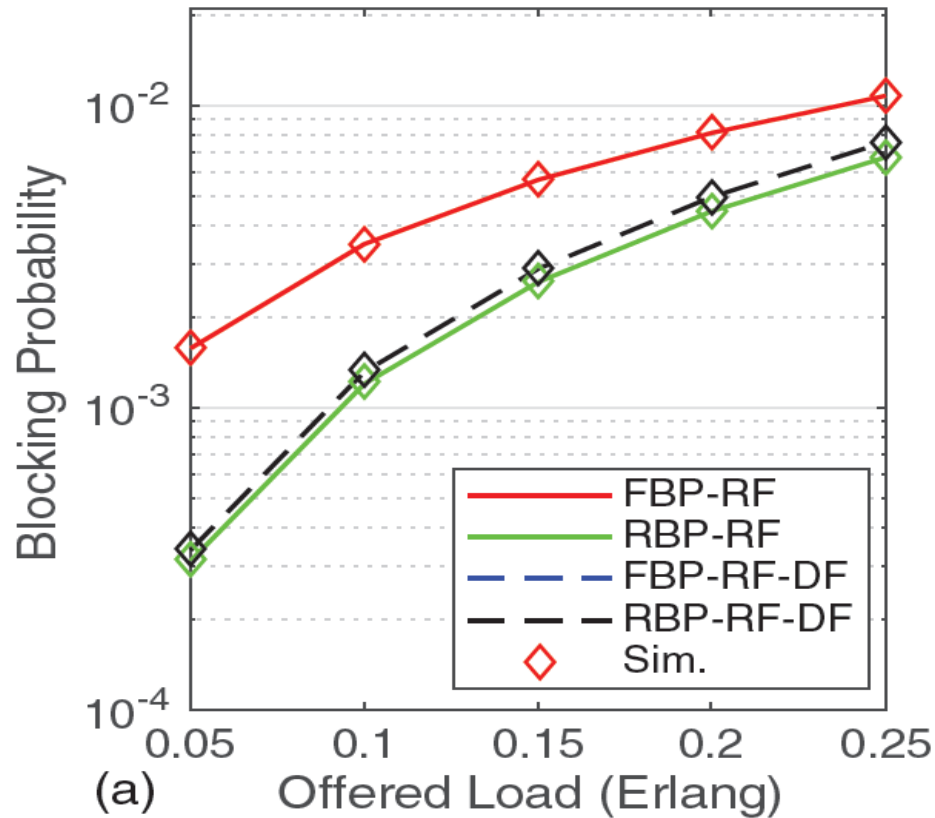


# Impact of DF in Networks with Bus Topology

## 2-link network with 3 OD pairs (Exact and Simulation results)

Link capacity  $C=10$  slices, demands  $(d) = (3,4)$  slices

Zero Fragmentation blocking under DF



# Impact of Defragmentation in Ring/Mesh Networks

## 3-node ring network with 6 OD pairs (Exact and Simulation results)

Link capacity  $C=7$  slices, demands  $(d) = (3,4)$  slices

