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Energy Efficiency Optimization In Cell-free Massive MIMO With Normalized Conjugate Beamforming

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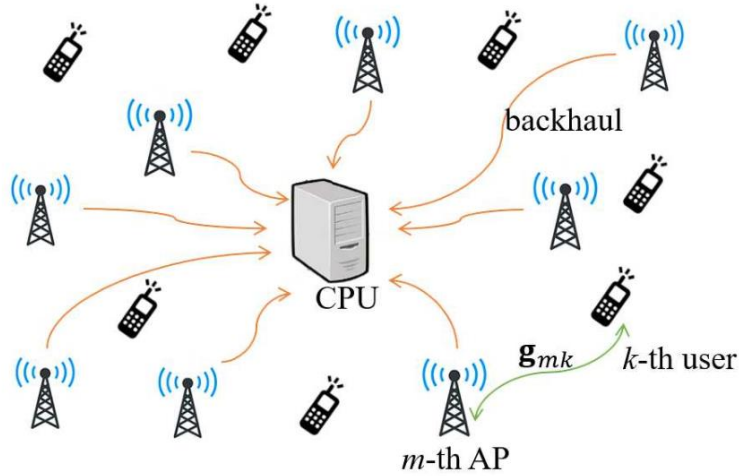
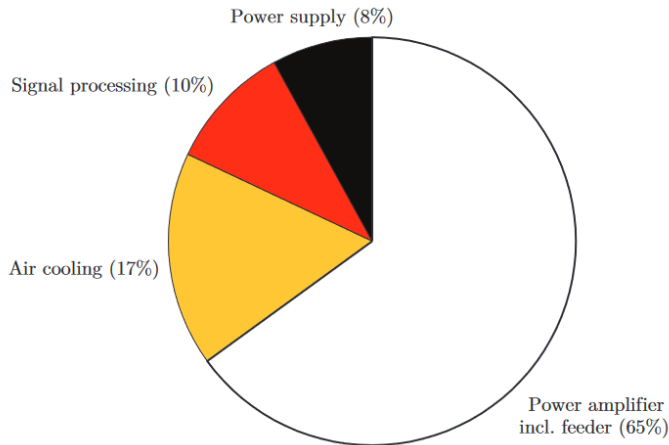


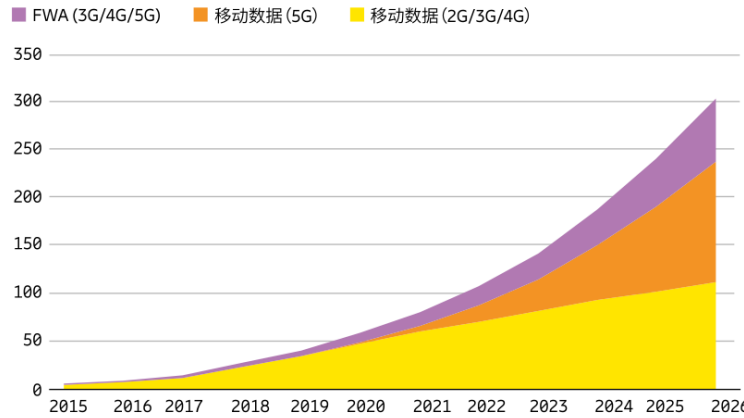
Fig. 1. Illustration of the cell free massive MIMO downlink systems.

Cell free massive MIMO

- Embrace the user-centric idea
- Eliminate the concept of cell boundaries
- Coverage enhancement
- Increased flexibility



Percentage of power consumed by different components



Growing demand for mobile traffic

Challenging issues about energy consumption and environmental pollution



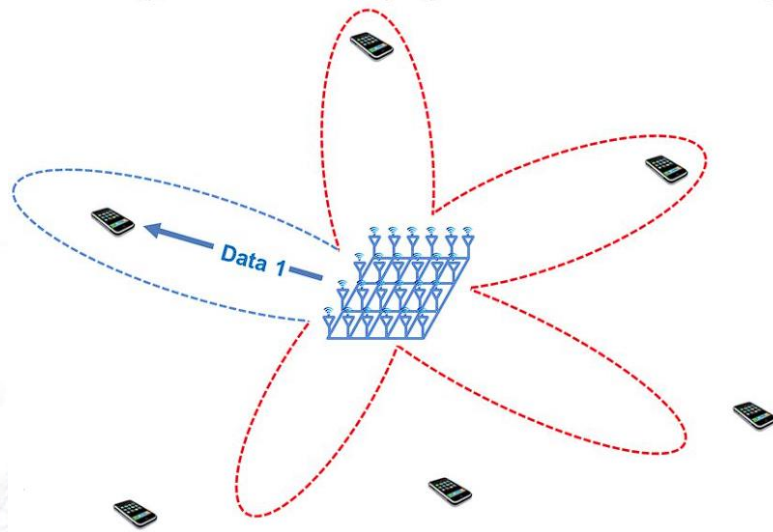
Energy efficiency optimization problem

Precoding

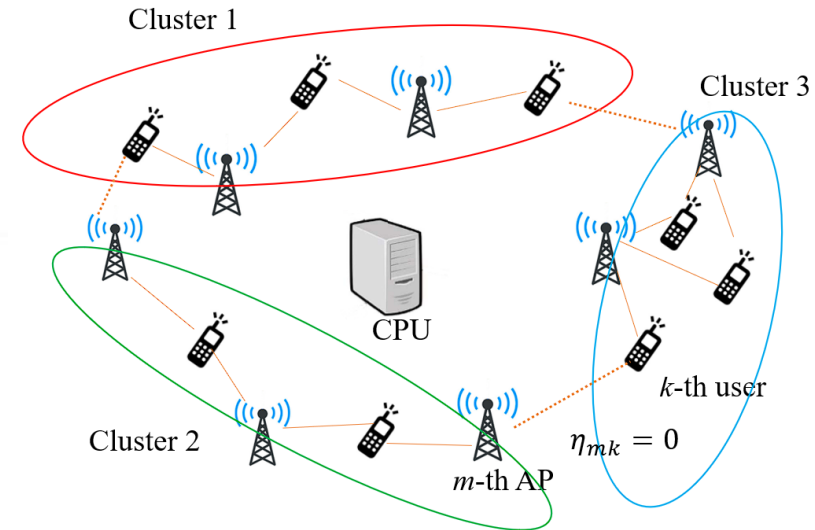
- Conjugate beamforming \Rightarrow high interference
- Zero forcing \Rightarrow complex matrix inversion
- Normalized conjugate beamforming

AP selection

- Reduce extra energy consumption
- Suitable for the practical implementation
- Avoid substantial pilot contamination



Conjugate beamforming: directional transmission of signals for users



Cell free massive MIMO systems with AP selection

Downlink energy efficiency (EE)

$$E_e(\{\eta_{mk}\}) = \frac{BS_e(\{\eta_{mk}\})}{P_{total}} = \frac{B \sum_{k=1}^K S_{ek}(\{\eta_{mk}\})}{P_{total}}$$



How much energy it takes to reliably transmit a certain amount of information.

- Each AP with multiple antennas
- Presence of pilot contamination
- Channel estimation errors

$$SINR_k(\{\eta_{mk}\}) = \frac{\rho_d \Gamma_N^2 \left(\sum_{m=1}^M \sqrt{\eta_{mk} \alpha_{mk}} \right)^2}{1 + \rho_d \sum_{k'=1}^K \sum_{m=1}^M \eta_{mk'} \beta_{mk} + (N - 1 - \Gamma_N^2) \rho_d \sum_{m=1}^M \eta_{mk} \alpha_{mk} + \rho_d \sum_{k' \neq k}^K \gamma_{kk'} |\psi_k^H \psi_{k'}|^2}$$

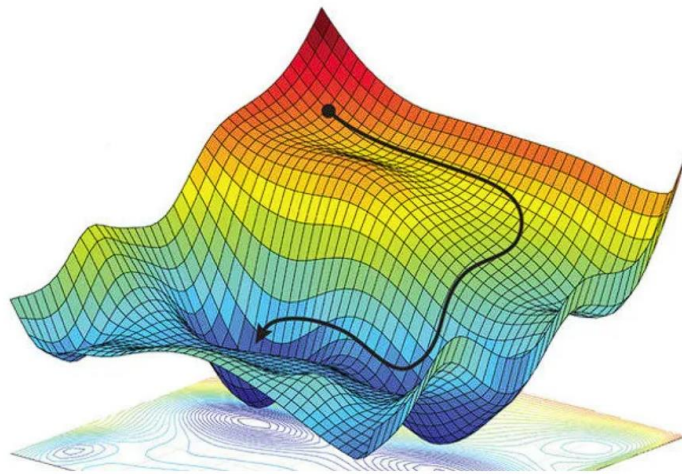
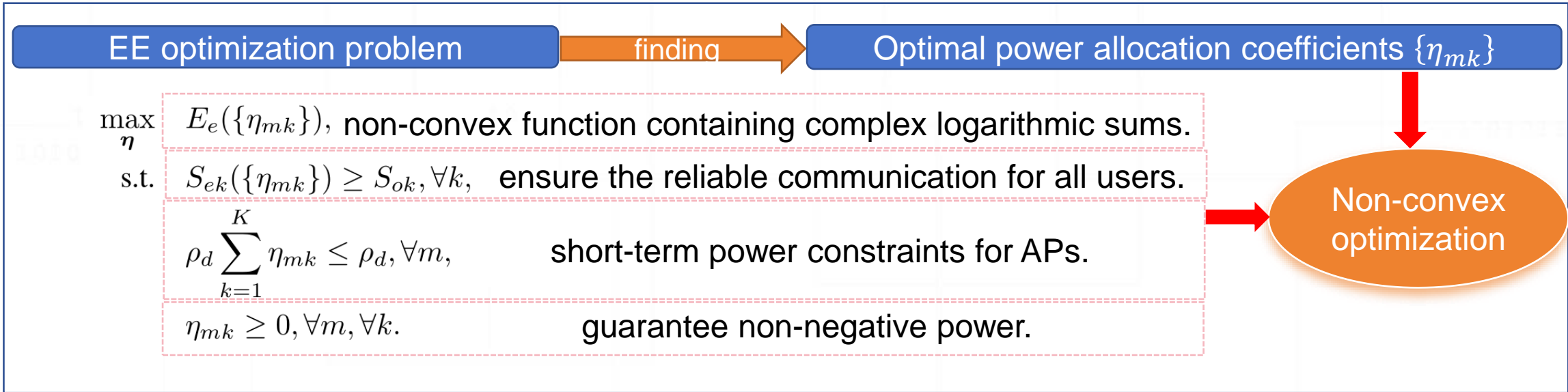
spectral efficiency

$$S_{ek}(\{\eta_{mk}\}) = \log_2 (1 + SINR_k(\{\eta_{mk}\}))$$

$$P_{total} = \rho_d N_0 \sum_{m=1}^M \frac{1}{\mu_m} \left(\sum_{k=1}^K \eta_{mk} \right) + P_{fix} + BS_e(\{\eta_{mk}\}) \sum_{m=1}^M P_{bt,m}$$

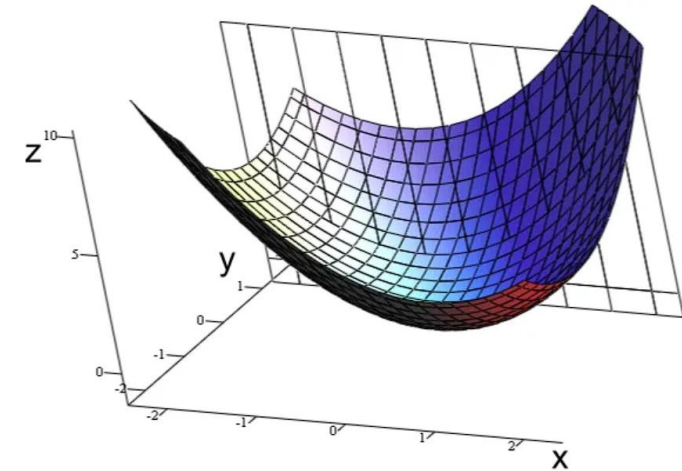
power consumption





non-convex optimization

How to transform it into a tractable form?



convex optimization

Main idea non-convex optimization problem \longrightarrow sequential convex approximation (SCA)

(1) Partial approximation of the objective function

$$SINR_k(\{\eta_{mk}\}) = \frac{\rho_d \Gamma_N^2 \left(\sum_{m=1}^M \sqrt{\eta_{mk} \alpha_{mk}} \right)^2}{1 + \rho_d \sum_{k'=1}^K \sum_{m=1}^M \eta_{mk'} \beta_{mk} + (N - 1 - \Gamma_N^2) \rho_d \sum_{m=1}^M \eta_{mk} \alpha_{mk} + \rho_d \sum_{k' \neq k}^K \boxed{\gamma_{kk'}} |\psi_k^H \psi_{k'}|^2}$$

user interference caused by the pilot contamination

$$\begin{aligned} \gamma_{kk'} \triangleq & (N - 1) \sum_{m=1}^M \eta_{mk'} \alpha_{mk'} \frac{\beta_{mk}^2}{\beta_{mk'}^2} \\ & + \Gamma_N^2 \sum_{m=1}^M \sum_{n \neq m}^M \sqrt{\eta_{mk'} \eta_{nk'} \alpha_{mk'} \alpha_{nk'}} \frac{\beta_{mk} \beta_{nk}}{\beta_{mk'} \beta_{nk'}} \end{aligned}$$

$$\gamma_{kk'} \approx (N - 1) \sum_{m=1}^M \eta_{mk'} \alpha_{mk'} \frac{\beta_{mk}^2}{\beta_{mk'}^2} + \Gamma_N^2 \left(\sum_{m=1}^M \sqrt{\eta_{mk'} \alpha_{mk'} \frac{\beta_{mk}}{\beta_{mk'}}} \right)^2$$

Dual summation: complex and intractable.

First-order approximation:

- High approximate accuracy.
- The objective function achieves a more tractable form.

(2) Transforming optimization variables and functions

Power allocation coefficients $\{\eta_{mk}\}$ \rightarrow Denote $\{c_{mk}\} = \{\sqrt{\eta_{mk}}\}$: promote the quadratic convex transformation

Original problem	Optimization problem after introducing auxiliary variable $\{t_k\}$
$\max_{\eta} \frac{BS_e(\{\eta_{mk}\})}{\rho_d N_0 \sum_{m=1}^M \frac{1}{\mu_m} \left(\sum_{k=1}^K \eta_{mk} \right) + P_{fix}},$ <p>s.t. $S_{ek}(\{\eta_{mk}\}) \geq S_{ok}, \forall k,$</p> $\rho_d \sum_{k=1}^K \eta_{mk} \leq \rho_d, \forall m,$ $\eta_{mk} \geq 0, \forall m, \forall k.$	$\max_{\mathbf{c}, \mathbf{t}} \frac{B \sum_{k=1}^K t_k}{\ln 2}, \quad \text{Simpler objective function}$ <p>s.t. $\frac{\ln(1 + SINR_k)}{P_{abs}} \geq t_k, \forall k,$ non-convex constraint</p> $S_{ek}(\{c_{mk}\}) \geq S_{ok}, \forall k,$ $\sum_{k=1}^K c_{mk}^2 \leq 1, \forall m,$ $c_{mk} \geq 0, \forall m, \forall k. \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \text{convex constraints}$

Second-order cone (SOC) constraint:

$$(\boldsymbol{\alpha}_k^T \mathbf{c}_k)^2 \geq (2^{S_{ok}} - 1) \left(\frac{1}{\rho_d \Gamma_N^2} + \frac{\sum_{k'=1}^K \|\boldsymbol{\beta}_k \mathbf{c}_{k'}\|_2^2}{\Gamma_N^2} + \frac{(N-1 - \Gamma_N^2)}{\Gamma_N^2} \|\boldsymbol{\alpha}_k \cdot \mathbf{c}_k\|_2^2 + \sum_{k' \neq k}^K (\boldsymbol{\xi}_{kk'}^T \mathbf{c}_{k'})^2 \zeta_{kk'} + \frac{(N-1)}{\Gamma_N^2} \sum_{k' \neq k}^K \|\boldsymbol{\xi}_{kk'} \cdot \mathbf{c}_{k'}\|_2^2 \zeta_{kk'} \right), \forall k.$$

(3) Approximate non-convex constraint

➤ Original form

$$\frac{\ln(1 + SINR_k)}{P_{abs}} \geq t_k, \forall k,$$

$$\left\{ \begin{aligned} SINR_k(\{c_{mk}\}) &= \frac{\Gamma_N^2(\alpha_k^T \mathbf{c}_k)^2}{\frac{1}{\rho_d} + \sum_{k'=1}^K \|\beta_k \mathbf{c}_{k'}\|_2^2 + (N-1 - \Gamma_N^2) \|\alpha_k \cdot \mathbf{c}_k\|_2^2 + \sum_{k' \neq k}^K \left((N-1) \|\xi_{kk'} \cdot \mathbf{c}_{k'}\|_2^2 + \Gamma_N^2 (\xi_{kk'}^T \mathbf{c}_{k'})^2 \right) \zeta_{kk'}} \\ P_{abs} &\triangleq \rho_d N_0 \sum_{m=1}^M \frac{1}{\mu_m} \left(\sum_{k=1}^K c_{mk}^2 \right) + P_{fix} \in \mathbb{R}, \end{aligned} \right.$$

➤ Step 1

$$\frac{\ln(1+x)}{t} \geq a - \frac{b}{x} - d \cdot t, \quad \forall x > 0, t > 0.$$

$$\frac{\ln(1 + SINR_k)}{P_{abs}} \geq a_k^n - d_k^n P_{abs} - \frac{b_k^n}{\rho_d \Gamma_N^2 (\alpha_k^T \mathbf{c}_k)^2} \frac{\sum_{k'=1}^K \left[\|\beta_k \mathbf{c}_{k'}\|_2^2 + (N-1 - \Gamma_N^2) \|\alpha_k \cdot \mathbf{c}_k\|_2^2 \right]}{\Gamma_N^2 (\alpha_k^T \mathbf{c}_k)^2 (b_k^n)^{-1}} - \frac{\sum_{k' \neq k}^K \left((N-1) \|\xi_{kk'} \cdot \mathbf{c}_{k'}\|_2^2 + \Gamma_N^2 (\xi_{kk'}^T \mathbf{c}_{k'})^2 \right) \zeta_{kk'}}{\Gamma_N^2 (\alpha_k^T \mathbf{c}_k)^2 (b_k^n)^{-1}}$$

- ✦ Separate fractions containing logarithmic and quadratic term
- ✦ The right-hand-side (RHS) is still a non-convex function

➤ Step 2

$$\frac{x^2}{t} \geq 2 \frac{\hat{x}x}{\hat{t}} - \frac{\hat{x}^2}{\hat{t}^2} t, \quad \forall x > 0, \hat{x} > 0, t > 0, \hat{t} > 0$$

- ✦ The quadratic term become the first-order term.
- ✦ Transform denominator term with quadratic optimization variables.

➤ Step 3

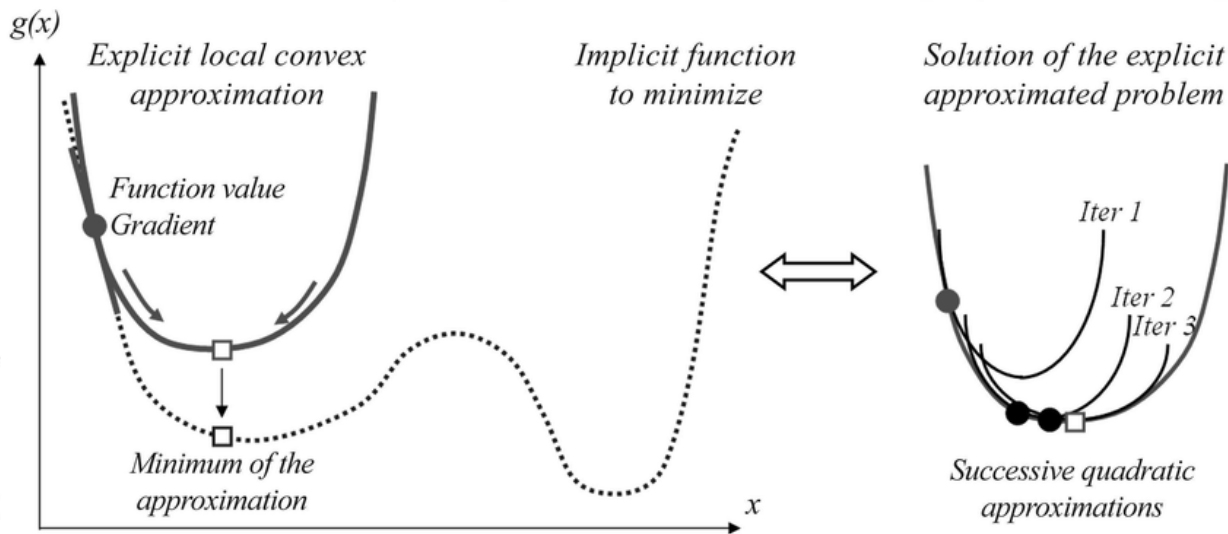
$$x^2 \geq 2\hat{x}x - \hat{x}^2, \quad \forall x \geq 0, \hat{x} \geq 0, 2x \geq \hat{x}$$

$$\begin{aligned} (\alpha_k^T \mathbf{c}_k)^2 &\geq 2(\alpha_k^T \mathbf{c}_k^n)(\alpha_k^T \mathbf{c}_k) - (\alpha_k^T \mathbf{c}_k^n)^2, \forall k, \\ 2c_{mk} &\geq c_{mk}^n, \quad \forall m, \forall k. \end{aligned}$$

- ✦ more computationally efficient formulation.

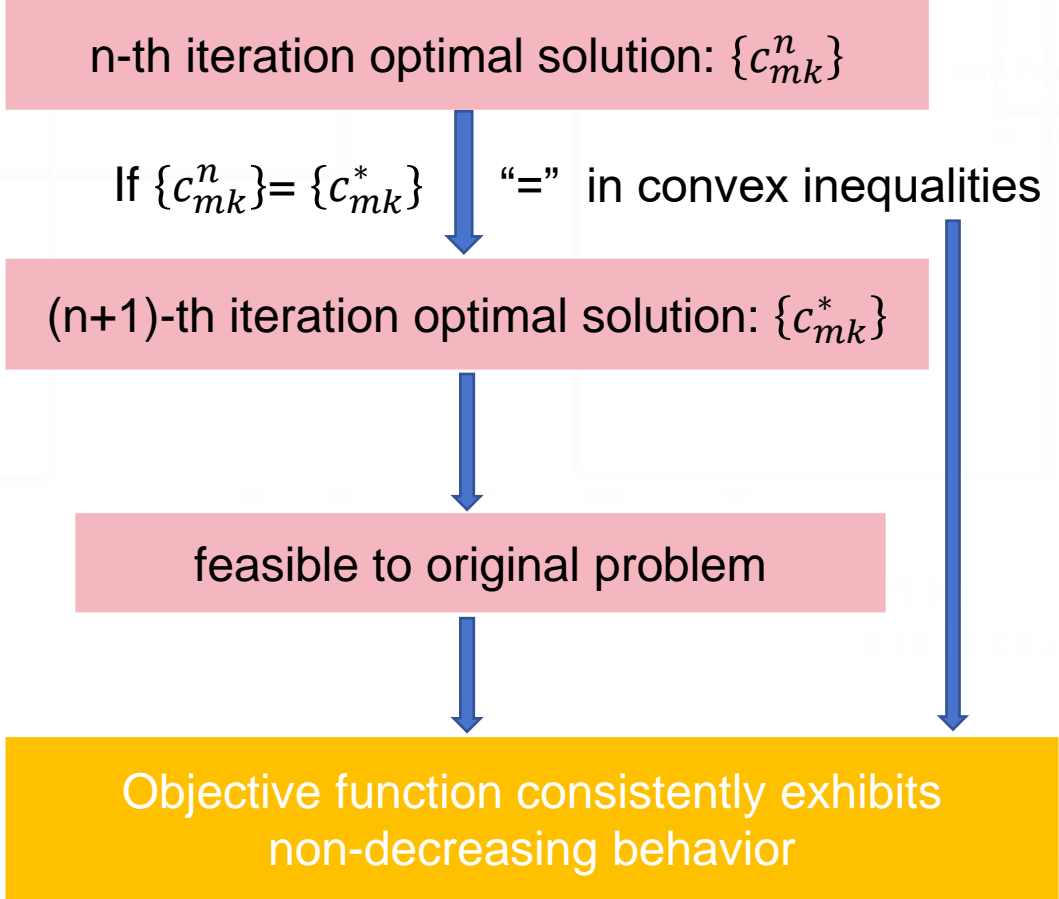
(4) Sequential convex approximation

Convex constraints after approximate transformation



- Iterative solve a series of accessible SOCP problems.
- The initial feasible solution is easy to obtain.

(5) Convergence analysis



Motivation

- Shortage of pilot resources
- Additional backhaul overhead
- More severe interference between users with similar channels

- Initialization: set the number of clusters as $L = \left\lceil \frac{k}{\tau_u} \right\rceil$
- User clustering: clustering based on the similarity of $\{\beta_{mk}\}$
- Centroid position update: whether the clustering result is stable
- Modify cluster size: ensure the cluster use orthogonal pilots
- AP selection: services for clusters with the best channel quality

Algorithm 1 The Proposed Power Allocation Scheme For EE Optimization With NCB

Input: $S_{ok}, \rho_d, N, \{\alpha_{mk}\}, \{\beta_{mk}\}, N_I$

Output: power allocation coefficients $\{\eta_{mk}\} = \{c_{mk}^2\}$

Step 1: perform AP selection, go to Step 2; without AP selection, go to Step 4

Step 2: perform AP selection scheme based on the K-means++ to obtain the connectivity matrix \mathbf{X}

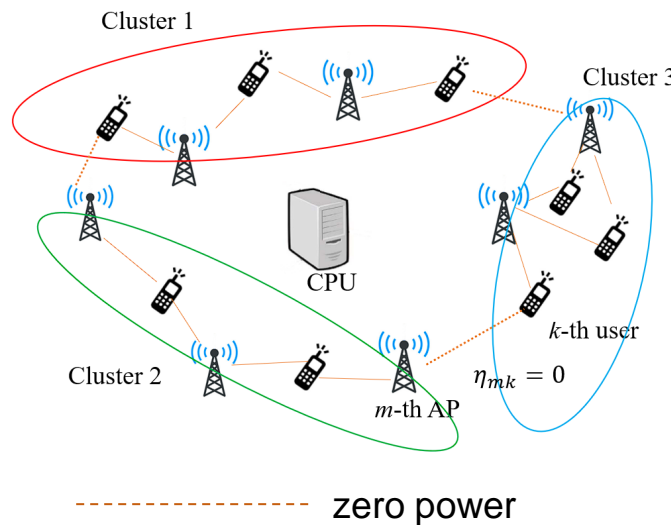
Step 3: if $\mathbf{X}_{mk} = 1$, let $\hat{\alpha}_{mk} = \alpha_{mk}$; else $\hat{\alpha}_{mk} = 0, \forall m, \forall k$. Replace $\{\alpha_{mk}\}$ with $\{\hat{\alpha}_{mk}\}$ as Step 4 input, proceed to the next step

Step 4: obtain an initial feasible solution \mathbf{c}^0 by solving (25), set $n = 1$

Step 5: perform the n -th iteration: solving problem (24) by using SOCP solver, obtain optimal solution \mathbf{c}^*

Step 6: when $n = N_I$, terminate the algorithm; else go to Step 7

Step 7: update $\mathbf{c}^n = \mathbf{c}^*, n = n + 1$, go to Step 5



Pilot contamination:

✦ Same cluster: $|\psi_k^H \psi_{k'}| = 0$

✦ Different clusters:

$$|\psi_k^H \psi_{k'}| \neq 0$$

$$\eta_{mk} \eta_{mk'} = 0 \rightarrow \gamma_{kk'} = 0$$

Eliminate the interference caused by pilot contamination

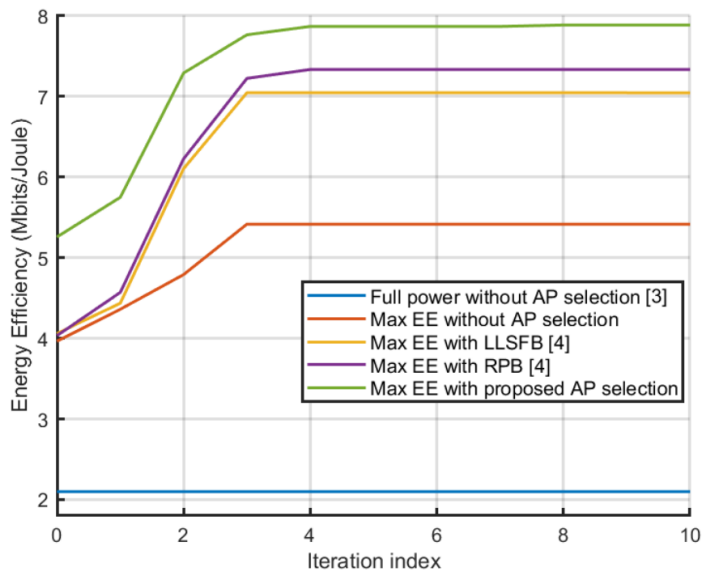


Fig. 2. Illustration of the energy efficiency versus the number of iterations ($M = 100, N = 1, K = 20, \tau_u = 5, D = 1$).

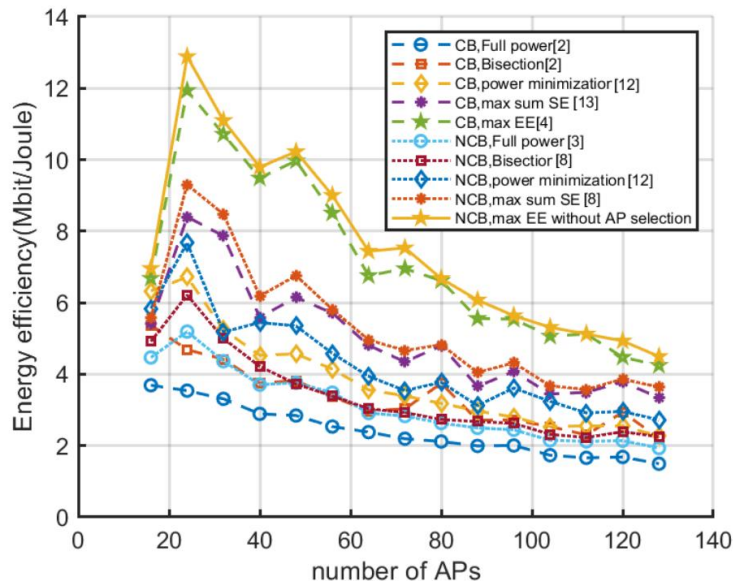


Fig. 3. Illustration of downlink energy efficiency versus the number of APs, comparison of proposed scheme with other power allocation schemes ($N = 2, K = 16, \tau_u = 16, D = 1$).

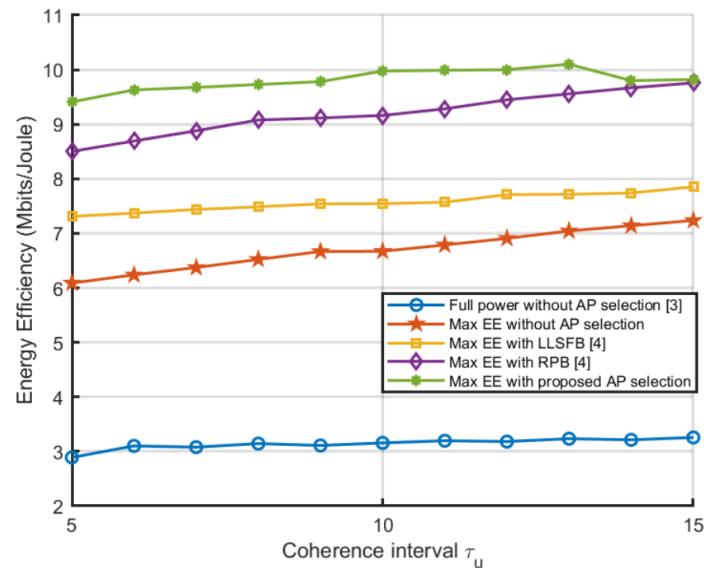


Fig. 4. Illustration of the downlink energy efficiency versus the coherence interval ($M = 100, N = 1, K = 40, D = 1$).

Further improved by AP selection:

- Reduce the backhaul power consumption
- Eliminate the interference caused by pilot contamination.

Converges with only a few iterations.

Significant improvement in energy efficiency compared to other power allocation schemes.

THANK FOR YOUR WATCHING

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