Performance Limits of Coded Caching under Heterogeneous Settings

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The Importance of Caching



Source: Cisco VNI Mobile, 2014

- Data traffic continues to grow at significant rates
- A major fraction (60-80%) of traffic will be generated by multimedia content, such as video
- Caching is important for reducing backhaul requirement in serving large volumes of content that multiple users are interested in



Coded Caching: Global Caching Gains



Homogeneous vs Heterogeneous Settings

- [Maddah-Ali and Niesen '14] shows that the worst-case transmission rate $K \cdot \left(1 - \frac{M}{N}\right) \cdot \frac{1}{1 + \frac{KM}{N}}$ is at most a *constant factor* (12x) away from the (information-theoretic) minimum possible
- Generalized to
 - Decentralized/probabilistic caching schemes [Maddah-Ali and Niesen `14]
 - Hierarchical caching [Karamchandani et al `14]
 - Online caching [Pedarsani et al `13]
- These studies assume a *homogeneous* setting where all files are equally important and are with the same parameters

Homogeneous vs Heterogeneous Settings

• In practice, heterogeneity arises naturally

In *homogeneous* settings, all files are cached uniformly

- In *heterogeneous* settings:
 - Should *more popular* files be cached more aggressively [Niesen and Maddah-Ali `14, Ji et al `14, Hachem et al `14]?
 - Should *larger* files be cached more aggressively?

Our Contribution

- Coded caching needs to be adapted in different ways to different aspects of heterogeneity
- Heterogeneous *popularity*:
 - Only files above a popularity *threshold* are cached
 - However, all popular files are cached *uniformly* (similar to [Ji et al '14])
 - We show constant-factor bounds that are independent of the popularity distribution
- Heterogeneous *file-sizes*
 - (Roughly) *quadratically* more content is cached for larger files
 - We show *logarithmic-factor* bounds
- While the new achievable schemes are quite intuitive, the corresponding lower bounds are more involved and reveal useful insights

Outline

- Coded Caching under Arbitrary Popularity
 Distributions
 - System Model
 - Achievable Bounds and Intuitions
 - Lower Bounds
- Coded Caching under Distinct File Sizes
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- Conclusion and Discussions



Network Model: Heterogeneous Popularity

Average-Case vs. Worst-Case

- Expected rate: $\sum_{i=1}^{N} r(W_i) P(W_i)$
- Worst-case rate [Maddah-Ali and Niesen `14]:

$$\max_{i=1,\ldots,N^K} r(W_i)$$

- Obviously: $\sum_{i=1}^{N^{K}} r(W_{i})P(W_{i}) \leq \max_{i=1,\dots,N^{K}} r(W_{i})$
- However, constant-factor results do NOT carry over from the worst case to the average case

$$\frac{\max_{i=1,\dots,N^{K}} r(W_{i})}{\max_{i=1,\dots,N^{K}} r^{*}(W_{i})} \leq c \quad \Longrightarrow \quad \frac{\sum_{i=1}^{N^{K}} r(W_{i}) P(W_{i})}{\sum_{i=1}^{N^{K}} r^{*}(W_{i}) P(W_{i})} \leq c$$

Related Work on the Average Case

- U. Niesen, and M.A. Maddah-Ali, "Coded Caching with Nonuniform Demands", arXiv:1308.0178v2 [cs.IT], Mar. 2014.
 - Divide the files into groups
 - The gap between the lower bound and the achievable (upper) bound increases with # of groups (unbounded)
- J. Hachem, N. Karamchandani and S. Diggavi, "Multi-level Coded Caching", arXiv:1404.6563 [cs.IT], Apr. 2014.
 - Popularity has *multiple levels*
 - The gap increases with # of levels (unbounded)
- M. Ji, A. Tulino, J. Llorca and G. Caire, "On the Average Performance of Caching and Coded Multicasting with Random Demands", arXiv:1402.4576v2 [cs.IT], Jul. 2014.

> **Zipf** popularity distribution $p_i \propto \frac{1}{i^{\alpha}}$

> The gap increases with $\frac{1}{\alpha-1}$ when $\alpha > 1$ (unbounded)

Our Main Results

• **Constant-factor gap** between the lower bound (R_{lb}) and the achievable (upper) bound (R_{ub}) of the expected backhual transmission rate:

 $R_{ub} \le 87R_{lb} + 2$

- The achievable bound (*R_{ub}*) is attained by a simple coded caching scheme similar to [Ji et al '14]
 - Perform coded caching only among the *most popular N*₁ *files*
 - However, all N₁ popular files are treated *uniformly*



• The key step is to show a matching lower bound

Arbitrary Popularity Distribution!

Main Intuition: An "Insensitivity" Property

The "best" worst-case rate for serving N files can be achieved by uniform caching [Maddah-Ali and Niesen '14]

The "best" worst-case rate
for serving N files can be
achieved by uniform caching
$$\frac{N}{M} - 1$$

[Maddah-Ali and Niesen '14] $\frac{M}{M} - 1$
 $K \bullet (1 - \frac{M}{N}) \bullet \frac{1}{1 + \frac{KM}{N}} \approx \frac{N}{M} (1 - \frac{M}{N}) = \frac{N}{M} - 1$
 $\frac{N}{M} = \frac{N}{M} + \frac{N}{M} = \frac{N}{M} + \frac{N}{M} + \frac{N}{M} = \frac{N}{M} + \frac{N}{M}$

whenever K >> N/M

- **Key Insight:** Beyond K=N/M, the above rate is independent of the number of users K
- Due to its global caching gain, coded caching significantly reduce the threshold for this insensitivity to arise

Main Intuition: Average Case

- Consider the following scheme:
 - Only perform coded caching am most "popular" files 1 to N_1
- The average transmission rate for the "popular" files will be upperbounded by the worst-case rate:

- Only perform coded caching among
most "popular" files 1 to
$$N_1$$

The average transmission rate for
the "popular" files will be upper-
bounded by the worst-case rate:
 $K' \bullet (1 - \frac{M}{N_1}) \bullet \frac{1}{1 + \frac{K'M}{M}} \approx \frac{N_1}{M} (1 - \frac{M}{N_1}) = \frac{N_1}{M} - 1$?

Popularity

whenever $K' >> N_1/M$

If these files are indeed very popular, K' will be large. Thus, the expected rate will likely be close to this upper bound $\frac{N_1}{M_1} - 1$ M

Once a file is "popular", its popularity does not matter!

Achievable Bound

An achievable rate:



Proposition 1: Assume $M \ge 2$. There exists an achievable scheme whose average transmission rate satisfies:

$$R(K,\mathcal{F},\mathcal{P}) \leq R_{ub} = \left[\frac{N_1}{M} - 1\right]_+ + \sum_{i > N_1} K p_i$$

where N_1 satisfies $p_{N_1} \ge \frac{1}{KM}$ and $p_{N_1+1} < \frac{1}{KM}$.

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Lower Bound: Statement



 For all possible coded caching schemes, the average transmission rate is lower bounded by

$$R(K, \mathcal{F}, \mathcal{P}) \ge R_{lb} = \max\{\frac{1}{29} \begin{bmatrix} N_1 \\ M \end{bmatrix} - 1 \end{bmatrix}_{+}, \frac{1}{58} \begin{bmatrix} \sum_{i > N_1} Kp_i - 2 \end{bmatrix}_{+}\}$$

Lower bound for
serving "popular files" Lower bound for
serving "popular files" $N_1 + 1$ to N

Lower Bound: Challenges

Need to show:

- For popular files, popularity does not matter
 - Reduce to uniform popularity 1/KM and use stochastic dominance



- With uniform popularity for all popular files, their worst-case rate and average-case rate are on the same order [Niesen and Maddah-Ali `14]
- For unpopular files, it is a good idea not to use any caching

Popular Files: Popularity Does Not Matter

$$\mathcal{F}, \mathcal{P} \xrightarrow{F_1, p_1} \cdots \xrightarrow{F_{N_1}, p_{N_1}} \xrightarrow{F_{N_1+1}, p_{N_1+1}} \cdots \xrightarrow{F_N, p_N} R(K, \mathcal{F}, \mathcal{P})$$
Remove all unpopular files
$$V$$

$$\mathcal{F}_1, \mathcal{P}_1 \xrightarrow{F_1, p_1} \cdots \xrightarrow{F_{N_1}, p_{N_1}} \emptyset \text{ (empty file), } 1 - \sum_{i=1}^{N_1} p_i$$
Reduce all popularity to the lowest value p_{N_1}

$$V$$

$$\mathcal{F}_1, \mathcal{P}_2 \xrightarrow{F_1, p_{N_1}} \cdots \xrightarrow{F_{N_1}, p_{N_1}} \emptyset \text{ (empty file), } 1 - N_1 p_{N_1}$$
R(K, $\mathcal{F}_1, \mathcal{P}_2$)
System 2

System 2 ($K, \mathcal{F}_1, \mathcal{P}_2$): Worst-case vs. Average-Case

- Each of the K user request one of the N_1 popular files with equal probability $p_{N_1} \ge \frac{1}{KM}$
- The average number of users requesting popular files is $KN_1p_{N_1} \ge \frac{N_1}{M}$
- **Property 1**: with reasonable probability, the *number of users* K_r requesting popular files is no smaller than $\left|\frac{N_1}{M}\right|$. More precisely,

$$P\left(K_r \ge \left\lfloor \frac{N_1}{M} \right\rfloor\right) \ge 0.5$$

• **Property 2:** with reasonable probability, the number of distinct files K_d requested is no smaller than $0.5K_r$. More precisely

3

$$P\left(K_{d} \ge \left|\frac{1}{2}K_{r}\right| | K_{r}\right) \ge 0.56$$

$$P\left(K_{d} \ge \left|\frac{1}{2}\left|\frac{N_{1}}{M}\right|\right)\right) \ge 0.28$$
denoted as *K*

Popular Files: Further Reduction from System 2





Unpopular Files: Reduction to System 2

$$\begin{split} \hline \emptyset, 1 - \sum_{i=N_{1}+1}^{N} p_{i} & F_{N_{1}+1}, p_{N_{1}+1} & \cdots & V_{k}, p_{i} + p_{j} & \cdots & F_{N}, p_{N} \\ \hline \text{Merge files until the sum popularity is just above 1/KM} & & & & & & & \\ \hline \emptyset, 1 - \sum_{i=1}^{N_{2}} p_{i} & V_{1}, v_{1} & V_{2}, v_{2} & \cdots & V_{N_{2}}, v_{N_{2}} \\ \hline \emptyset, 1 - \sum_{i=1}^{N_{2}} p_{i} & V_{1}, v_{1} & V_{2}, v_{2} & \cdots & V_{N_{2}}, v_{N_{2}} \\ \hline \vdots \\ R_{KM} > v_{i} \geq \frac{1}{KM}, & & & & & & \\ \hline N_{2} \geq (\sum_{i>N_{1}} p_{i})KM/2 & & & & & & \\ \hline \emptyset, 1 - \frac{N_{2}}{KM} & V_{1}, \frac{1}{KM} & V_{1}, \frac{1}{KM} & \cdots & V_{N_{2}}, \frac{1}{KM} \\ \hline \emptyset, 1 - \frac{N_{2}}{KM} & V_{1}, \frac{1}{KM} & V_{1}, \frac{1}{KM} & \cdots & V_{N_{2}}, \frac{1}{KM} \\ \hline R(K, \mathcal{F}_{4}, \mathcal{P}_{5}) & & & \\ \hline R(K, \mathcal{F}, \mathcal{P}) \geq \frac{1}{29} \left[\frac{N_{2}}{M} - 1 \right]_{+} \geq \frac{1}{29} \left[\frac{\Sigma_{i>N_{1}} Kp_{i}}{2} - 1 \right]_{+} \\ \end{array}$$

Constant Factor

• We have shown the lower bound:

$$R(K,\mathcal{F},\mathcal{P}) \ge R_{lb} = \max\{\frac{1}{29} \left[\frac{N_1}{M} - 1\right]_+, \frac{1}{58} \left[\sum_{i>N_1} Kp_i - 2\right]_+\}$$

• Recall the achievable bound

$$R(K,\mathcal{F},\mathcal{P}) \le R_{ub} = \left[\frac{N_1}{M} - 1\right]_+ + \sum_{i>N_1} Kp_i$$

• Constant-factor:

 $R_{ub} \leq 87R_{lb} + 2$

Numerical Comparison

- **LFU** [Lee et al '01]:
 - Cache the *M* most popular contents (No coding)
- Uniform-caching [Maddah-Ali and Niesen '14]:
 - > Randomly cache $\frac{M}{N}$ portion of every content, regardless of popularity
- **Group-caching** [Niesen and Maddah-Ali, '14]:
 - Divide the files into groups with similar popularity; perform coded caching within each group
 - Include an additional cache-allocation optimization
- RLFU (Random LFU) [Ji et al '14]:
 - Assume Zipf popularity distribution; perform coded caching among the most popular files 1 to N_1
 - > Numerically optimize N_1 based on some upper bound
- For uniform-caching, group-caching and RLFU, we plot the upper bound on the average transmission rate



Note: For RLFU, Group-Caching, Uniform-Caching, we plot the upper bound

Non-Zipf Distribution

• Zipf-Mandelbrot law distribution



Summary: Arbitrary Popularity Distributions

- We study the expected transmission rate of coded caching *under arbitrary popularity distributions*
- We obtain achievable bounds that differ from the information-theoretic lower bound by at most a *constant factor* (except for a small additive term)
- *Threshold* Structure:
 - Perform coded caching only among popular files

$$\geq p_{N_1} \approx \frac{1}{KM}$$

- However, all popular files are cached uniformly
- Similar to [Ji et al '14], but use a different N_1

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- Network Model: Distinct File Sizes
- Server with a broadcast channel
- K users: cache size M
- *N* files: $\mathbb{F} = \{\mathcal{F}_1, \dots, \mathcal{F}_N\}$ **File-size (non increasing)**: $|\mathcal{F}_i| = F_i$ $F_i \ge F_j$, if $i \le j$
- Request pattern: $W_i = \{f_{i1}, \dots, f_{iK}\}, f_{ik} \in \mathbb{F}$ Rate for W_i is $r(W_i)$
- Worst-case rate:

 $R = \max_i r(W_i)$



Power-of-2 Simplification

- File-sizes differ by power-of-2 factors
- *l*-th type: $F_l = F_1/2^{l-1}$
 - N_l files of type l



- T distinct types
 - The total number of files $\sum_{l=1}^{l} N_l = N$
- $\bar{T} = \min\{T, \log_2 K\}$
 - Files of type $l > \overline{T}$ can be virtually neglected
 - Their sizes $\leq F_1/K$

Our Main Results

• Logarithmic-factor gap between the lower bound (R_{LB2}) and the achievable (upper) bound (R_{UB2}) for the worst-case transmission rate:

 $R_{UB2} \leq 32 \log_2 K \cdot R_{LB2} + 22$

- The achievable bound (R_{UB2}) is attained by caching larger files more aggressively
 - Quadratically more content is cached for larger files
- The key step is to show a tighter lower bound, which involves careful use of entropy inequalities



Recall "Insensitivity" under Unit File-size

 The worst-case rate with uniform file size of 1 [Maddah-Ali & Niesen `14] is given by

$$K \cdot \left(1 - \frac{M}{N}\right) \cdot \frac{1}{1 + \frac{KM}{N}} \approx \frac{N}{M} - 1 \approx \frac{N}{M}, \text{ when } K >> \frac{N}{M} \text{ and } M << N$$

• Each user caches every "bit" of each file with probability $q = \frac{M}{N}$

➢ Worst-case rate ≈
$$\frac{N}{M} = \frac{1}{q}$$

• When the uniform file-size is *F* , these numbers become:

➤ Caching probability
$$q = \frac{M}{NF}$$
,

Worst-case rate:

$$KF \cdot \left(1 - \frac{M}{NF}\right) \cdot \frac{1}{1 + \frac{KM}{NF}} \approx \frac{NF^2}{M} = \frac{F}{q}$$
, when $q \ll 1$ and $K \gg 1/q$

Two Achievable Schemes (UB1 vs UB2)

Achievable Scheme 1 (UB1)

- All files are cached with an equal probability q
 - Linearly more content is cached for larger files
- Cache constraint: $q \sum_{l=1}^{\overline{T}} N_l F_l = M \text{ with}$ $\overline{T} = \min(T, \log_2 K)$



Achievable Scheme 2 (UB2)

- > Let the caching probability $q_l = F_l/c$
 - Quadratically more content is cached for larger files
- Cache constraint:

$$\sum_{l=1}^{\bar{T}} q_l N_l F_l = M \implies \sum_{l=1}^{\bar{T}} N_l F_l^2 = cM$$
$$R_{UB2} \cdot \sum_{l=1}^{\bar{T}} \frac{F_l}{q_l} + KF_{\bar{T}+1}$$
$$= \bar{T}c + KF_{\bar{T}+1}$$
$$\cdot (\bar{T}+1) \frac{\sum_{l=1}^{\bar{T}} N_l F_l^2}{M}$$

Two Achievable Schemes (UB1 vs UB2)



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Lower Bound: Uniform File-size



In order to serve all request patterns, we must have [Maddah-Ali & Niesen `14]

$$\frac{2M}{F}R^*(\mathbf{F}) + \frac{NF}{2M}M \ge NF \implies R^*(\mathbf{F}) \ge \frac{NF^2}{4M} = \frac{F}{4q}$$

Lower Bound: First Try



In order to serve all request patterns, we must have

$$\sum_{l \in \Phi} \frac{\left|\sum_{l \in \Phi} N_{l}\right|}{L} \cdot R^{*}(\mathbb{F}) + L \cdot M \geq \sum_{l \in \Phi} N_{l} F_{l}$$

$$Maximizing \text{ over } L \text{ and } \Phi$$

$$R^{*}(\mathbb{F}) \geq \max_{\Phi} \frac{\left(\sum_{l \in \Phi} N_{l} F_{l}\right)^{2}}{4M \sum_{l \in \Phi} N_{l}}$$

$$(LB1)$$

LB1 vs UB2

Lower Bound 1 (LB1) $R \ge R_{LB1} = \max_{\Phi} \frac{(\sum_{l \in \Phi} N_l F_l)^2}{4M \sum_{l \in \Phi} N_l}$

 $\blacktriangleright \quad \text{Consider } N_{l+1} = 4N_l$ $R_{LB1} \cdot \quad \frac{N_1 F_1^2}{4M}$

LB1 fails to account for heterogeneous caching probabilities! Achievable Bound 2 (UB2)

$$R \cdot R_{UB2} = (\bar{T}+1) \frac{\sum_{l=1}^{\bar{T}} N_l F_l^2}{M}$$



An Improved Lower Bound (LB2) Proposition 3: Under

• Assumption 1: $\frac{2M}{F_1}$ and $\frac{N_l F_l}{2M}$ are integers for all $1 \le l \le \overline{T}$

• Assumption 2:
$$\sum_{l=1}^{\overline{T}} \frac{N_l F_l}{2M} \le K$$

We must have



Intuition for LB2 (Two types: $F_2 = F_1/2$) $R^*(\mathbb{F}) \ge \sum_{l=1}^{\bar{T}} \frac{N_l F_l^2}{4M} = \frac{N_1 F_1^2}{4M} + \frac{N_2 F_2^2}{4M}$ (LB2) > Assumption 2 ensures that $s_1 + s_2 = \frac{N_1 F_1}{2M} + \frac{N_2 F_2}{2M} \leq K$ $s_1 = \frac{N_1 F_1}{2M}$ users (U₁) $s_2 = \frac{N_2 F_2}{2M}$ users (U_2) $\frac{N_1}{s_1} = \frac{2M}{F_1}$ N_1 files of type 1 (F_1) N_2 files of type 2 (F_2) $\frac{N_1}{s_1} = \frac{2M}{F_1} \xrightarrow{\longrightarrow}$ $R^* (\mathbf{F})^{\text{files}} \mathcal{N}_1 \mathcal{F}_1^2$ patterns



$$\frac{I \vee I}{s_1} R^*(\mathbb{F}) \geq H(\mathcal{R}_{\mathcal{D}_1}) + H(\mathcal{R}_{\mathcal{D}_2})
= H(\mathcal{R}_{\mathcal{D}_1} | \mathbb{F}_1) + I(\mathcal{R}_{\mathcal{D}_1}; \mathbb{F}_1) + H(\mathcal{R}_{\mathcal{D}_2} | \mathbb{F}_1) + I(\mathcal{R}_{\mathcal{D}_2}; \mathbb{F}_1)
\geq H(\mathcal{R}_{\mathcal{D}_1} \cup \mathcal{R}_{\mathcal{D}_2} | \mathbb{F}_1) + I(\mathcal{R}_{\mathcal{D}_1}; \mathbb{F}_1) + I(\mathcal{R}_{\mathcal{D}_2}; \mathbb{F}_1)
= I(\mathcal{R}_{\mathcal{D}_1} \cup \mathcal{R}_{\mathcal{D}_2}; \mathbb{F}_2 | \mathbb{F}_1) + I(\mathcal{R}_{\mathcal{D}_1}; \mathbb{F}_1) + I(\mathcal{R}_{\mathcal{D}_2}; \mathbb{F}_1).$$

$$I(\mathcal{R}_{\mathcal{D}_1}; \mathbb{F}_1) \geq \frac{N_1 F_1}{2}, I(\mathcal{R}_{\mathcal{D}_2}; \mathbb{F}_1) \geq \frac{N_1 F_1}{2}$$

$$I(\mathcal{R}_{\mathcal{D}_1} \cup \mathcal{R}_{\mathcal{D}_2}; \mathbb{F}_2 | \mathbb{F}_1) \ge \frac{N_2 F_2}{2}$$



• From
$$\frac{2N_1}{s_1}R^*(\mathbb{F}) \ge N_1F_1 + \frac{N_2F_2}{2}$$

 $\frac{4M}{F_1}R^*(\mathbb{F}) \ge N_1F_1 + \frac{N_2F_2}{2}$
 $R^*(\mathbb{F}) \ge \frac{N_1F_1^2}{4M} + \frac{N_2F_2F_1}{8M}$
 $R^*(\mathbb{F}) \ge \frac{N_1F_1^2}{4M} + \frac{N_2F_2^2}{4M}$ (LB2)

Beyond Power-of-2

- Original file-set: $\mathbb{F} = \{\mathcal{F}_1, \dots, \mathcal{F}_N\}, |\mathcal{F}_i| = F_i$
- Upper-quantized version

$$\mathbb{F}^{UB} = \left\{ \mathcal{F}_i^{UB} | F_i^{UB} = F_1 \cdot 2^{-\left\lfloor \log_2 \frac{F_1}{F_i} \right\rfloor} \right\}$$

Lower-quantized version

$$\mathbb{F}^{LB} = \left\{ \mathcal{F}_i^{LB} | F_i^{LB} = F_1 \cdot 2^{-\left\lfloor \log_2 \frac{F_1}{F_i} \right\rfloor - 1} \right\}$$

- $R^*(\mathbb{F}^{LB}) \leq R^*(\mathbb{F}) \leq R^*(\mathbb{F}^{UB})$
- Under certain conditions (Assumption 2), R*(F^{LB}) is in a constant gap with R*(F^{UB})

Т

Index



K/log² K

logK

logk

logK

With Assumptions 1 & 2	
	• • • • •
	Number of types

Rate	LB2/LB1	Gain	UB1/UB2	Gain
2 types	4/3.6	11%	15.3757/13.2523	16%
3 types	5/3.8462	30%	21.2593/16.2196	31%

General Result (without Assumptions 1 and 2): $R_{UB2} \leq (32\log_2 K + 22)R_{LB2}$

Comparisons

Conclusions

- Heterogeneous *popularity*:
 - A simple *threshold-based* policy (similar to [Ji et al '14]): files above a popularity threshold are cached uniformly
 - We show *constant-factor* gap that is *independent* of the popularity distribution
- Heterogeneous *file-sizes*
 - **Quadratically** more content is cached for larger files
 - We show *logarithmic-factor* gap
- While the new achievable schemes are quite intuitive, the corresponding *lower bounds* more involved and reveal useful insights

Potential Future Directions

- Refinements:
 - Combining heterogeneous popularity and file-sizes?
 - Reduce the logarithmic factor?
- Other aspects of heterogeneity:
 - Heterogeneous cache size?
 - The role of cache location?
- Wireless environments (HetNet):
 - Is coded caching still helpful?
- Practical considerations:
 - Low-complexity coding/transmission schemes
 - Real-time transmissions [Niesen and Maddah-Ali `15]

Thank you!

- J. Zhang, X. Lin and X. Wang, "Coded Caching under Arbitrary Popularity Distributions," in *ITA Workshop*, UCSD, February 2015.
- J. Zhang, X. Lin, C.-C. Wang and X. Wang, "Coded Caching for Files with Distinct File Sizes," in ISIT, June 2015 (to appear).