# Contents on the Move Content Caching and Delivery at the Wireless Network Edge

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## Information Processing and Communications Lab (IPC-LAB)



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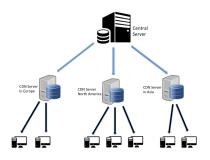
- Video demand dominates traffic (78% by 2021)
- 75% of Facebook video browsing, 40% of Netflix downloads performed on smartphones
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- Asymmetric resource usage
- Delay-tolerant, asynchronous access
- Most traffic due to a few viral/ popular video files
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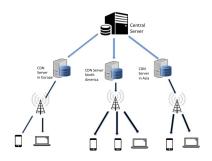
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## Content Distribution Networks



- Content provider (e.g. Netflix, BBC, Facebook) contracts with a CDN (e.g. Akamai, LimeLight)
- Balance traffic, reduce latency, ...
- This is in the core network

#### Content Distribution Networks - Wireless



- Content provider (e.g. Netflix, BBC, Facebook) contracts with a CDN (e.g. Akamai, LimeLight)
- Balance traffic, reduce latency, ...
- This is in the core network
- Bring content to the edge (e.g., Netflix Open Connect)

# **Coded Proactive Content Caching**



- Two-phase protocol:
  - Placement phase: off-peak hours, user demands unknown
  - Delivery phase: peak hours, demands revealed
- Library of N files, each consisting of F bits
- K users, each equipped with a cache of size M
- Each user requests one file
- Error-free shared delivery link: Satisfy all demands simultaneously
- What is the minimum number of bits that must be delivered sufficient to satisfy all demand combinations?
- What is the trade-off between cache capacity and number of delivered bits?

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# **Coded Proactive Content Caching**



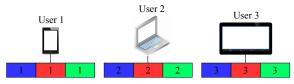
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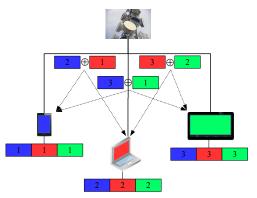
- N = 3 files
- K = 3 users
- Cache capacity: M = 1
- Split each file into 3 non-overlapping equal-size subfiles:



• Cache contents after placement phase:



• Delivery phase:

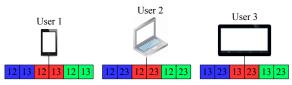


• Delivery rate:  $R_{\text{MAN}}(1) = 1$ 

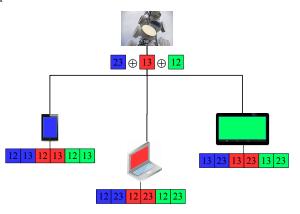
- N = 3 files
- K = 3 users
- Cache capacity: M = 2
- Split each file into 3 non-overlapping equal-size subfiles:

$W_1$	12	13	23
$W_2$	12	13	23
$W_3$	12	13	23

• Cache contents after placement phase:

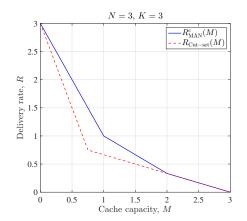


• Delivery phase:



•  $R_{\text{MAN}}(2) = 1/3$ 

# Delivery Rate-Cache Capacity Trade-off



Many improvements and variations since then...

M. Mohammadi Amiri and D. Gündüz, Fundamental limits of caching: Improved delivery rate-cache capacity trade-off, IEEE Trans. on Communications, vol. 65, no. 2, pp. 806-815, Feb. 2017.
 M. Mohammadi Amiri, Q. Yang and D. Gündüz, Decentralized coded caching with distinct cache

## Caching and Delivery for Heterogeneous Devices



- Devices have different resolution/processing capabilities
- They may request the same file, but at different resolutions
- $\bullet$   $D_k$ : distortion requirement of user k. Without loss of generality, let

$$D_1 \geq D_2 \geq \cdots \geq D_K$$

• Devices have distinct cache capacities:  $M_k$ 

Q. Yang and D. Gündüz, Coded caching and content delivery with heterogeneous distortion requirements, to appear, IEEE Trans. on Information Theory.

# Scalable Coded Caching

D <sub>1</sub> D <sub>2</sub> D <sub>3</sub> :	$r_1$	$r_{2}-r_{1}$	$r_{3} - r_{2}$
	$D_1$		
: D <sub>3</sub>		D <sub>2</sub>	
i i		$D_3$	
	:		

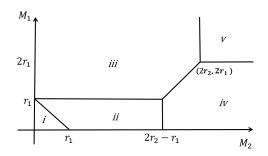
Compress video into multiple quality layers; e.g., scalable video coding (SVC) in H264/ MPEG

- First layer:  $r_1$  bits/sample
- k—th layer:  $r_k r_{k-1}$  bits/sample
- User k wants  $D_k \to \text{needs first } k$  layers

# Centralized Lossy Coded Caching (N = K = 2)

Given  $(r_1, r_2)$ , five cases depending on cache capacities  $M_1$  and  $M_2$ :

- Case i&ii: coded placement
- Case iii&iv: coded placement and coded delivery
- Case v: uncoded caching



- Proposed layered caching scheme is optimal.
- Requires coded caching and delivery simultaneously.

# Scalable Coded Caching and Delivery

#### Two subproblems:

- Cache allocation among different layers
- Lossless caching/delivery of each layer with heterogeneous cache sizes

#### Cache capacity allocation:

- Proportional Cache Allocation (PCA)
  - Allocate cache capacity proportionally (to sizes) among requested layers
- Ordered Cache Allocation (OCA)
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# Scalable Coded Caching and Delivery

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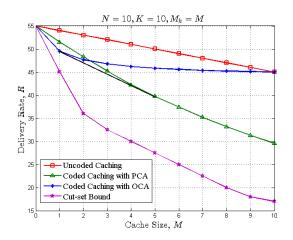
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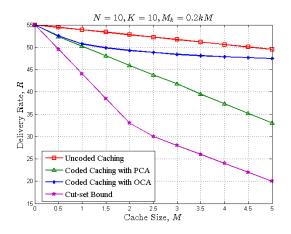
## **Identical Cache Capacities**

- $D_1 \ge D_2 \ge \cdots \ge D_{10}$ :  $r_k = k, k = 1, ..., 10$ ;
- Identical cache capacities,  $M_k = M$ .

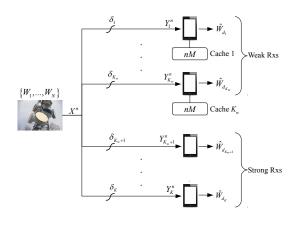


## Heterogeneous Cache Capacities

- $D_1 \ge D_2 \ge \cdots \ge D_{10}$ :  $r_k = k, k = 1, ..., 10$ ;
- Heterogeneous cache capacities,  $M_k = 0.2kM$ .



## Cache-Aided Wireless Content Delivery



$$\delta_k = egin{cases} \delta_w & ext{if } k \in [K_w] \ \delta_s & ext{if } k \in [K_w+1:K] \end{cases}$$

S. Saeedi Bidokhti, M. Wigger, and R. Timo, Noisy Broadcast Networks with Receiver Caching, submitted.

## Packet Erasure Broadcast Channel

- Library of N files:  $W_1, \ldots, W_N$
- Each file is distributed uniformly over  $\left[2^{nR}\right] \stackrel{\Delta}{=} \left\{1, \dots, 2^{nR}\right\}$
- Packet erasure broadcast channel

$$P(Y_k = y_k | X = x) = \begin{cases} 1 - \delta_k, & \text{if } y_k = x, \\ \delta_k, & \text{if } y_k = \Delta \end{cases}$$

- $\bullet \ P_e \stackrel{\triangle}{=} \max_{(d_1, \dots, d_K) \in [N]^K} \Pr \left\{ \bigcup_{k=1}^K \left\{ \hat{W}_{d_k} \neq W_{d_k} \right\} \right\}$
- (M, R) is achievable, if for every  $\varepsilon > 0$ ,  $\exists n$  large enough, s.t.  $P_e < \varepsilon$

$$C \stackrel{\Delta}{=} \sup \{R : (M, R) \text{ is achievable}\}$$

• Cache capacity of *M* only at weak receivers

## Main Result: Achievable Rate-Memory Pairs

Memory-rate pairs  $(M_{(p,q)}, R_{(p,q)})$  are achievable for any  $p \in [0:K_w]$  and  $q \in [p:K_w]$ :

$$\begin{split} R_{(p,q)} & \stackrel{\Delta}{=} \frac{F \sum\limits_{i=p}^{q} \left( \gamma \left( p,i \right) \right)}{\frac{1}{1-\delta_{w}} \sum\limits_{i=p}^{q} \left( \frac{K_{w}-i}{i+1} \gamma \left( p,i \right) \right) + \frac{K_{s}}{1-\delta_{s}}}, \\ M_{(p,q)} & \stackrel{\Delta}{=} \frac{N \sum\limits_{i=p}^{q} i \gamma \left( p,i \right)}{K_{w} \sum\limits_{i=p}^{q} \gamma \left( p,i \right)} R_{(p,q)}, \end{split}$$

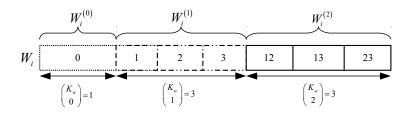
where

$$\gamma(p,i) \stackrel{\Delta}{=} \frac{\binom{K_w}{i}}{\binom{K_w}{p}K_s^{i-p}} \left(\frac{1-\delta_s}{1-\delta_w} - 1\right)^{i-p}, \text{ for } i = p,...,q.$$

M. Mohammadi Amiri and D. Gündüz, Cache-aided data delivery over erasure broadcast channels, *IEEE Transactions on Communications*, vol. 66, no. 1, pp. 370 - 381, Jan. 2018.

# Successive Joint Cache-Channel Coding (SCC) Scheme

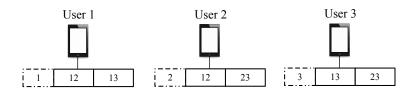
- $K_w = 3$  weak RXs
- $K_s = 2$  strong RXs
- p = 0, q = 2



- Rate of  $W_i^{(k)}$  is  $R^{(k)}$ , k = 0, 1, 2
- $R^{(0)} + R^{(1)} + R^{(2)} = R$

## Successive Joint Cache-Channel Coding (SCC) Scheme

#### Placement phase:

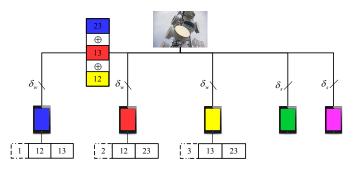


• Cache capacity:  $M = R^{(1)}/3 + 2R^{(2)}/3$ 

- q p + 2 = 4 distinct messages delivered by **time division** multiplexing
- Codewords of *i*-th message are of length  $\beta_i n$  channel uses, i = 1, ..., 4:

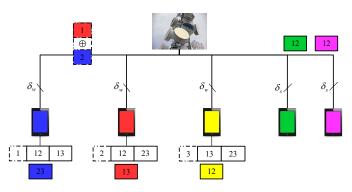
$$\sum_{i=1}^{4} \beta_i = 1$$

#### Message 1:



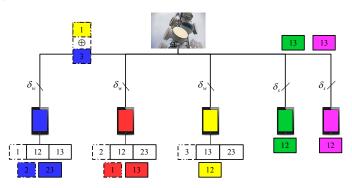
$$\frac{R^{(2)}/3}{(1-\delta_w)F} \le \beta_1$$

## Message 2, Part 1:



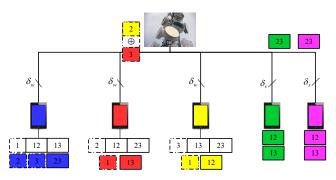
$$\max\left\{\frac{R^{(1)}/3}{(1-\delta_w)F}, \frac{R^{(1)}/3+2R^{(2)}/3}{(1-\delta_s)F}\right\} \le \beta_{2,1}$$

## Message 2, Part 2:



$$\max\left\{\frac{R^{(1)}/3}{(1-\delta_w)F}, \frac{R^{(1)}/3 + 2R^{(2)}/3}{(1-\delta_s)F}\right\} \le \beta_{2,2}$$

## Message 2, Part 3:



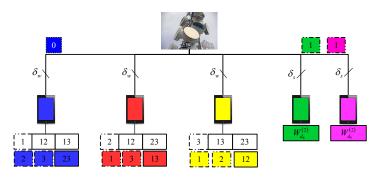
Correct decoding if

$$\max\left\{\frac{R^{(1)}/3}{(1-\delta_w)F}, \frac{R^{(1)}/3 + 2R^{(2)}/3}{(1-\delta_s)F}\right\} \le \beta_{2,3}$$

Equivalently:

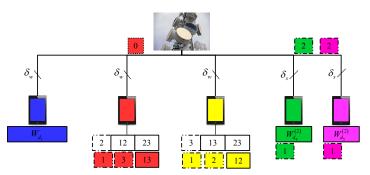
$$\max\left\{\frac{R^{(1)}}{(1-\delta_w)F}, \frac{R^{(1)}+2R^{(2)}}{(1-\delta_s)F}\right\} \le \beta_2$$

## Message 3, Part 1:



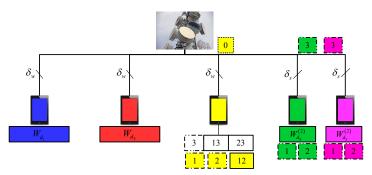
$$\max\left\{\frac{R^{(0)}}{(1-\delta_w)F}, \frac{R^{(0)}+2R^{(1)}/3}{(1-\delta_s)F}\right\} \le \beta_{3,1}$$

## Message 3, Part 1:



$$\max\left\{\frac{R^{(0)}}{(1-\delta_w)F}, \frac{R^{(0)}+2R^{(1)}/3}{(1-\delta_s)F}\right\} \le \beta_{3,2}$$

## Message 3, Part 3:



Correct decoding if

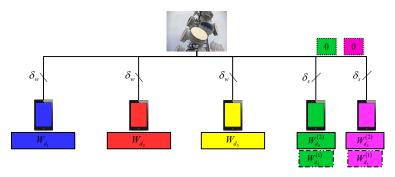
$$\max\left\{\frac{R^{(0)}}{(1-\delta_w)F}, \frac{R^{(0)}+2R^{(1)}/3}{(1-\delta_s)F}\right\} \le \beta_{3,3}$$

Equivalently:

$$\max\left\{\frac{3R^{(0)}}{(1-\delta_w)F}, \frac{3R^{(0)}+2R^{(1)}}{(1-\delta_s)F}\right\} \le \beta_3$$

# SCC Scheme: Delivery phase

### Message 4:



Correct decoding if

$$\frac{2R^{(0)}}{(1-\delta_s)F} \le \beta_4$$

# Achievable Memory-Rate Pair Analysis

• Message 1:  $\frac{R^{(2)}/3}{(1-\delta_w)F} \le \beta_1$ 

• Message 2: max  $\left\{\frac{R^{(1)}}{(1-\delta_w)F}, \frac{R^{(1)}+2R^{(2)}}{(1-\delta_s)F}\right\} \le \beta_2$ 

• Message 3: max  $\left\{ \frac{3R^{(0)}}{(1-\delta_w)F}, \frac{3R^{(0)}+2R^{(1)}}{(1-\delta_s)F} \right\} \le \beta_3$ 

• Message 4:  $\frac{2R^{(0)}}{(1-\delta_s)F} \le \beta_4$ 

 $\beta_i$ s chosen such that:

$$\frac{R^{(2)}/3}{(1-\delta_w)F} + \frac{R^{(1)}}{(1-\delta_w)F} + \frac{3R^{(0)}}{(1-\delta_w)F} + \frac{2R^{(0)}}{(1-\delta_s)F} = 1$$

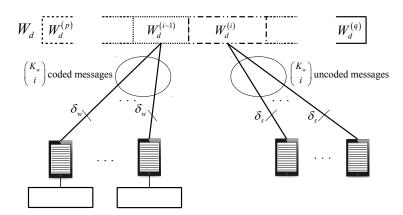
Choose rates s.t. max achieved by equality.

$$R^{(0)} + R^{(1)} + R^{(2)} = R$$

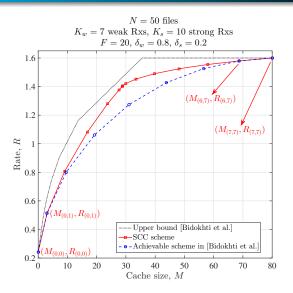
Required cache capacity:

$$M = \frac{R^{(1)}}{3} + \frac{2R^{(2)}}{3}$$

# Summary of SCC Scheme

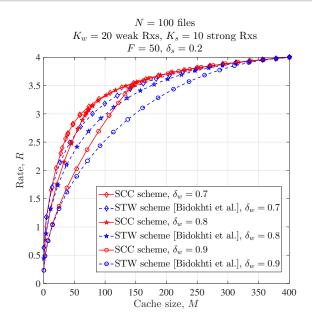


### Rate-Memory Trade-off



$$N = 3, K = 3$$

# Rate-Memory Trade-off



# Cache-Aided Delivery over the Wireless Edge

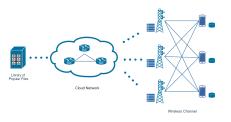
### System overview

- $K_T \times K_R$  interference channel
- Transmitter cache:  $M_TF$
- Receiver cache  $M_RF$

### Sum Degrees-of-Freedom

$$DoF(M_T, M_R) = \liminf_{P \to \infty} \frac{C(M_T, M_R, P)}{\log(P)}$$

 Decentralized caching at user terminals (RXs)



Novel scheme combining:

- Zero-forcing
  - Interference cancellation
  - Interference alignment

### Fog-Aided Radio Access Networks

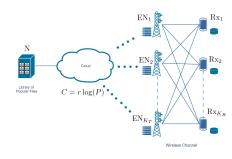
### System overview

- Fronthaul connections to base stations
- Uncached contents can be delivered from the cloud server

### Normalized Delivery Time

$$\delta(M_T, M_R) = \lim_{P \to \infty} \lim_{F \to \infty} \frac{T_F + T_E}{F/\log(P)}.$$

- Orthogonal backhaul links
- Fronthaul capacity r unknown during placement
- Serial/ pipelined fronthaul delivery



- Hard-transfer fronthauling
- Joint edge and cloud delivery

A. Sengupta, R. Tandon, and O. Simeone, Cloud and cache-aided wireless networks: Fundamental latency trade-offs, IEEE Trans. on Information Theory, Nov. 2017.

J. Pujol-Roig, F. Tosato, and D. Gündüz, **Storage-latency trade-off in cache-aided fog radio access networks**, to appear in IEEE Int'l Conf. on Communications, Kansas City, MI, May. 2018.

# **Proactive Caching for Resource Optimization**



- Channel and network conditions vary over time
- State of the art: Reactive content delivery
- User behaviour (demands and mobility) are highly predictable
- Contents can be pushed in advance when channel is good.

A. C. Gungor and D. Gündüz, **Proactive wireless caching at mobile user devices for energy efficiency**, Int'l Symp. on Wireless Comm. Systems (ISWCS), 2015.

M. Gregori, J. Gomez-Vilardebo, J. Matamoros, and D. Gündüz, Wireless content caching for small cell and D2D networks. IEEE Journal on Selected Areas in Communications. May 2016.

# Proactive Caching for Energy Efficiency

- Demands known/ predicted in advance
- Finite capacity cache at user terminal
- System model:
  - Duration of time slot i:  $\tau_i$
  - User demand rate:  $d_i$
  - Channel state:  $h_i$
  - Cache capacity: B
  - Rate-power function:  $r(t) = \log(1 + h(t)p(t))$

-- Demand Curve

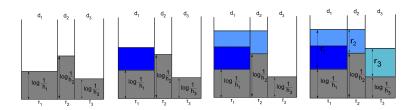
• Objective: Minimize energy consumption over *N* timeslots:

$$\min_{r_i \geq 0} \sum_{i=1}^N \tau_i \frac{e^{r_i} - 1}{h_i}$$
s.t. 
$$\sum_{i=1}^n \tau_i (d_i - r_i) \leq 0, \text{ for } n = 1, \dots, N,$$

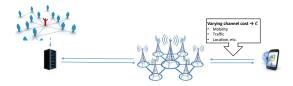
$$\sum_{i=1}^{n} \tau_{i}(r_{i} - d_{i}) - B \leq 0, \text{ for } n = 1, \dots, N.$$

# Sequential Backwards Waterfilling

- Download demands over a longer period, and in better channel conditions
- Each file can be downloaded only in advance, not later than when it is requested
- Proactive caching amount is limited by cache memory



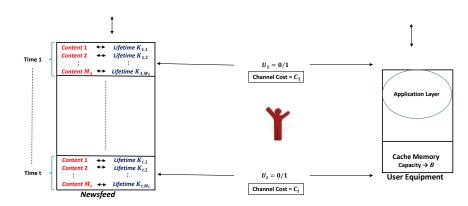
# Proactive Caching in a Dynamic Environment



- Contents generated randomly, with random lifetime
- User accesses at random time instants to download all relevant contents (e.g., online social network)
- Cost = Channel cost of download × downloaded data
- Goal: Minimize long-term average cost
- Proactively cache content at favourable channel conditions

S. Somuyiwa, A. Gyorgy and D. Gündüz, A Reinforcement-Learning Approach to Proactive Caching in Wireless Networks, revised. *IEEE Journal on Selected Areas in Communications*, 2018.

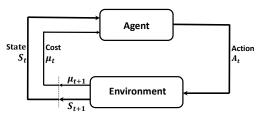
### System Model



#### System State:

- Relevant contents outside cache  $\Rightarrow \mathcal{O}_t$ .
- Contents inside cache  $\Rightarrow \mathcal{I}_t (|\mathcal{I}_t| \leq B)$ .
- Elapsed time since last user access  $\Rightarrow E_t$ .
- Energy cost of downloading a content  $\Rightarrow C_t$  (0 <  $C_t \le C_{max}$ ): i.i.d. over time.

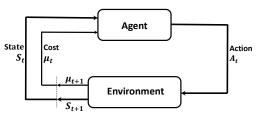
#### MDP Model



### Markov decision process with side information (MDP-SI).

- ▶ State ( $s \in S$ ):
  - Controllable state:  $(\mathcal{O}_t, \mathcal{I}_t, E_t)$ .
  - Uncontrollable state:  $C_t \Rightarrow$  side information
- ▶ Action ( $a \in A_s$ ):  $A_t = (A_t^{(1)}, A_t^{(2)})$ .
- ▶ Transition probability:  $P(S_{t+1}|S_t, A_t)$ .
- ▶ Cost function:  $\mu(S_t, A_t) = C_t \cdot |A_t^{(1)}|$ .
- ▶ Objective function:  $\rho = \lim_{T \to \infty} \mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}\mu(S_t, A_t)\right].$

### MDP Model



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# Structure of Optimal Policy

For any state  $s = (\mathcal{O}, \mathcal{I}, E) \in \mathcal{S}$ , the optimal policy  $\pi^*(s)$  has a threshold structure with respect to cost C.

- ▶ Let
  - $l_1 \leq \cdots \leq l_B$  :contents in the cache  $(\mathcal{I})$ .
  - $L_1 \ge \cdots \ge L_B$ : B contents out of cache ( $\mathcal{O}$ ) with highest lifetimes.
- $ightharpoonup \exists B' \leq B$  and corresponding threshold values:

$$\mathcal{T}(a_{B'}) \leq \mathcal{T}(a_{B'-1}) \leq \cdots \leq \mathcal{T}(a_1) \leq C_{max}$$

and the optimal policy performs simple actions  $a_i = (l_i | L_i)$ , if  $C \leq \mathcal{T}(a_i)$  and E > 0.

# Structure of Optimal Policy

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# LISO: A Suboptimal Policy

### ► Longest lifetime in-Shortest lifetime out:

- Swap largest  $L \in \mathcal{O}$  with the smallest  $l \in \mathcal{I}$ , if  $C_t \leq \mathcal{T}(a)_{a=(l|L)}$ , until no more swaps can be performed.
- Single threshold value for each pair (l|L) of lifetimes.
- $oldsymbol{\circ}$  Parametrized by threshold values:  $oldsymbol{ heta} = \mathcal{T}(l|L)$  for all L>l.

### Policy Representation

Threshold values obtained using linear function approximation (LFA) as

$$\mathcal{T}(a)_{a=(l|L)} = \sum_{i=0}^{K_{max}} \phi(i)\theta_i(l,L) = \Phi^\top \theta(l,L) ,$$

 $K_{max}$ : maximum lifetime

 $\Phi_t = [\phi_t(0), \phi_t(1), \dots, \phi_t(K_{max})]$ : frequency vector

$$\phi(i) \triangleq \frac{\sum_{l \in \mathcal{C}} \mathbb{I}_{\{l=i\}}}{B}, \quad \text{for} \quad i = 0, 1, \dots, K_{max},$$

 $\theta_i(l,L)$ : coefficients to be optimized for each simple action.

### Policy Search

▶ A model free policy search technique using stochastic gradient descent.

### Policy Gradient Algorithm

- generate "samples" with P(s'|s,a) and the probability density function  $f_C(c)$ 
  - Results in trajectory  $\tau_{\pi_{\theta}} = (S_1, C_1, A_1), \dots, (S_T, C_T, A_T)$  i.e.,  $\tau_{\pi_{\theta}, T} \sim P_{\theta, T}(\tau_{\pi_{\theta}}) = P(\tau_{\pi_{\theta}, T}|\theta).$
- Evaluate average sample cost  $J_{\pi_{\theta}} = \frac{1}{T} \sum_{t=1}^{T} \mu(S_t, A_t)$
- Update  $\theta$  in the direction that decreases  $\rho^{\pi_{\theta}} = \mathbb{E}[J_{\pi_{\theta}}]$ :

$$\boldsymbol{\theta}_{j+1} = \boldsymbol{\theta}_j - \lambda \nabla_{\boldsymbol{\theta}} \rho^{\pi_{\boldsymbol{\theta}}},$$

where  $\lambda > 0$  is the step size, j is the current iteration step and

$$\nabla_{\boldsymbol{\theta}} \rho^{\pi_{\boldsymbol{\theta}}} = \int_{\tau} \nabla_{\boldsymbol{\theta}} P_{\boldsymbol{\theta}}(\tau_{\pi_{\boldsymbol{\theta}}}) J_{\pi_{\boldsymbol{\theta}}} d\tau .$$

### Performance Bounds

- Unlimited cache capacity (LB-UC)
  - Decouples actions for contents,  $A_t^{(2)} = \emptyset$ ,  $\forall t$
  - Threshold  $\mathcal{T}_L$ : Content with lifetime L is downloaded if  $C \leq \mathcal{T}_L$ .

$$0 \leq \mathcal{T}_1 \leq \cdots \leq \mathcal{T}_{K_{max}} \leq C_{max}$$

- Threshold obtained using value iteration algorithm (VIA)
- Non-causal knowledge of user access times (LB-NCK)
  - For any time-to-user access t', contents are downloaded if  $C_t \leq \mathcal{T}_{t'}$ .

$$0 \leq \mathcal{T}_{D_{max}} \leq \cdots \leq \mathcal{T}_1 \leq C_{max}$$

- where  $D_{max}$  is the bound on the user access interval.
- Threshold values obtained using VIA.

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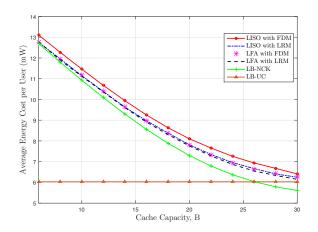
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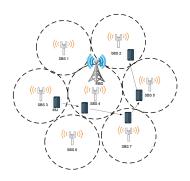
## Average Energy Cost Vs Cache Capacity



Percentage Improvement over LISO with FDM:

- ► LFA with LRM  $\rightarrow$  up to 5.6%. ► LFA with FDM  $\rightarrow$  up to 4.4%.
  - ▶ LISO with LRM  $\rightarrow$  up to 4.2%.

# Mobility and Popularity Aware Small Cell Caching

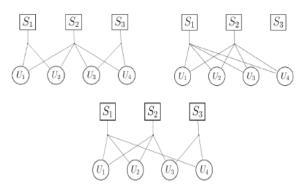


- Random mobility patterns
- Maximum distance separable (MDS) coded content storage
- How to allocate cached to contents with different popularities?

M. Ozfatura and D. Gündüz, Mobility and popularity aware coded small-cell caching, IEEE Communication Letters, vol. 22, no. 2, pp. 288 - 291, Feb. 2018.

K. Shanmugam, N. Golrezaei, A. G. Dimakis, A. F. Molisch, and G. Caire. Femtocaching: Wireless content delivery through distributed caching helpers. IEEE Trans. Inf. Theory, Dec. 2013.

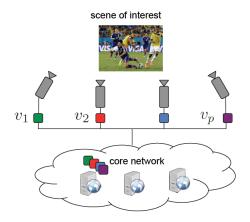
# Multi-Server System with Random Topology



- Each user connects to  $\rho$  out of P servers
- Each server can cache  $N/\rho$  files
- Both coded caching and MDS coded storage need to be utilised

N. Mital, D. Gündüz and C. Ling, Coded caching in a multi-server system with distributed storage, to appear in Int'l Wireless Communications and Networking Conference, Barcelona, Spain, Apr. 2018.

### Cache-Aided Interactive Multiview Video Streaming in Small Cell Networks



- Interactive multiview streaming
- How to optimally cache and deliver multiview video content to improve the free viewpoint streaming experience?

E. Bourtsoulatze and D. Gündüz, Cache-aided interactive multiview video streaming in small cell networks, submitted for publication.

Thank You for Your Attention!