

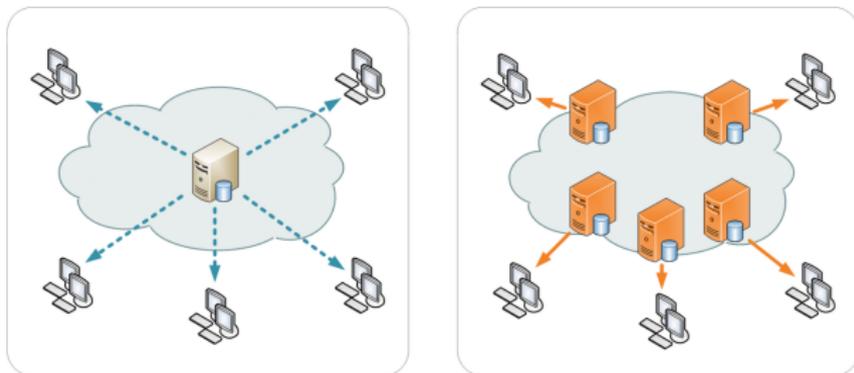
# An Information-Theoretic View of Cache-Aided Networks: Part 1 – Coded Caching

Michèle Wigger

Telecom ParisTech, 9 february 2021

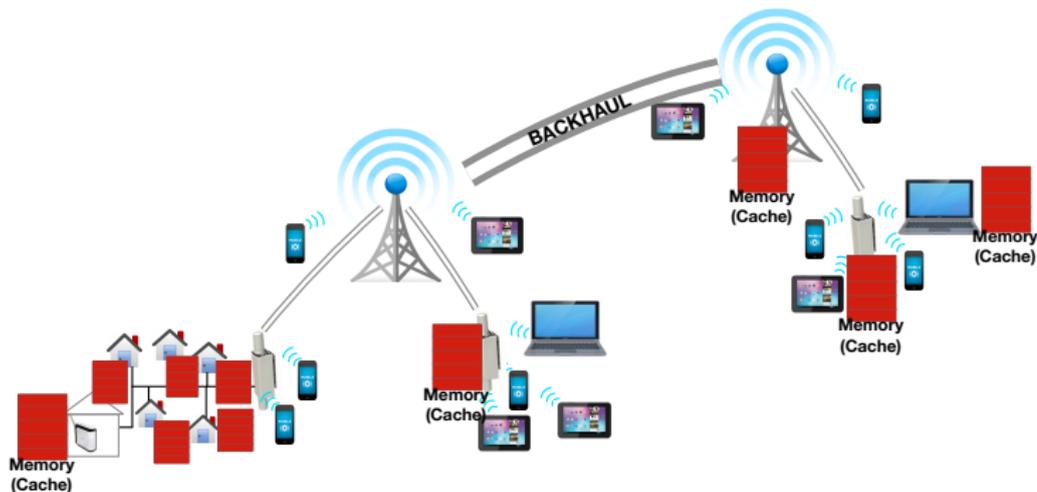


# Content Delivery Networks



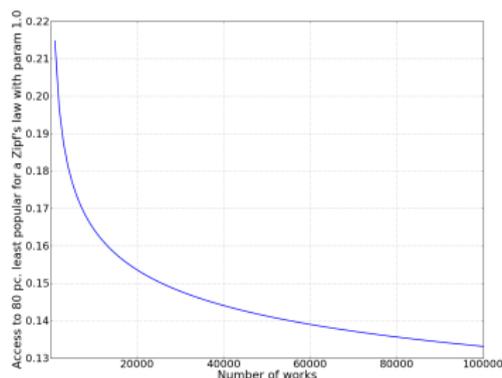
- Store contents in caches before file demands even known
- Reduce network load and latency during high-congestion periods
- Idea useful if certain files very popular and known in advance

# Distributed Caches: Promising Solution for Cellular Networks



- Can cache at main BSs, picoBSs, femtoBSs, or directly at end users

# File Popularities



- Static file popularity follows a Zipf distribution  $P(x) = Cx^{-\alpha}$
- Evolution of file popularities (youtube videos) can also be predicted

Use pro-active caching to improve cellular systems!

- decrease network load
- decrease latency

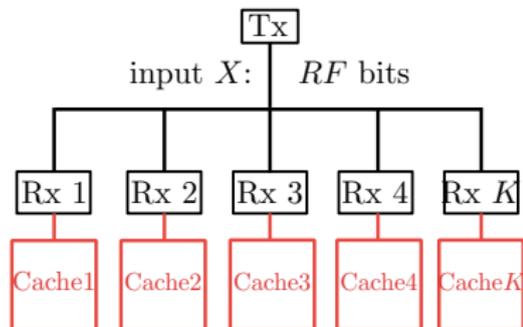
## Information-Theoretic View of Caching

- Go beyond obvious *local* caching gain
- Create coding opportunities through smart cache placement
- Exploit multi-cast opportunities
  - serve many users/demands with same signals
- **Global caching gain**
  - receivers can profit from other receivers' cache memories

## A Simple Network

Library:

Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each



- All files equally popular  $\rightarrow$  we consider only most popular files
- All files equally large.
- Before the actual transmission there is an idle period where the transmitter (server) can fill the receivers' cache memories.
- Cache placement phase only constrained by memory sizes.

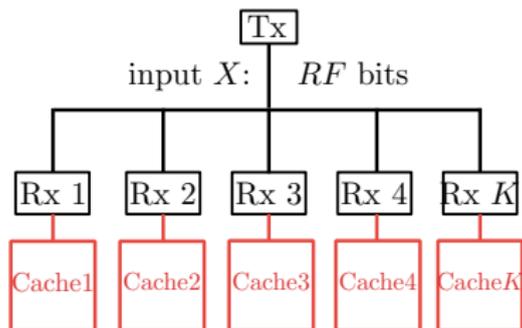
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[1] M. A. Maddah-Ali, U. Niesen, "Fundamental Limits of Caching." *IEEE Transactions on Information Theory*.

## A Simple Network

Library:

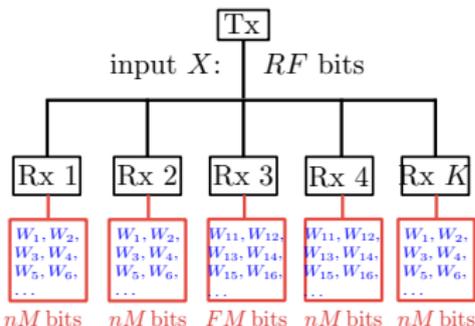
Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each



Communication in two phases:

# A Simple Network

Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each

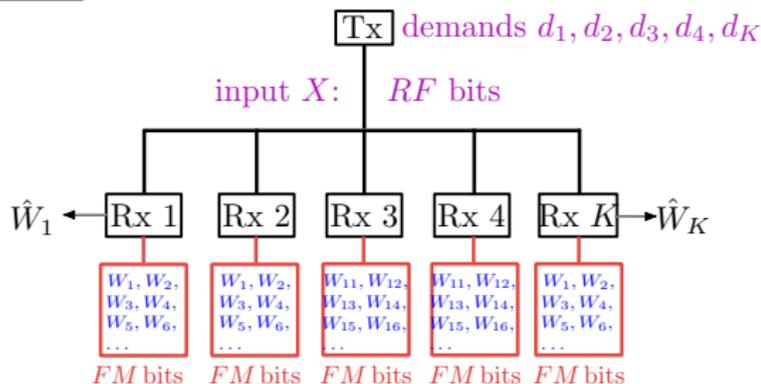


Communication in two phases:

- **Placement phase**: Tx fills caches without knowing which receiver demands which message

## A Simple Network

Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each



Communication in two phases:

- **Placement phase**: Tx fills caches without knowing which receiver demands which message
- **Delivery phase**:
  - Receiver  $k$  wants file  $W_{d_k}$   $\rightarrow$  sends demand  $d_k$  to transmitter
  - Tx describes  $W_{d_1}, \dots, W_{d_K}$  to Rxs  $1, \dots, K$  through input  $X$
  - Tx describes demands  $d_1, \dots, d_K$  to all receivers

## Fundamental Rate-Memory Tradeoff $R^*(M)$

$$R^*(M) := \min \left\{ R : \text{such that for } (R, M) \text{ goal can be achieved} \right. \\ \left. \text{for all demands } d_1, \dots, d_K. \right\}$$

Some properties:

- $R^*(M)$  is decreasing in  $M$ .
- $R^*(M)$  is bounded above by  $\min\{N, K\}$ . Moreover:

$$R^*(M = 0) = \min\{N, K\}.$$

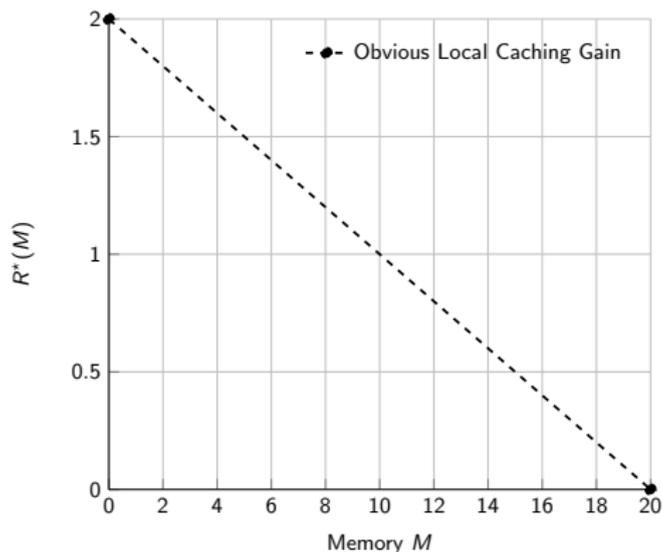
- $R^*(M)$  is nonnegative. Moreover:

$$R^*(M) = 0, \quad \forall M \geq N.$$

## Obvious Upper Bound on $R^*(M)$

$$R^*(M) \leq \min\{K, N\} \left(1 - \frac{M}{N}\right).$$

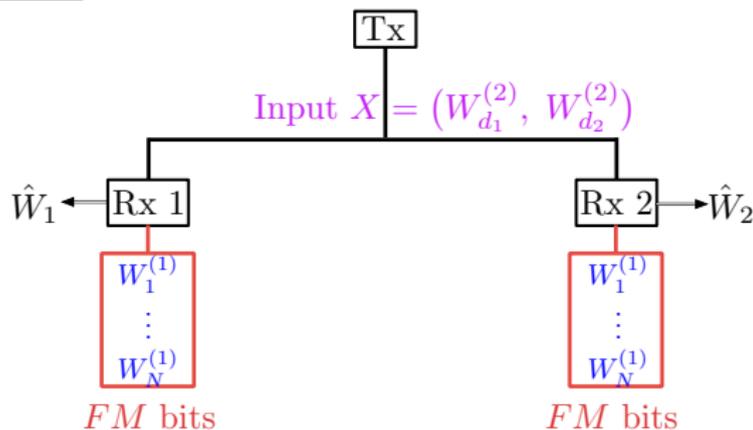
- Example with  $N = 20$  files and  $K = 2$  users



- Achieved through time/memory sharing or by the following naive scheme...

## Naive Scheme for $K = 2$ Receivers

Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each



- Split  $W_d$  into two parts  $(W_d^{(1)}, W_d^{(2)})$  of sizes  $\frac{FM}{N}$  and  $F\left(1 - \frac{M}{N}\right)$  bits
- For  $d = 1, \dots, N$ : cache part  $W_d^{(1)}$  at both Rxs 1 and 2
- Delivery input  $X = (W_{d_1}^{(2)}, W_{d_2}^{(2)})$  (if  $d_1 \neq d_2$ )

In the worst-case, delivery rate needs to be  $R = 2\left(1 - \frac{M}{N}\right)$ .

## Naive Scheme for $K$ Receivers

- Split  $W_d$  into two parts  $(W_d^{(1)}, W_d^{(2)})$  of sizes  $F \frac{M}{N}$  and  $F(1 - \frac{M}{N})$  bits
- For  $d = 1, \dots, N$ : cache part  $W_d^{(1)}$  at all rxs
- Deliver part  $W_d^{(2)}$  for each demanded message  $W_d$ .
  - If  $K \geq N$ , in the worst case:

$$X = (W_{d_1}^{(2)}, W_{d_2}^{(2)}, \dots, W_{d_K}^{(2)}).$$

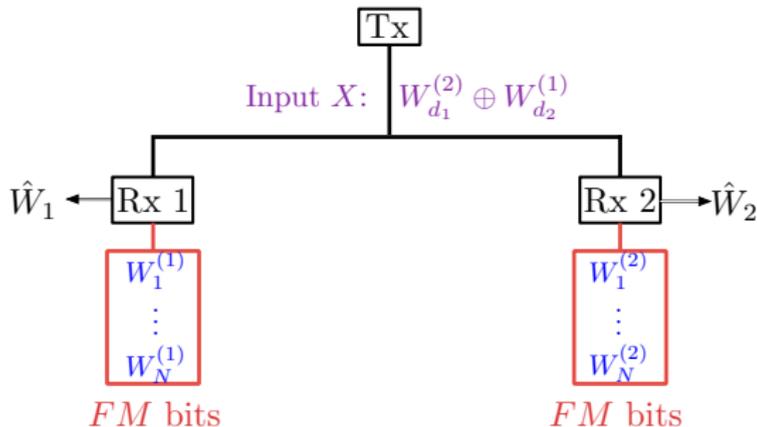
- If  $K < N$ , in the worst case:

$$X = (W_1^{(2)}, W_2^{(2)}, \dots, W_N^{(2)}).$$

Required Delivery Rate is  $R = \min\{K, N\} \cdot (1 - \frac{M}{N})$ .

## Coded caching for $K = 2$ Receivers [Maddah-Ali&Niesen 2014]

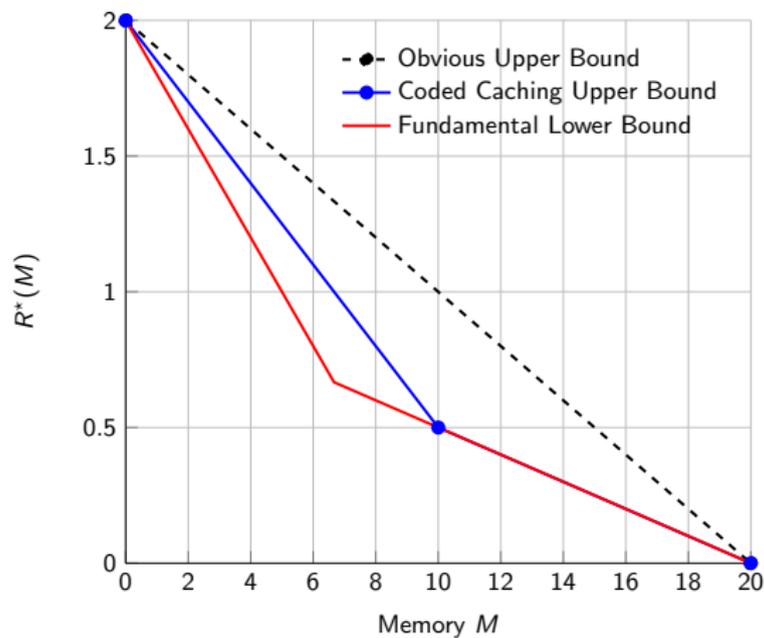
Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each



- Split  $W_d$  into two parts ( $W_d^{(1)}, W_d^{(2)}$ ) each of  $\frac{F}{2}$  bits
- For  $d = 1, \dots, N$ : cache part  $W_d^{(1)}$  at Rx1 and part  $W_d^{(2)}$  at Rx2
- Delivery input  $X = W_{d_1}^{(2)} \oplus W_{d_2}^{(1)}$

Achieves Rate-Memory Pair  $M = \frac{N}{2}$  and  $R = \frac{1}{2}$ .

## Fundamental Limit and Bounds for 2 Users ( $N = 20$ files)



## Time- and Memory Sharing for arbitrary parameter $\alpha \in [0, 1]$

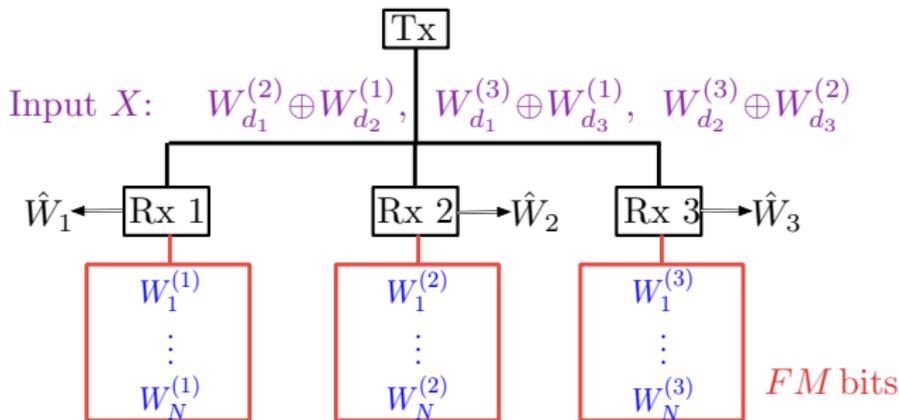
- Assume Scheme 1 achieves  $(R_1, M_1)$  and Scheme 2 achieves  $(R_2, M_2)$
- Split each file  $W_d = (W_d^{(1)}, W_d^{(2)})$  consisting of  $\alpha F$  and  $(1 - \alpha)F$  bits each.
- Apply placements of Scheme 1 to  $\{W_d^{(1)}\}$  using  $M_1\alpha F$  bits of memory and placements of Scheme 2 to  $\{W_d^{(2)}\}$  using  $M_2(1 - \alpha)F$  bits of memory
- Apply delivery of Scheme 1 to  $\{W_d^{(1)}\} \rightarrow$  signal  $X^{(1)}$  of  $R_1\alpha F$  bits; and apply delivery of Scheme 2 to  $\{W_d^{(2)}\} \rightarrow$  signal  $X^{(2)}$  of  $R_2(1 - \alpha)F$  bits

Total cache memory  $MF = M_1\alpha F + M_2(1 - \alpha)F$ ; and  
total number of delivery bits  $RF = R_1\alpha F + R_2(1 - \alpha)F$

$\Rightarrow (\alpha M_1 + (1 - \alpha)M_2, \alpha R_1 + (1 - \alpha)R_2)$  is achievable  $\forall \alpha \in [0, 1]$

## Coded caching for $K = 3$ Receivers, Parameter $t = 1$

Library: Files  $W_1, W_2, \dots, W_N$  of  $FR$  bits each

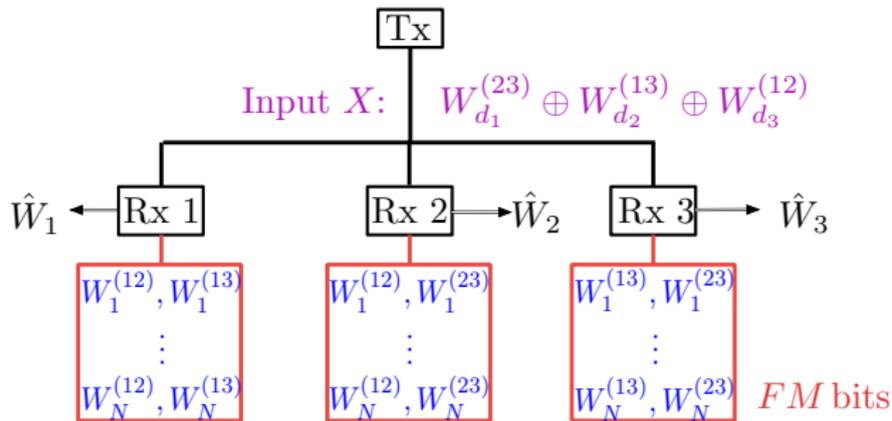


- Split  $W_d$  into three parts ( $W_d^{(1)}, W_d^{(2)}, W_d^{(3)}$ ) each  $\frac{F}{3}$  bits
- For  $d = 1, \dots, N$ : cache part  $W_d^{(1)}$  at Rx1, part  $W_d^{(2)}$  at Rx2, and part  $W_d^{(3)}$  at Rx3
- Delivery input  $X = W_{d_1}^{(2)} \oplus W_{d_2}^{(1)}, W_{d_1}^{(3)} \oplus W_{d_3}^{(1)}, W_{d_2}^{(3)} \oplus W_{d_3}^{(2)}$

Achieves Rate-Memory Pair  $M = \frac{N}{3}$  and  $R = 1$ .

## Coded caching for $K = 3$ Receivers, Parameter $t = 2$

Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each

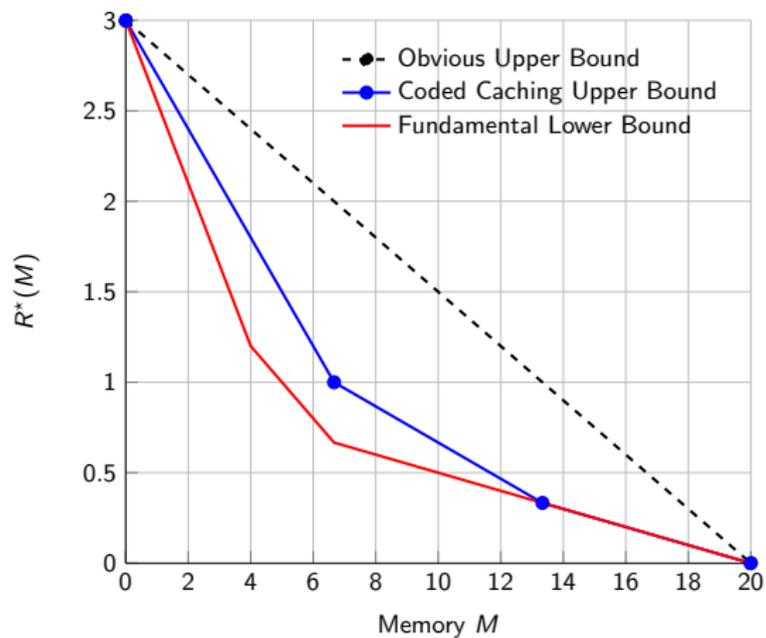


- Split  $W_d$  into three parts ( $W_d^{(12)}$ ,  $W_d^{(13)}$ ,  $W_d^{(23)}$ ) each of  $\frac{F}{3}$  bits
- For  $d = 1, \dots, N$ : cache part  $W_d^{(12)}$  at Rxs 1 and 2, part  $W_d^{(13)}$  at Rxs 1 and 3, and part  $W_d^{(23)}$  at Rxs 2 and 3
- Delivery input  $X = W_{d_1}^{(23)} \oplus W_{d_2}^{(13)} \oplus W_{d_3}^{(12)}$

Achieves Rate-Memory Pair

$$M = \frac{2N}{3} \quad \text{and} \quad R = \frac{1}{3}.$$

## Bounds for 3 Users ( $N = 20$ files)



## Coded Caching for $K$ Users

- Parameter  $t \in \{1, \dots, K - 1\}$
- *Placement*: Split each  $W_d$  into  $\binom{K}{t}$  parts and save each part at a different subset of receivers  
Let for each size- $t$  subset  $\mathcal{G}$  denote  $W_d^{\mathcal{G}}$  the part of  $W_d$  placed in caches of all receivers in  $\mathcal{G}$ .
- *Delivery transmission*: For each set  $\mathcal{S} = \{s_1, \dots, s_{t+1}\}$  of size  $t + 1$ , send

$$W_{\text{XOR}, \mathcal{S}} := \bigoplus_{\ell=1}^{t+1} W_{d_{s_\ell}}^{(\mathcal{S} \setminus \{s_\ell\})}$$

- *Delivery reception*: Receiver  $s_j$  has stored in its cache memory

$$W_{d_{s_\ell}}^{(\mathcal{S} \setminus \{s_\ell\})}, \quad \forall \ell \in \{1, \dots, j-1, j+1, \dots, t\}.$$

So, with  $W_{\text{XOR}, \mathcal{S}}$  it can recover  $W_{d_{s_j}}^{(\mathcal{S} \setminus \{s_j\})}$ .

This way it can recover all missing parts of  $W_{d_{s_j}}$ .

## Analysis of Coded Caching for $K$ Users

- Fix parameter  $t \in \{1, \dots, K - 1\}$
- Each part of a file is of size

$$F \cdot \binom{K}{t}^{-1} \text{ bits.}$$

- Each receiver stores  $\binom{K-1}{t-1}$  parts of each file. So the placement requires cache memory

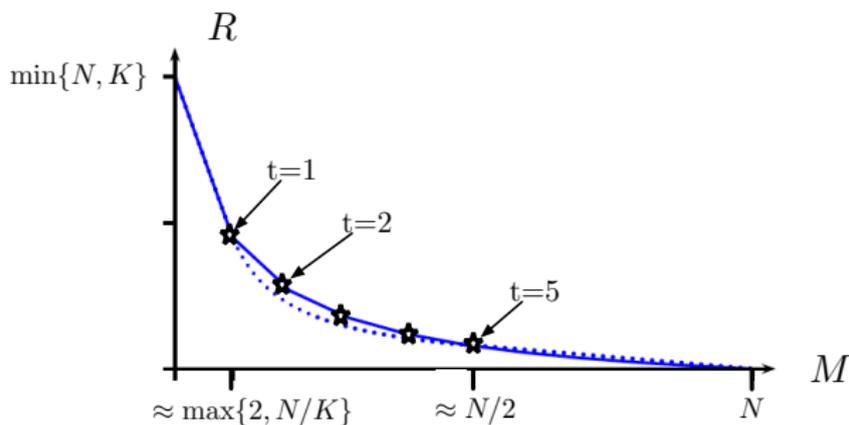
$$M = N \frac{\binom{K-1}{t-1}}{\binom{K}{t}} = N \frac{t}{K}. \quad (\text{increasing in } t)$$

- Coded caching sends an XOR-message to each subset of  $t + 1$  receivers. So the total rate is

$$R = \frac{\binom{K}{t+1}}{\binom{K}{t}} = \frac{K - t}{t + 1}. \quad (\text{decreasing in } t)$$

## Performance of Coded Caching

- $K = 6$



### Coded Caching Upper Bound

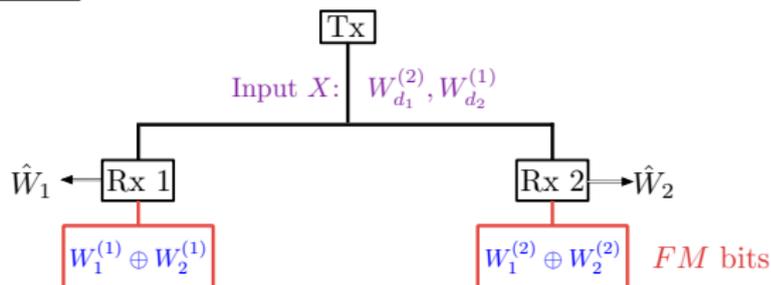
For all  $M \in \frac{N}{K} \cdot \{0, 1, \dots, K-1, K\}$ :

$$R^*(M) \leq \min \left\{ K \left( 1 - \frac{M}{N} \right) \left( 1 + \frac{MK}{N} \right)^{-1}, N \left( 1 - \frac{M}{N} \right) \right\}.$$

## Achievability can be Improved!

Example:  $K = 2$  and  $N = 2$ :

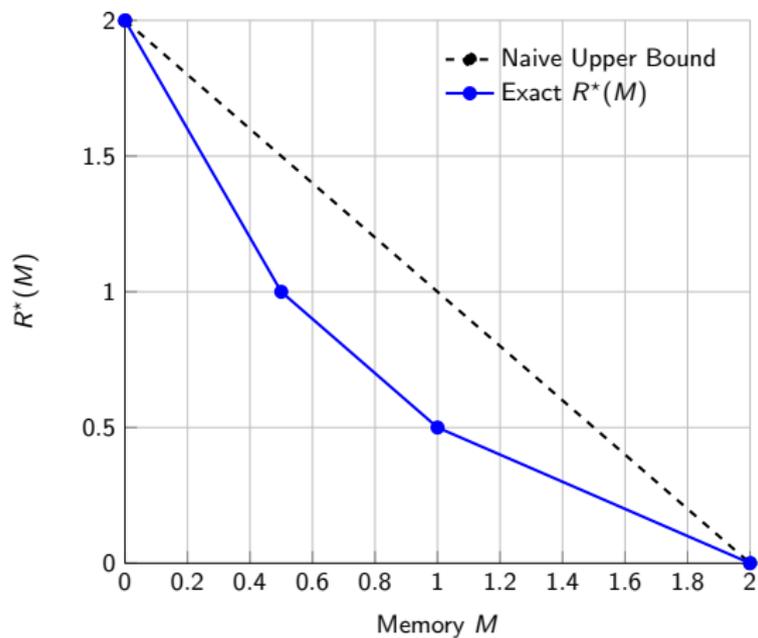
Library: Files  $W_1$  and  $W_2$  of  $F$  bits each



- Split  $W_d$  into two parts ( $W_d^{(1)}$ ,  $W_d^{(2)}$ ) each of  $\frac{F}{2}$  bits
- For  $d = 1, \dots, N$ : cache part  $W_d^{(1)}$  at Rx1 and part  $W_d^{(2)}$  at Rx2
- Delivery input  $X = W_{d_1}^{(2)}, W_{d_2}^{(1)}$

Achieves Rate-Memory Pair  $M = \frac{1}{2}$  and  $R = 1$ .

## Exact Rate-Memory Tradeoff $R^*(M)$ for $K = N = 2$



## First Lower Bound $R \geq s - \frac{s}{\lfloor N/s \rfloor} M$

- Consider only Receivers  $1, \dots, s$ , where  $s \leq \min\{N, K\}$
- Consider demand vectors

$$\begin{aligned} \mathbf{d}^{(1)} &:= (1, \dots, s) \\ \mathbf{d}^{(2)} &:= (s+1, \dots, 2s) \\ &\dots \\ \mathbf{d}^{(\lfloor N/s \rfloor)} &:= ((\lfloor N/s \rfloor - 1)s + 1, \dots, \lfloor N/s \rfloor s) \end{aligned}$$

- Let  $X^{(\ell)}$  denote the delivery input for demand  $\mathbf{d}^{(\ell)}$
- From  $X^{(1)}, \dots, X^{(\lfloor N/s \rfloor)}$  and  $Z_1, \dots, Z_s$  one can calculate  $W_1, \dots, W_{\lfloor N/s \rfloor s}$ :

$$\begin{aligned} &H(X^{(1)}, \dots, X^{(\lfloor N/s \rfloor)}, Z_1, \dots, Z_s) \geq H(W_1, \dots, W_{\lfloor N/s \rfloor s}) \\ \iff &H(X^{(1)}, \dots, X^{(\lfloor N/s \rfloor)}) + H(Z_1, \dots, Z_s) \geq H(W_1, \dots, W_{\lfloor N/s \rfloor s}) \\ \iff &FR \lfloor N/s \rfloor + sFM \geq \lfloor N/s \rfloor sF \\ \iff &R \geq s - \frac{s}{\lfloor N/s \rfloor} M. \end{aligned}$$

## Second Lower Bound $R \geq s - \frac{s^2}{N} M$

- Consider only Receivers  $1, \dots, s$ , where  $s \leq \min\{N, K\}$
- $\mathcal{D}_s$ : all demand vectors  $\mathbf{d}_s$  of  $s$  users having  $s$  different demands.

$$|\mathcal{D}_s| = \binom{N}{s} s!$$

- Let  $X^{(\mathbf{d}_s)}$  denote the delivery input for demand  $\mathbf{d}_s$
- $\forall \mathbf{d}_s \in \mathcal{D}_s$  it holds that:

$$\begin{aligned} F \cdot R &\geq H(X^{\mathbf{d}_s}) \geq I(X^{\mathbf{d}_s}; W_{d_1}, \dots, W_{d_s}, Z_1, \dots, Z_s) \\ &\geq I(X^{\mathbf{d}_s}; W_{d_1}, \dots, W_{d_s} | Z_1, \dots, Z_s) \\ &= H(W_{d_1}, \dots, W_{d_s} | Z_1, \dots, Z_s). \end{aligned}$$

- By averaging over all demands  $\mathbf{d}_s \in \mathcal{D}_s$  and by Han's inequality:

$$\begin{aligned} FR &\geq \frac{s}{N} \left( H(W_1, \dots, W_N) - I(W_1, \dots, W_N; Z_1, \dots, Z_s) \right) \\ &\geq F \left( s - \frac{s^2}{N} M \right). \end{aligned}$$

### Third Lower Bound $R \geq s - \sum_{i=1}^s \frac{i}{N-i+1} M$

- Consider only Receivers  $1, \dots, s$ , where  $s \leq \min\{N, K\}$
- $\mathcal{D}_s$ : all demand vectors  $\mathbf{d}_s$  of  $s$  users having  $s$  different demands.

$$|\mathcal{D}_s| = \binom{N}{s} s!$$

- Let  $X^{(\mathbf{d}_s)}$  denote the delivery input for demand  $\mathbf{d}_s$
- $\forall \mathbf{d}_s \in \mathcal{D}_s$  it holds that:

$$F \cdot R \geq sF - \sum_{i=1}^s I(W_{d_i}; Z_1, \dots, Z_i | W_{d_1}, \dots, W_{d_{i-1}})$$

- By averaging over all demands  $\mathbf{d}_s \in \mathcal{D}_s$

$$\begin{aligned} FR &\geq sF - \sum_{i=1}^s \frac{1}{\binom{N}{s} s!} \sum_{\mathbf{d}_s \in \mathcal{D}_s} I(W_{d_i}; Z_1, \dots, Z_i | W_{d_1}, \dots, W_{d_{i-1}}) \\ &\geq F \left( s - \sum_{i=1}^s \frac{i}{N-i+1} M \right). \end{aligned}$$

## Gap between Upper and Lower Bounds

- Multiplicative Gap:

$$\frac{\text{Best Upper Bound}}{\text{Best Lower Bound}} \leq 2.315.$$

- Upper bound from a more constrained scenario

$$R_{\text{Dec}} := \frac{N - M}{M} \left( 1 - \left( 1 - \frac{M}{N} \right)^{\min\{K, N\}} \right)$$

- Third lower bound is piecewise linear over intervals  $[M_{\ell+1}, M_{\ell}]$  with

$$M_{\ell} = \begin{cases} \frac{N-\ell}{\ell+1} & \text{if } \ell \in \{0, 1, \dots, \min\{K, N\} - 1\}, \\ 0 & \text{if } \ell = \min\{N, K\}. \end{cases}$$

- Make  $R_{\text{Dec}}$  piecewise linear over same intervals.
- Ratio is bilinear and it suffices to consider end-points of intervals.  
→ check all end-points!

# An Information-Theoretic View of Cache-Aided Networks: Part 2 – Decentralized Coded Caching

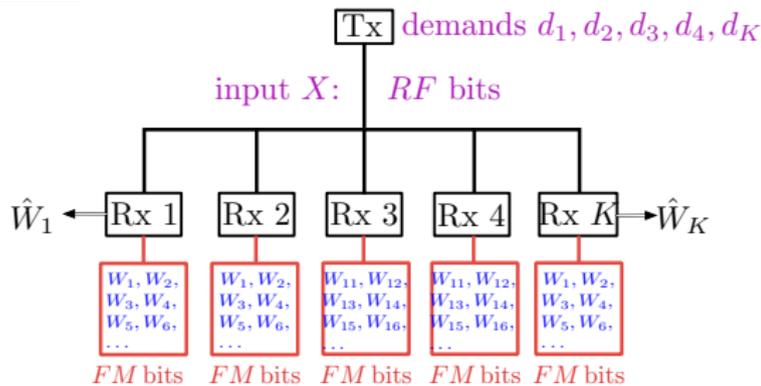
Michèle Wigger

Telecom ParisTech, 9 February 2021



## Decentralized Placement

Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each

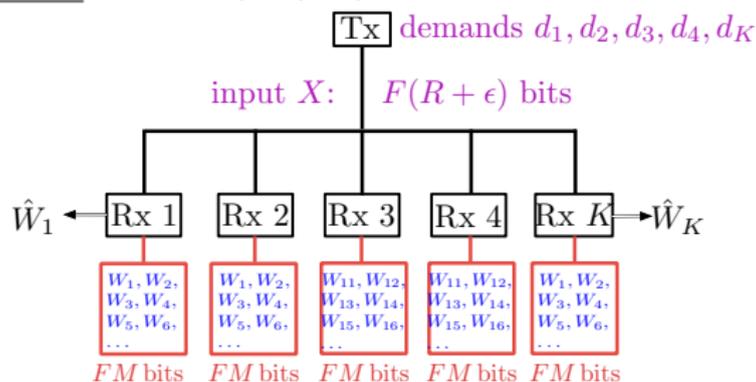


Communication in two phases:

- **Placement phase**: Each receiver randomly downloads bits from library without knowing demands or the number of receivers  $K$ !  
 $\rightarrow Z_k = g(W_1, \dots, W_N, \Theta_k)$
- **Delivery phase**:
  - Receiver  $k$  wants file  $W_{d_k} \rightarrow$  sends demand  $d_k$  to transmitter
  - Tx describes  $W_{d_1}, \dots, W_{d_K}$  to Rxs  $1, \dots, K$  through input  $X$

## Formal Problem Statement

Library: Files  $W_1, W_2, \dots, W_N$  of  $F$  bits each



- Cache placement  $Z_k = g(W_1, \dots, W_N, \Theta_k)$ ,  
where  $\Theta_k$  is a randomness known to everyone
- Delivery encoding  $X = f(W_1, \dots, W_N, d_1, \dots, d_K, \Theta_1, \dots, \Theta_K)$
- Delivery decoding  $\hat{W}_k = \varphi_k(X, Z_k, d_1, \dots, d_K, \Theta_1, \dots, \Theta_K)$ .
- Goal:  $\hat{W}_k = W_{d_k}$  for all  $k = 1, \dots, K$  **with high probability**

## Fundamental Rate-Memory Tradeoff $R_{\text{Dec}}^*(M)$

$$R_{\text{Dec}}^*(M) := \min \left\{ R : \text{such that for } (R, M) \text{ each receiver } k \in \{1, \dots, K\} \right. \\ \left. \text{learns } W_{d_k} \text{ with high probability} \right\}$$

Obvious bounds:

- Local caching gain achievable  $R_{\text{Dec}}^*(M) \leq \min\{K, N\} \left(1 - \frac{M}{N}\right)$
- Cannot improve on centralized setup:  $R_{\text{Dec}}^*(M) \geq R^*(M)$ .

## Decentralized Coded Caching Algorithm

- Placement: Each Receiver sequentially samples and stores each bit of the library with probability  $p = \frac{M}{N}$
- For each subset  $\mathcal{S} \subseteq \{1, \dots, K\}$ , define now  $W_d^{\mathcal{S}}$  the set of all bits of message  $W_d$  exclusively cached at all receivers of set  $\mathcal{S}$ .
- *Delivery Encoding:*
  - Send all demanded bits that are not cached anywhere
  - For  $t = 1, \dots, K - 1$  use the coded caching delivery scheme of parameter  $t$  to send the demanded bits

$$\{W_{d_1}^{\mathcal{S}}, \dots, W_{d_K}^{\mathcal{S}} : |\mathcal{S}| = t\}$$

Zero-padding might be required for this operation!

- Delivery decoding similar to coded caching scheme, but again for all parameters  $t = 0, 1, 2, \dots, K - 1$ .

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[2] M. A. Maddah-Ali, U. Niesen, "Decentralized coded caching attains order-optimal memory-rate tradeoff," *IEEE Trans. on Inf. Theory*.

## Analysis of Decentralized Scheme

- By the weak law of large numbers, for all  $\epsilon > 0$ :

$$\Pr \left[ \left| \left| W_d^S \right| - p^{|S|} (1-p)^{K-|S|} F \right| > \epsilon \right] \rightarrow 0 \quad \text{as} \quad F \rightarrow \infty.$$

- Expected storage

$$\begin{aligned} M &= N \sum_{t=1}^K \binom{K-1}{t-1} p^t (1-p)^{K-t} \\ &= Np \sum_{t'=0}^{K-1} \binom{K-1}{t'} p^{t'} (1-p)^{K-1-t'} = Np = M. \end{aligned}$$

- Expected rate:

$$\begin{aligned} R &= \sum_{t=0}^{K-1} \binom{K}{t+1} p^t (1-p)^{K-t} = \frac{1-p}{p} \sum_{t'=1}^K \binom{K}{t'} p^{t'} (1-p)^{K-t'} \\ &= \frac{1-p}{p} \sum_{t'=0}^K \binom{K}{t'} p^{t'} (1-p)^{K-t'} - \frac{1-p}{p} (1-p)^K \\ &= \frac{1-p}{p} (1 - (1-p)^K) = \frac{N-M}{M} \left( 1 - \left( 1 - \frac{M}{N} \right)^K \right) \end{aligned}$$

## Results for Decentralized Caching

### Upper Bound $R^*(M)$ for Decentralized Caching

$$R^*(M) \leq K \left(1 - \frac{M}{N}\right) \min \left\{ \frac{N}{KM} \left(1 - \left(1 - \frac{M}{N}\right)^K\right), \frac{N}{K} \right\}$$

- Above upper bound matches  $R^*(M)$  up to a factor of at least 12. (Proved analytically in [2].)
- Above upper bound matches coded caching upper bound for centralized scenario up to a factor of 1.6. (Shown numerically.)

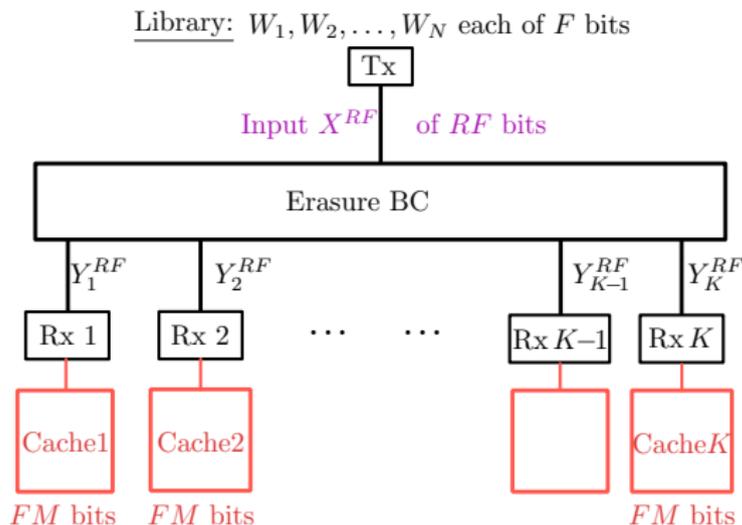
# An Information-Theoretic View of Cache-Aided Networks: Part 3 – Caching for Erasure Broadcast Channels

Michèle Wigger

Telecom ParisTech, 17 November 2020



## Delivery over Homogeneous Erasure Broadcast Channels (BC)



- Each transmitted bit erased at each receiver with probability  $\delta > 0$ , irrespective of all other bits:

$$\Pr[Y_{k,t} = X_t] = 1 - \delta \quad \text{and} \quad \Pr[Y_{k,t} = \Delta] = \delta$$

## Noisy Setup Requires Vanishing Error Probability

- Cache placement  $Z_k = g_k(W_1, \dots, W_N)$  consists of  $FM$  bits
- Delivery encoding  $X^{RF} := (X_1, \dots, X_{RF}) = f(W_1, \dots, W_N, d_1, \dots, d_K)$
- Delivery decoding:
  - Receiver  $k$  receives  $Y_k^{RF} = (Y_{k,1}, \dots, Y_{k,RF})$
  - It produces  $\hat{W}_k = \varphi_k(Y_k^{RF}, Z_k, d_1, \dots, d_K)$ .
- Goal:  $\Pr[\hat{W}_k = W_{d_k}] \rightarrow 0$  as  $F \rightarrow \infty$  for all  $k = 1, \dots, K$

→ Need to tolerate errors because the channel has nonzero probability of experiencing lots of erasures.

## Rate-Memory Tradeoff for Erasure Broadcast Channels ( $N \geq K$ )

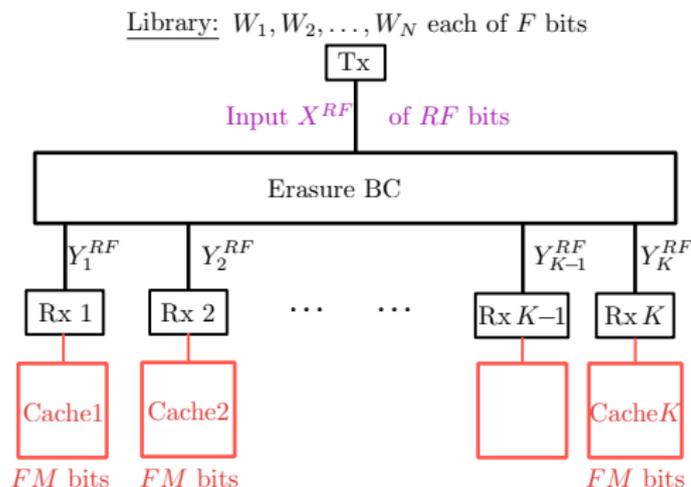
- Capacity of a erasure broadcast channel (EBC) is  $(1 - \delta)$ , both for sending common messages and private messages.
- That means, by sending  $FR$  inputs, we can convey  $FR(1 - \delta)$  information bits with arbitrarily small probability of error as  $F \rightarrow \infty$ .
- So now each XOR packet requires  $(1 - \delta)^{-1}$  times more slots for reliable transmission than before. As a consequence:

$$R^*(M) \leq \text{convhull} \left\{ (M_t, R_t) : t = 0, 1, \dots, K \right\}$$

where

$$M_t = \frac{tN}{K}$$
$$R_t = \frac{K}{1 - \delta} \left( 1 - \frac{M_t}{N} \right) \left( \frac{KM_t}{N} \right)^{-1}.$$

## Delivery over Heterogeneous Erasure BCs



- Erasure probability at receivers  $1, \dots, K_w$  is  $\delta_1$ , where  $K_w < K$
- Erasure probability at receivers  $K_w + 1, \dots, K$  is  $\delta_2 < \delta_1$
- Adapt the coded caching upper bound on  $R^*(M)$  to this setup!

## Rate-Memory Tradeoff for Asymmetric EBCs

- The XOR packets that are meant for at least one weak receiver require  $(1 - \delta_1)^{-1}$  times more channel inputs than over the noise-free channel
- The XOR packets that are meant for only strong receivers require  $(1 - \delta_2)^{-1}$  times more channel inputs than over the noise-free channel
- $\binom{K_s}{t+1}$  XOR packets intended for only the strong receivers, where  $t$  denotes the parameter of the coded caching scheme
- $\binom{K}{t+1} - \binom{K_s}{t+1}$  of the XOR packets intended for at least one weak receiver

## EBCs with Unequal Channel Strengths

- In this coded caching algorithm requires:

$$R^*(M) \leq \text{convhull}\{(M_t, R_t): t = 0, 1, \dots, N\}$$

where

$$M_t = \frac{tN}{K}$$

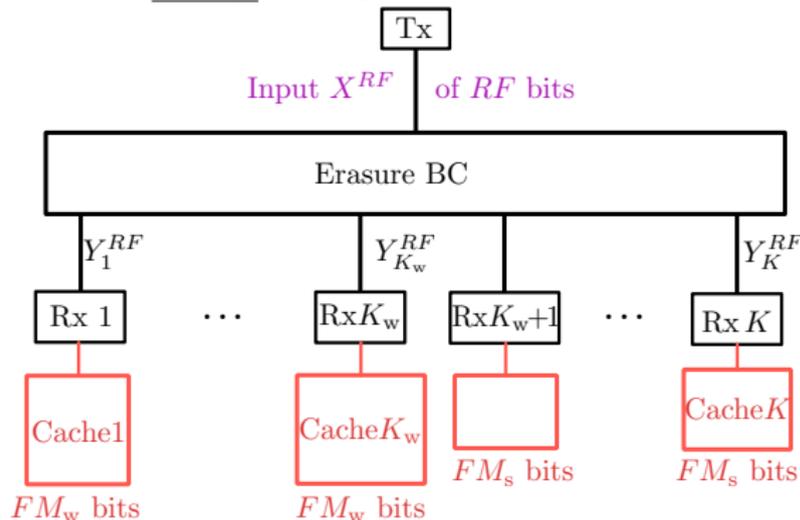
$$R_t = \frac{K}{(1 - \delta_2)} \left(1 - \frac{M_t}{N}\right) \left(1 + \frac{KM_t}{N}\right)^{-1} \\ + \underbrace{\frac{\binom{K}{t+1} - \binom{K-K_w}{t+1}}{\binom{K}{t}} \left(\frac{1}{1 - \delta_1} - \frac{1}{1 - \delta_2}\right)}_{\text{penalty caused by weak receivers}}$$

## Penalty Caused By Weak Receivers

- Weak receivers cause rate-penalty!
- As we will see, this holds only for asymmetric situations
- Penalty caused by weak receivers can partially be removed in asymmetric setups where weak receivers need less information or when they have larger cache memories
- Efficient elimination of rate penalty requires new coding approach!

## Can Cache Assignment Resolve Penalty?

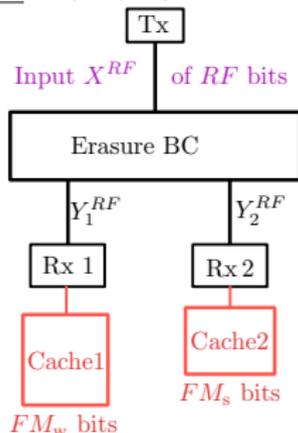
Library:  $W_1, W_2, \dots, W_N$  each of  $F$  bits



- Move part of the cache memories from strong receivers to weak receivers  
→ Idea: Help more the weak receivers to make the network more balanced
- How to exploit additional cache memories?  
→ Coded caching only works with equal cache memories at all receivers

## Two-User Example with Asymmetric Cache Memories

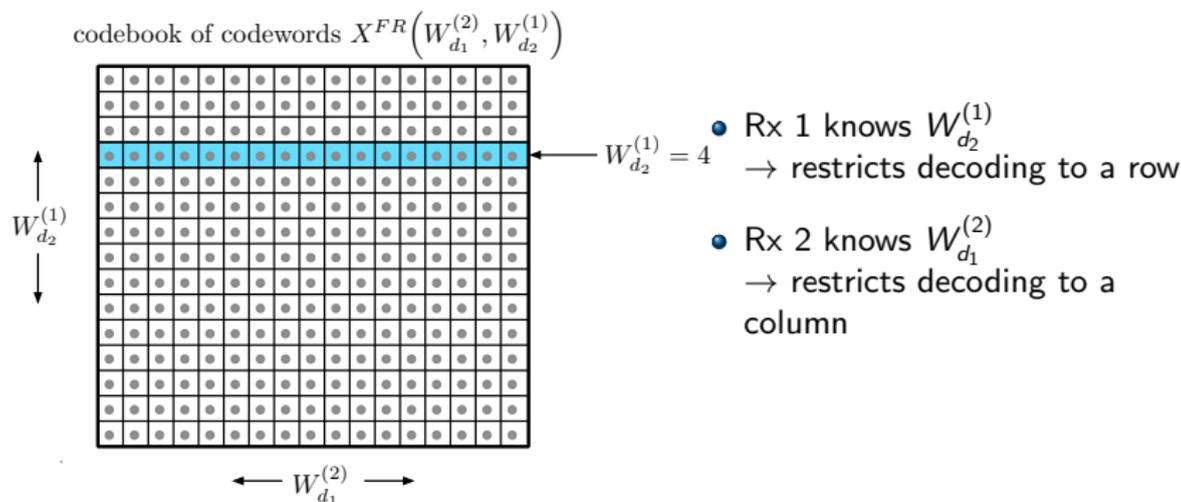
Library:  $W_1, W_2, \dots, W_N$  each of  $F$  bits



- Assign cache memory inverse proportionally to channel capacities:  
 $M_w = \frac{1-\delta_2}{2-\delta_1-\delta_2} N$  and  $M_s = \frac{1-\delta_1}{2-\delta_1-\delta_2} N$
- Split each  $W_d = (W_d^{(1)}, W_d^{(2)})$  with sizes  $\frac{1-\delta_2}{2-\delta_1-\delta_2} F$  and  $\frac{1-\delta_1}{2-\delta_1-\delta_2} F$  bits
- Placement: store  $\{W_d^{(1)}\}_{d=1}^N$  in Cache 1 and  $\{W_d^{(2)}\}_{d=1}^N$  in Cache 2

## Use a “Piggyback-Code” to Delivery $W_{d_1}^{(2)}$ and $W_{d_2}^{(1)}$

Randomly generate all codewords IID by choosing all entries IID.



- For Rx 1 to be able to decode,  $X^{FR}$  needs to be of size  $\frac{F}{2-\delta_1-\delta_2}$  bits
- For Rx 2 to be able to decode,  $X^{FR}$  needs to be of size  $\frac{F}{2-\delta_1-\delta_2}$  bits

## “Piggyback-Code” removes Penalty caused by Weak Receiver

For  $p(\text{error}) \rightarrow 0$  as  $n \rightarrow \infty$  we need  $R \geq \frac{1}{2 - \delta_1 - \delta_2}$

- Same performance as if only one of the receivers was present!
- Weaker receiver does not penalize stronger receiver!

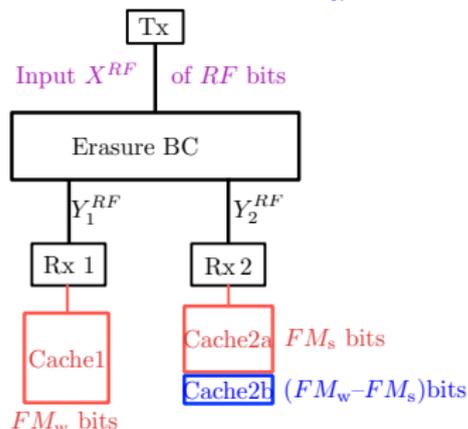
## Piggyback-Coding Extends to $K$ Receivers

- Piggyback coding extends to arbitrary number of receivers and different erasure probabilities
- Extends to general degraded broadcast channels (BC)
- Modify coded caching as follows:
  - Size of a subpart of files depends on channel strengths of the receivers caching this subpart
  - Choose  $t + 1$ -dimensional piggyback codebook for delivery communication to each subset of  $t + 1$  receivers

Size of subparts (and thus of cache memories) is chosen so that each piggyback codebook is decoded using the same time by each of the involved receivers

## Higher Resolution At Stronger Receivers

Library:  $W_1, W_2, \dots, W_N$  each of  $F$  bits  
 $T_1, T_2, \dots, T_N$  each of  $F \frac{M_w - M_s}{N}$  bits



- Let  $T_1, \dots, T_N$  be higher resolution info. required at Rx 2, not at Rx 1
- Store all  $T_1, \dots, T_N$  in Rx 2's cache memory
- Apply placement and delivery strategies described before