

Designing Wireless Broadband Access for Energy Efficiency

Are Small Cells the Only Answer?

Emil Björnson¹, Luca Sanguinetti^{2,3}, Marios Kountouris^{3,4}

¹ Linköping University, Linköping, Sweden

² University of Pisa, Pisa, Italy

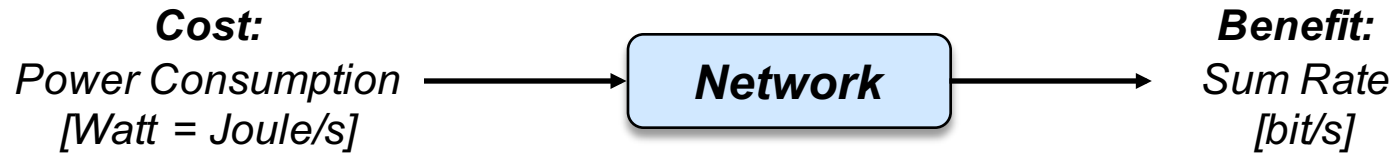
³ CentraleSupélec, Gif-sur-Yvette, France

⁴ Huawei Technologies, Paris, France

INTRODUCTION

Energy Efficiency

- Benefit-Cost Analysis of Networks



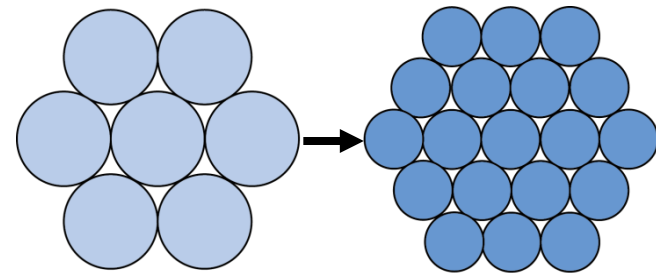
- Definition: Energy Efficiency (EE):

$$EE \text{ [bit/Joule]} = \frac{\text{Average Sum Rate [bit/s/km}^2\text{]}}{\text{Power Consumption [Joule/s/km}^2\text{]}}$$

- Future networks: 1000x more data → 1000x higher EE

- How to Improve Energy Efficiency?

- One approach: Reduce radiated power
Achieved by smaller cells



Is Smaller Cells the Only Answer?

1. Formulate EE maximization mathematically
2. Optimize cell density and other parameters – what do we get?

PROBLEM FORMULATION

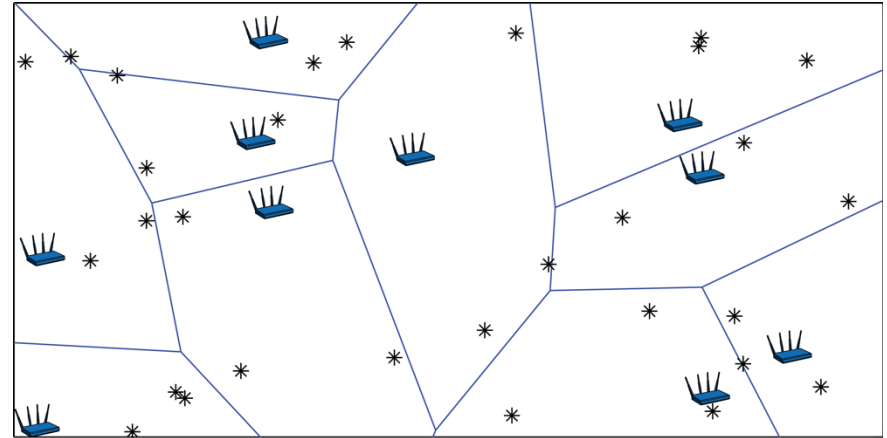
System Model and Average Rate

Random Network Deployment

Access points (AP) positions as
Poisson point process (PPP) Ψ_λ

M antennas per AP, K users per cell

Pathloss: $\omega^{-1}(\text{distance [km]})^{-\alpha}$



Scenario: Downlink Broadband Access

Perfect channel knowledge

Transmit power per user: ρ

Zero-forcing precoding

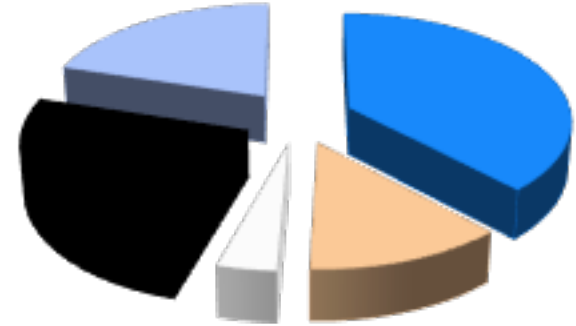
Hardware distortion at users: ϵ^2

Proposition 1: Lower Bound on Average Rate

$$\underline{R} = \underset{\substack{\uparrow \\ \text{Bandwidth}}}{B} \cdot \log_2 \left(1 + \frac{(1 - \epsilon^2)(M - K)}{\frac{2K}{\alpha - 2} + \epsilon^2(M - K) + \frac{M}{(\pi\lambda)^{\alpha/2}} \frac{\omega\sigma^2}{\rho}} \right)$$

Generic Power Consumption Model

- Many Components Consume Power
 - Radiated transmit power
 - Baseband signal processing (e.g., precoding)
 - Active circuits (e.g., converters, mixers, filters)
- Area Energy Consumption [Joule/s/km²]:



$$\text{AEC} = \lambda \left(\frac{K\rho}{\eta} + C_0 + D_0M + C_1K + D_1MK \right)$$

*Power amplifier
(η is efficiency)*

*Circuit power per
transceiver chain*

*Fixed power
(backhaul, load-ind. processing)*

*Cost of digital signal processing
(e.g., precoding)*

***Nonlinear increasing
function of M and K***

Many coefficients: η, C_i, D_i for $i = 0, 1$

Problem Formulation

Energy Efficiency Optimization

$$\begin{array}{ll} \text{maximize} & \frac{\lambda K \underline{R}}{\lambda \left(\frac{K \rho}{\eta} + C_0 + D_0 M + C_1 K + D_1 MK \right)} \\ \rho, \lambda, M, K & \\ \text{subject to} & \underline{R}/B = \gamma \end{array}$$

Optimization variables:

$$\begin{array}{ll} \rho = \text{transmit power}, & \lambda = \text{AP density}, \\ M = \text{antennas per AP}, & K = \text{users per AP} \end{array}$$

Spectral efficiency (SE) constraint γ needed to not get overly low rates

ANALYTICAL AND NUMERICAL RESULTS

Optimality of Small Cells

Theorem 1: Optimal AP Density

The EE increases with λ .

EE maximized as $\lambda \rightarrow \infty$ or at some upper value λ_{\max}

Saturation Property

Higher density $\lambda \rightarrow$ Less transmit power \rightarrow Eventually negligible

Simulations show saturation at $\lambda \geq 10^2$

50 meters between APs: Saturation appears in practice!

Optimization of Remaining Variables

Theorem 2: Optimal Transmit Power

Constraint satisfied if $\rho^* = \frac{\frac{2^\gamma - 1}{1 - 2^\gamma \epsilon^2} \frac{\omega \sigma^2 \Gamma(\alpha/2 + 1)}{(\pi \lambda)^{\alpha/2}}}{M - K - \frac{2^\gamma - 1}{1 - 2^\gamma \epsilon^2} \frac{2K}{\alpha - 2}}$

Removes ρ from
EE optimization problem
(Only M and K remain)

Theorem 3: Optimal Number of Antennas (fixed K)

EE maximized by $M^* = K + \frac{2K(2^\gamma - 1)}{(\alpha - 2)(1 - 2^\gamma \epsilon^2)} + \sqrt{\frac{2^\gamma - 1}{1 - 2^\gamma \epsilon^2} \frac{K \omega \sigma^2 \Gamma(\alpha/2 + 1)}{\eta(\pi \lambda)^{\alpha/2} (D_0 + D_1 K)}}$

Theorem 4: Optimal Number of Users (fixed $M/K = \beta$)

EE maximized by $K^* = \sqrt{\frac{\frac{2^\gamma - 1}{1 - 2^\gamma \epsilon^2} \frac{\omega \sigma^2 \Gamma(\alpha/2 + 1)}{\eta(\pi \lambda)^{\alpha/2}}}{\beta D_1 (\beta - 1 - \frac{2^\gamma - 1}{1 - 2^\gamma \epsilon^2} \frac{2}{\alpha - 2})}} + \frac{C_0}{\beta D_1}$

Iterate between
these till convergence:
**Find real-valued
global solution**

Tradeoffs and connections established formulas!

Simulation Parameters

Simulation Parameter	Symbol	Value
Pathloss exponent	α	3.76
Pathloss over noise at 1 km	ω/σ^2	33 dBm
Amplifier efficiency	η	0.39
Level of hardware impairments	ϵ	0.05
Bandwidth	B	20 MHz
Static power	C_0	10 W
Circuit power per active user	C_1	0.1 W
Circuit power per AP antenna	D_0	1 W
Signal processing coefficient	D_1	3.12 mW

Impact of Number of Antennas and Users

Simulation

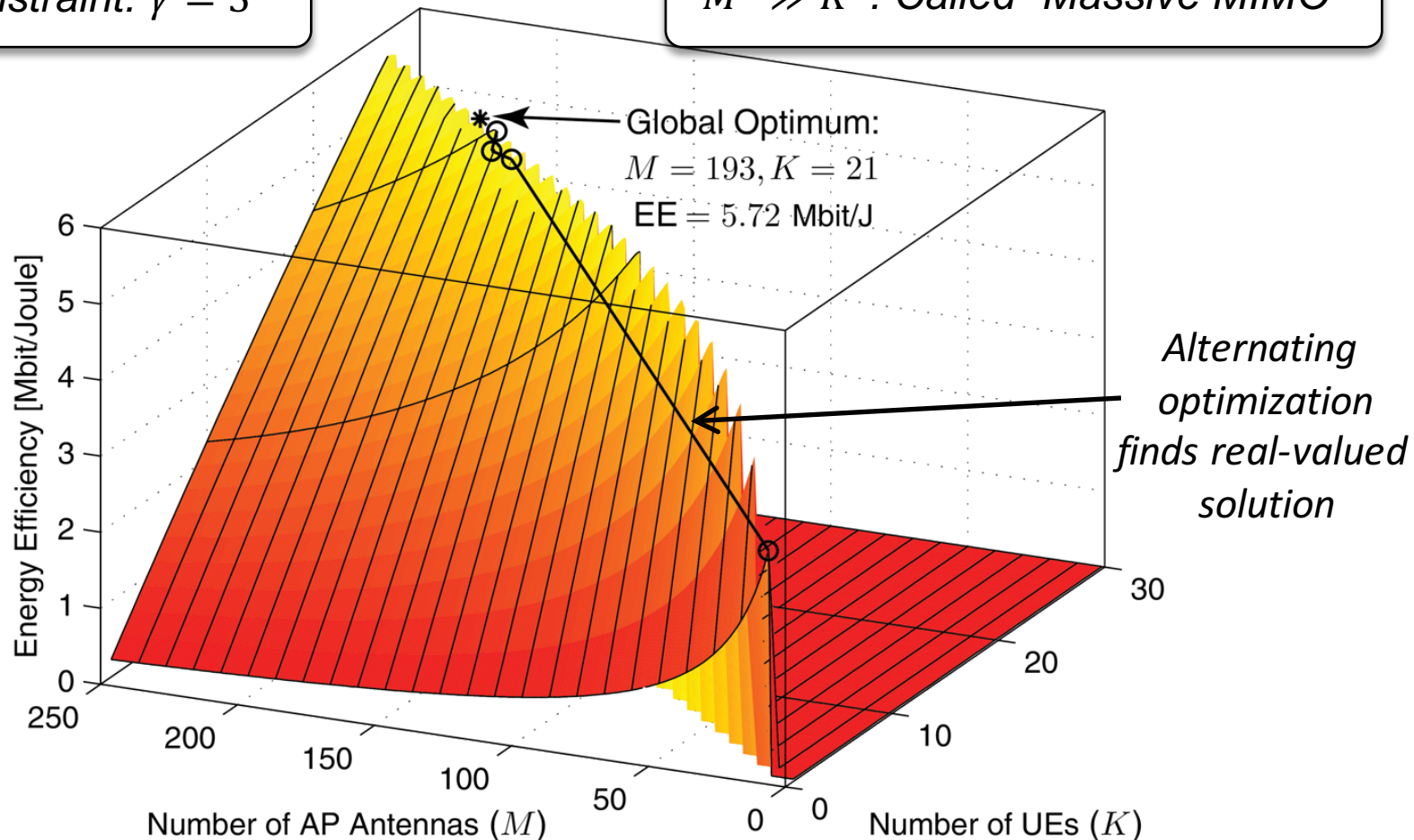
AP density $\lambda = 10^4$

SE constraint: $\gamma = 3$

Observations

Optimal: $M^* = 193, K^* = 21$

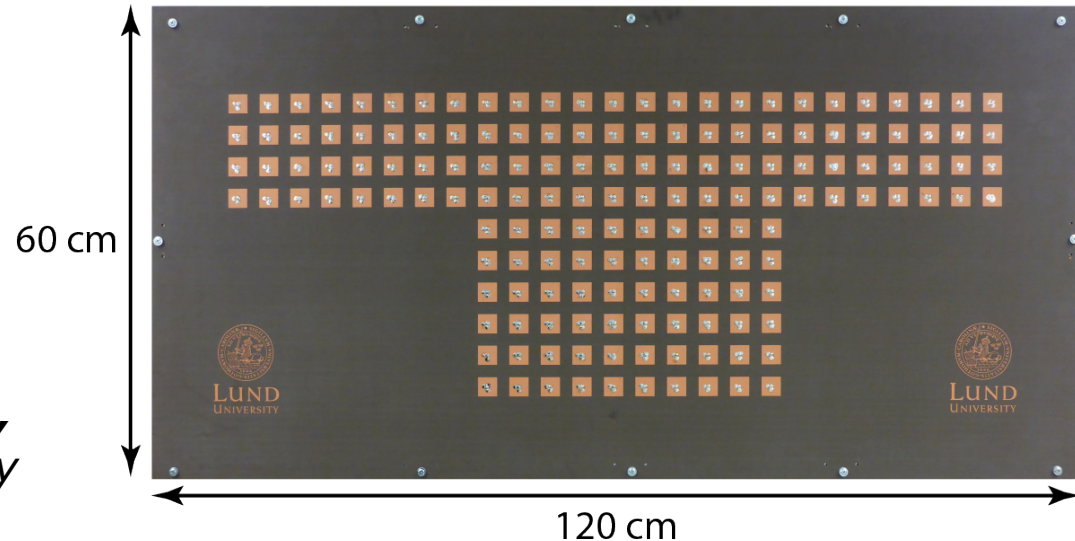
$M^* \gg K^*$: Called “Massive MIMO”



Is it Ridiculous with 200 Antennas?

- Dimensionality: Half-wavelength Antenna Spacing
 - Example: 3.7 GHz
Spacing: 4 cm
 - Array = Flat-screen TV

160 dual-polarized antennas,
LuMaMi testbed, Lund University

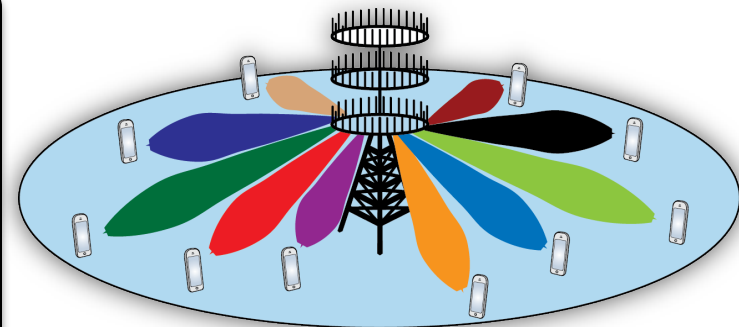


Why Massive MIMO, Not Only Small Cells?

Small cells improve SNR, but not SINR

Massive MIMO improves SINRs by precoding

Circuit power costs are shared between users



Impact of User Density

Simulation

Fixed user density μ users/km²

EE maximization with: $K\lambda = \mu$

Range: $\mu = 10^2$ (rural) to $\mu = 10^5$ (mall)

Low User Density

Add more cells with $K \approx 1$

Most important to reduce pathloss

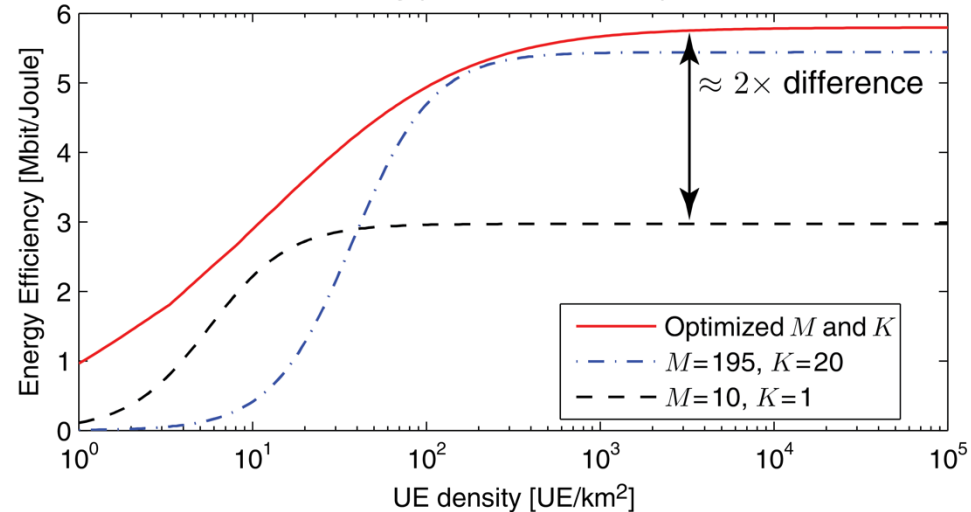
High User Density

Small cells with Massive MIMO

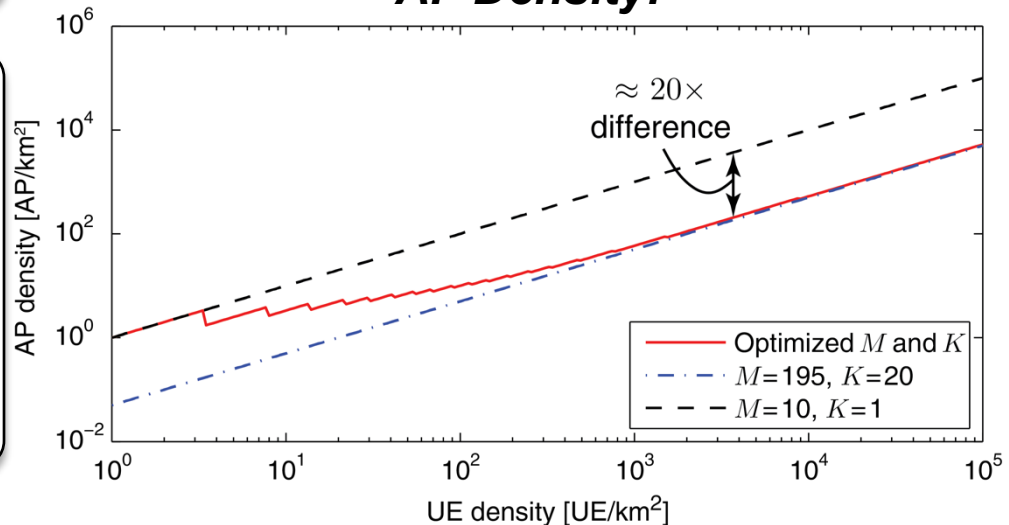
Saturation for $\mu \geq 100$

Covers most practical scenarios:
EE independent of user load!

Energy Efficiency:



AP Density:



SUMMARY

Summary

- Designing Networks for Energy Efficiency
 - Optimize: AP density, transmit power, and antennas/users per cell
 - Analytical optimization: EE maximizing network deployment was found!
 - Solution: Small cells with Massive MIMO capability
 - Intuition: Small cells → Negligible transmit power
Massive MIMO → Less interference, share costs over users
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- Further Results:
 - Take channel estimation and imperfect channel knowledge into account
 - 1. E. Björnson, L. Sanguinetti, M. Kountouris, “*Deploying Dense Networks for Maximal Energy Efficiency: Small Cells Meet Massive MIMO*,” Submitted to IEEE JSAC. (<http://arxiv.org/pdf/1505.01181>)
 - 2. E. Björnson, L. Sanguinetti, M. Kountouris, “*Energy-Efficient Future Wireless Networks: A Marriage between Massive MIMO and Small Cells*,” Proceedings of IEEE SPAWC, July 2015. (<http://arxiv.org/pdf/1506.01051>)

QUESTIONS?

Visit Emil Björnson online:

<http://www.commsys.isy.liu.se/en/staff/emibj29>