



Stochastic Geometry of RIS and NT Networks

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**Indo-French Seminar
6G Wireless Networks
Challenges and Opportunities
October 10, 2024**

Structure of the presentation

- **Zoom 0: Stochastic Geometry and Wireless Networks**
- **Zoom 1: RIS Enhanced Cellular Networks**
- **Zoom 2: NTN Cellular Networks**
- **Zoom 3: RIS & NTN Networks**

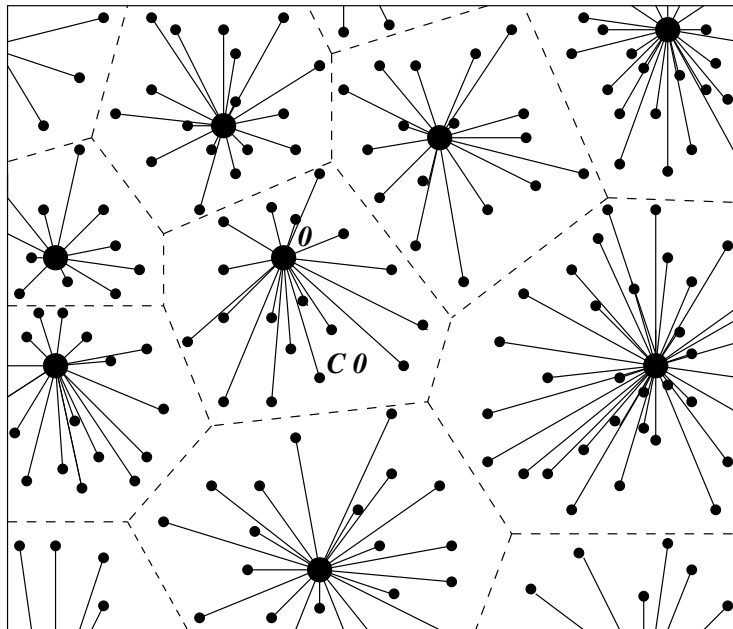
ZOOM 0: POISSON STOCHASTIC GEOMETRY

- A few basic models:
 - Spatial Poisson point process
 - Spatial Shot–noise fields : Interference
 - Poisson–Voronoi tessellation: “Connection to closest ”
- [Chiu, Stoyan, Kendall and Mecke 13]
Stochastic Geometry and its Applications
- [Baccelli, Blaszczyzyn, Karray 24]
Random Measures, Point Processes, & Stochastic Geometry

STOCHASTIC GEOMETRY & CELLULAR NETWORKS

- **Started around 2000**
branch of stochastic geometry dedicated to wireless networks
- **Extends to system level analysis of other networks**
WiFi, IoT, NTN
- **Google query *wireless stochastic geometry* → millions of hits**
more than 10 books, tens of thousands articles
- **Large scientific community in this field worldwide**

POISSON-VORONOI CELLULAR NETWORKS



- - Stations
- - Users
- - Connections
- - Cells

- Base stations (BSs) arranged according to an homogeneous Poisson point process of intensity λ in \mathbb{R}^2
- UEs
 - located according to some independent stationary point process
 - each user is served with the closest BS \rightarrow
Poisson Voronoi Cells

SHANNON RATE IN POISSON-VORONOI CELLULAR NETWORKS WITH RAYLEIGH FADES

- SINR experienced by typical user: $\text{SINR} := \frac{S}{I+N}$
 - S: Signal power: stems from closest BS
 - I: Interference power: from BSs outside Voronoi cell of typical user
 - N: thermal noise power
- Shannon rate offered to typical user: $\mathcal{T} \sim B \log(1 + \text{SINR})$
- Question: Law of the Shannon rate offered to typical user

COVERAGE/SHANNON RATE

$$p_c(\mathbf{T}, \lambda, \beta) = \Pr_{\mathbf{u}}^0[\mathbf{SINR} > \mathbf{T}] = \Pr_{\mathbf{u}}^0[\text{Shannon rate} > \mathbf{B} \log(\mathbf{1} + \mathbf{T})]$$

- As in statistical physics, this is **equivalent to spatial averages**
 - Average fraction of users who achieve SINR at least \mathbf{T}
 - Average fraction of the network area in “ \mathbf{T} -coverage”
- Assumptions on propagation for next result:
 - **Power law path loss** :
at distance r , $l(r) = r^\beta$, $\beta > 2$, **path loss exp.**
 - **Rayleigh fading model**: Exponential fade with mean $\frac{1}{\mu}$

P-V-S THEOREM FOR RAYLEIGH FADES

■ Theorem

J.G. Andrews, F.B and R.K. Ganti, IEEE Tr. Comm. 11

$$p_c(\mathbf{T}, \lambda, \beta) = \pi \lambda \int_0^{\infty} e^{-\pi \lambda v(1+\rho) - \mu \mathbf{T} N v^{\beta/2}} dv \quad \text{with} \quad \rho = \mathbf{T}^{\frac{2}{\beta}} \int_{\mathbf{T}^{\frac{-2}{\beta}}}^{\infty} \frac{1}{1+u^{\beta/2}} du$$

- Step 1 : Poisson–Voronoi : distance to closest BS: Rayleigh distr. → S
- Step 2 : Poisson–Voronoi : Poisson shot-noise outside a ball → I
- Step 3 : Poisson–Voronoi–Shannon : law of SINR and Shannon rate

■ Closed form expressions in e.g. interference limited case

PROOF

■ Step 1: Poisson-Voronoi

$$\begin{aligned}
 p_c(\mathbf{T}, \lambda, \beta) &= \mathbb{P}[\mathbf{SINR} > \mathbf{T}] \\
 &= \int_{r>0} \mathbb{P}[\mathbf{SINR} > \mathbf{T}] f_r(\mathbf{r}) d\mathbf{r} \\
 &= \int_{r>0} \mathbb{P} \left[\frac{\mathbf{S} r^{-\beta}}{\mathbf{N} + \mathbf{I}_r} > \mathbf{T} \right] e^{-\pi \lambda r^2} 2\pi \lambda r d\mathbf{r} \\
 &= \int_{r>0} e^{-\pi \lambda r^2} \mathbb{P}[\mathbf{S} > \mathbf{T} r^\beta (\mathbf{N} + \mathbf{I}_r)] 2\pi \lambda r d\mathbf{r}
 \end{aligned}$$

\mathbf{I}_r : interference power given the closest BS is at distance r

PROOF (continued)

■ **Step 2: Rayleigh.** Since $\mathbf{S} \sim \exp(\mu)$,

$$\mathbb{P}(\mathbf{S} > \mathbf{Tr}^\beta(\mathbf{N} + \mathbf{I}_r)) = \mathbb{E}[\exp(-\mu \mathbf{Tr}^\beta(\mathbf{N} + \mathbf{I}_r))] = e^{-\mu \mathbf{Tr}^\beta \mathbf{N}} \mathcal{L}_{\mathbf{I}_r}(\mu \mathbf{Tr}^\beta)$$

with $\mathcal{L}_{\mathbf{I}_r}(\mathbf{s})$ the Laplace transform of the interference

Thus

$$\mathbf{p}_c(\mathbf{T}, \lambda, \beta) = \int_{r>0} e^{-\pi \lambda r^2} e^{-\mu \mathbf{Tr}^\beta \mathbf{N}} \mathcal{L}_{\mathbf{I}_r}(\mu \mathbf{Tr}^\beta) 2\pi \lambda r dr$$

PROOF (*continued*)

■ **Step 3: Interference as shot noise field**

$$\mathcal{L}_{\mathbf{I}_r}(\mathbf{s}) = \exp \left(-2\pi\lambda \int_r^\infty (1 - \mathcal{L}_F(\mathbf{s}\mathbf{v}^{-\beta})) \mathbf{v} d\mathbf{v} \right)$$

with \mathcal{L}_F the Laplace Transform of the general fading

PROOF (continued)

- **Computational step 4: plugging in $s = \mu \mathbf{T} r^\beta$ gives**

$$\begin{aligned}
 \mathcal{L}_{I_r}(\mu \mathbf{T} r^\beta) &= \exp \left(-2\pi\lambda \int_r^\infty (1 - \mathcal{L}_F(\mu \mathbf{T} r^\beta \mathbf{v}^{-\beta})) \mathbf{v} d\mathbf{v} \right) \\
 &= \exp \left(-2\pi\lambda \int_0^\infty \left(\int_r^\infty (1 - e^{-\mu \mathbf{T} r^\beta \mathbf{v}^{-\beta} \mathbf{g}}) \mathbf{v} d\mathbf{v} \right) \mathbf{f}(\mathbf{g}) d\mathbf{g} \right) \\
 &= \exp \left(\lambda\pi \mathbf{r}^2 - \frac{2\pi\lambda (\mu \mathbf{T})^{\frac{2}{\beta}} \mathbf{r}^2}{\beta} \right. \\
 &\quad \left. \int_0^\infty \mathbf{g}^{\frac{2}{\beta}} [\Gamma(-2/\beta, \mu \mathbf{T} \mathbf{g}) - \Gamma(-2/\beta)] \mathbf{f}(\mathbf{g}) d\mathbf{g} \right)
 \end{aligned}$$

with $\mathbf{f}(\mathbf{g})$ the PDF of \mathbf{F}

EXAMPLE OF USE

- **Example of System Level Question** that can be answered:

When does BS densification lead to a decrease of spectral efficiency?

- When $N = 0$

- Constant spectral efficiency for all densities !

$$p_c(\mathbf{T}, \lambda, \beta) = \frac{1}{1 + \rho(\mathbf{T}, \beta)}$$

Scale invariance: only true for power law attenuation

- For bounded attenuation functions

$$p_c(\mathbf{T}, \lambda, \beta) \rightarrow 0 \text{ as } \lambda \rightarrow \infty$$

Joint work with **A. Alammouri, J. G. Andrews, IEEE Trans. Information Theory, 2019**

STATE OF THE ART – EXTENSIONS – FUTURE

- **Further point processes: beyond Poisson,**

Joint work w. J.G. Andrews, H. Dhillon, Y. Li, *IEEE Tr. Comm.* 15

- **Further propagation models: beyond scale invariance,**

Joint work w. A. Alammouri, J.G. Andrews, *IEEE Tr. Inf. Theory* 19

- **Obstacle/Shadowing models: essential for millimeter waves,**

Joint work w. J. Lee, *IEEE Infocom* 18

- **MIMO and BeamForming : optimal beam management**

Joint work w. S. Kalamkar and NBL, *IEEE Tr. Wireless* 22

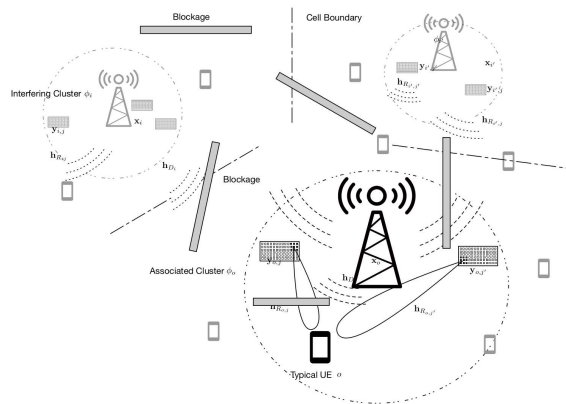
- **and also Power Control, OFDM, Coexistence with WiFi, Successive Interference Cancellation, CoMP, Vehicular networks, etc.**

- Stochastic Geometry for 6G and Beyond
 - RIS
 - NTN
 - Deterministic Latency Networking
 - JCAS
 - Cell Free
- Worldwide efforts engaged on the matter

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ZOOM 1: RIS ENHANCED CELLULAR NETWORK

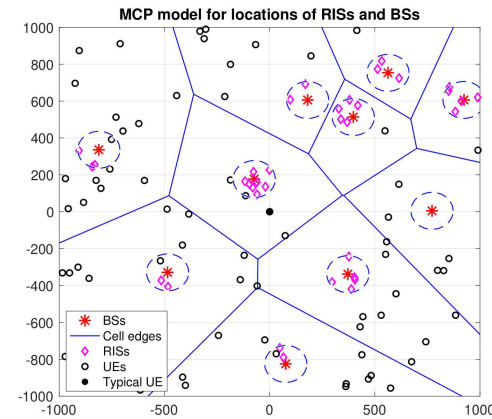


RIS cluster with each BS

Blockages BSs \rightarrow UEs

Each RIS

- reflects BS signal
- beamforms to UEs



Matérn Cluster Process model

- **BS PPP** of intensity λ_{BS}
- **RIS PPP** of intensity λ_{RIS} in ring $[r, R]$ around each BS
- **UE PPP** of intensity λ_{UE}

MAIN TECHNICAL NOVELTY

- Under **OFDM Assumptions**, Signal has two components
 - Direct path with power Q_{S_D}
 - Reflected paths with RIS beamforming with power Q_{S_R} characterized by its Laplace Transform
- Interference power Q_I characterized by its Laplace Transform (LT of MCP known in closed form)

■ Coverage Probability

$$P_c(T) \triangleq \mathbb{P}(\text{SIR} \geq T) = \mathbb{E}_r[\mathbb{P}(\text{SIR} \geq T|r)] = \mathbb{E}_r \left[\mathbb{P} \left(\frac{Q_{S_D}(r) + Q_{S_R}(r)}{Q_I(r)} \geq T \middle| r \right) \right].$$

$$\mathbb{P} \left[\frac{Q_{S_D}(r) + Q_{S_R}(r)}{Q_I(r)} \geq T \middle| r \right] = \mathbb{P}[Q_{S_D}(r) \geq TQ_I(r) - Q_{S_R}(r)|r].$$

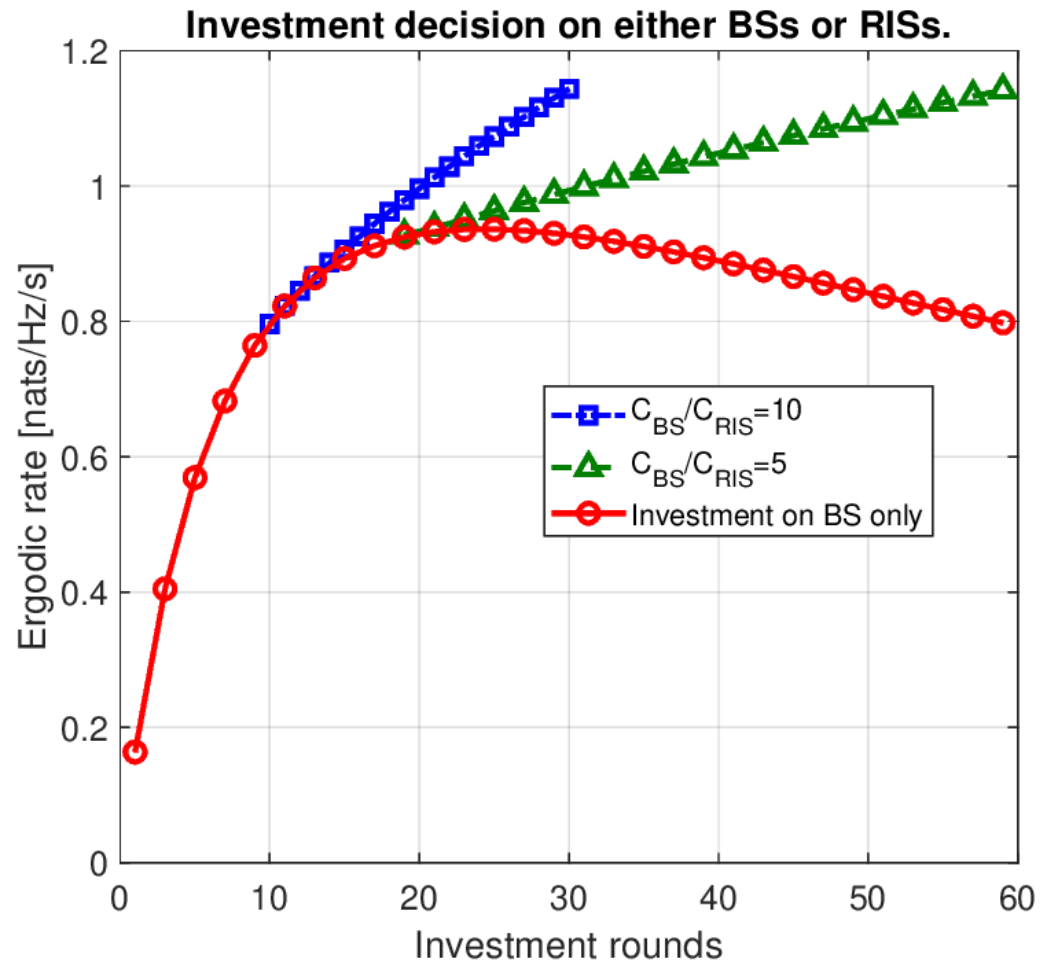
- Need to separate the positive and negative parts in the RHS of the last equation. **Wiener Hopf Factorization**

SYSTEM LEVEL QUESTIONS

- Influence on **Spectral Efficiency** of
 - Geometry of clusters : more or less spread?
 - RIS resources organization
 - Bigger and lesser RISs or the other way around?
- For optimal configuration
 - Mean spectral efficiency gain brought by RISs
 - Dependence of this gain in function of obstacle density

Performance Analysis of RIS-assisted MIMO-OFDM Cellular Networks Based on Matern Cluster Processes, G. Sun, F. Baccelli, K. Feng, L.G. Uzeda Garcia, S. Paris, ArXiv 2024

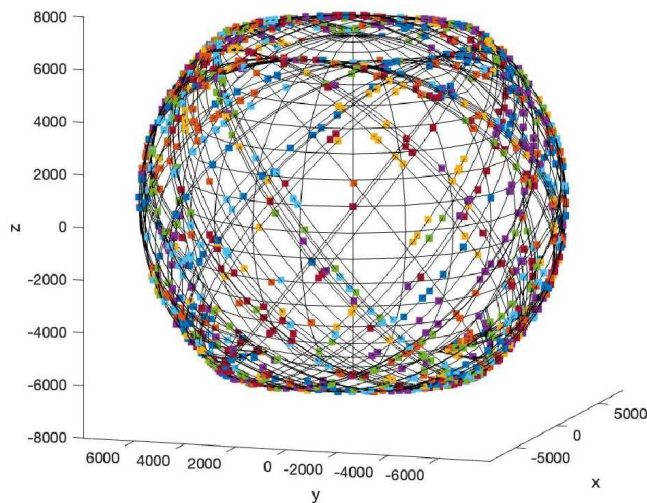
EXAMPLE OF ECONOMIC ANALYSIS



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ZOOM 2: NTN BASED CELLULAR NETWORKS



A constellation of LEO satellites

- Spherical geometry with orbiting BSs
- BSs move on orbits with given inclination

■ Need for new SG models

Orbits?, Voronoi?

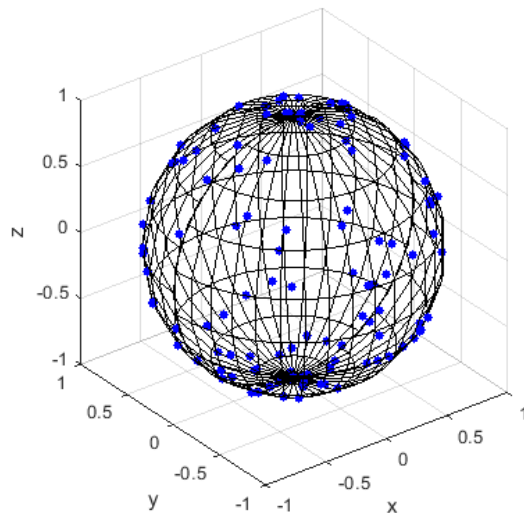
Coverage? SINR?

Spectral Efficiency?

■ System level questions

- Interaction/Interference between 5G and NTN
- Optimal orbit/satellite density
- 5G offloading analysis

FIRST STEPS



A Binomial PP on the sphere
[Okati et al. 20] IEEE Trans.
Comm.

Binomial p.p. with satellites
uniformly distributed on a
sphere

Basic questions similar to
those on the plane but in
spherical geometry

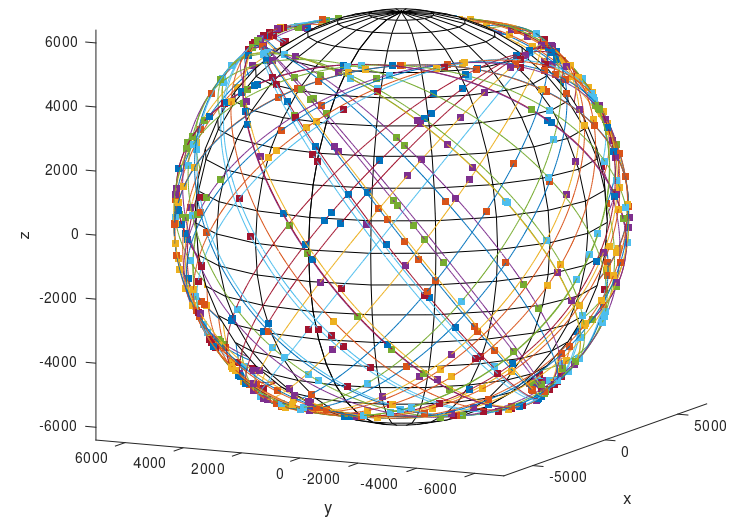
Limitations

1. Geometry: No orbital planes
2. Analysis: clustering of interference ignored

ONGOING STEPS 1

Build a stochastic geometry framework with requirements:

- Characterize orbital planes with various longitudes and inclinations
- Address distribution of LEO satellites on orbital planes
- Evaluate the SINR distribution and spectral efficiency



Cox Point Processes for
Multi-Altitude LEO Satellite
Networks, C.S. Choi, F.
Baccelli, *IEEE Trans. Veh.
Technol.*, 2024

ONGOING STEPS 2

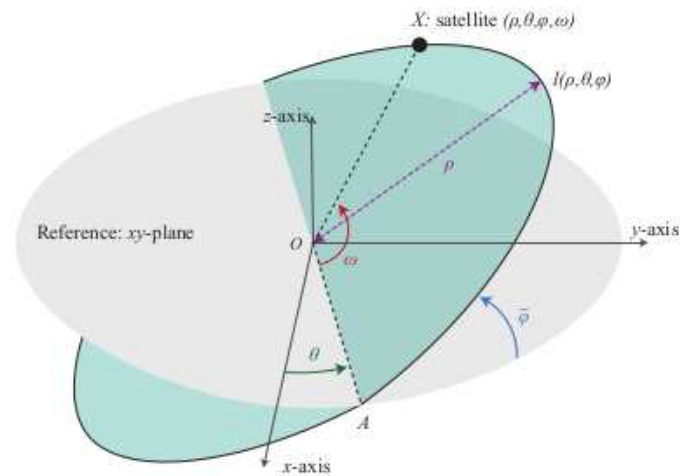
Build a stochastic geometry framework allowing one to:

- Evaluate impact of NTN on terrestrial
- Evaluate synergy between NTN and terrestrial

Ongoing work with J. Park and N. Lee [ArXiv 2024](#)

MAIN TECHNICAL NOVELTIES

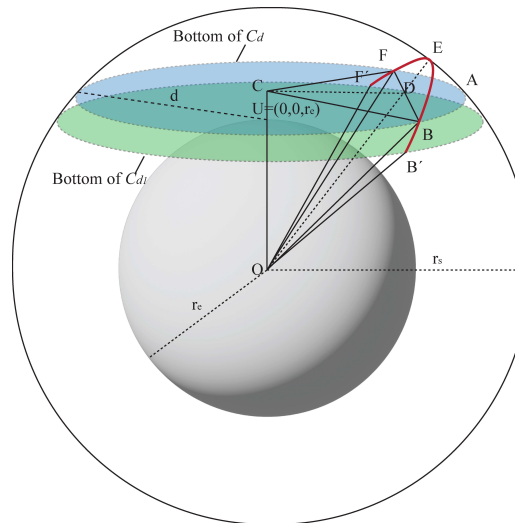
■ Representation of Orbits in SG



Inclination and Ascending Point

MAIN TECHNICAL NOVELTIES (*continued*)

■ Spherical Geometry



Visibility Cap, Visibility Arc of an Orbit

MAIN TECHNICAL NOVELTIES (*continued*)

- **Distributions Needed to Analyze Downlink SINR**
 - **Signal:** Distance to closest visible satellite
 - **Interference:** Shot Noise created by other visible satellites
- **Extension of the Laplace Transform approach to evaluate**
 - **Probability of Coverage**
 - **Spectral Efficiency**

Latitude dependent for most deployments

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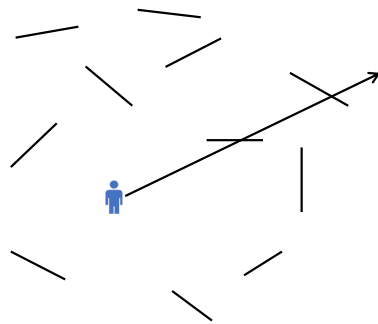
ZOOM 3: NTN AND RIS

Question: How Much Can Reconfigurable Intelligent Surfaces Augment Sky Visibility ?

- Urban environments
- Millimeter wave bands (either 5G or NTN) blocked by buildings
- Connectivity of terrestrial users to NTN entities
- Visibility and coverage extension provided by RIS installed on top of buildings
- Distribution of Visibility Angle
- Distribution of RIS Augmented Visibility Angle
- Metrics:
 - Angular
 - Linear
 - Coverage

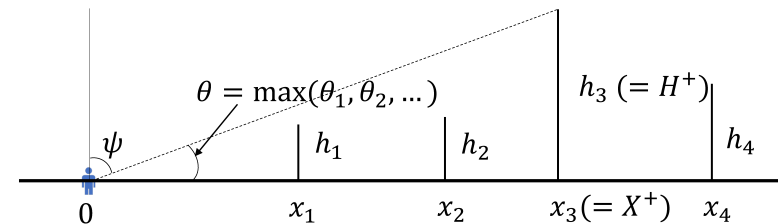
SYSTEM MODEL

Urban Environment 3D Model



Segments : buildings in a 2D plane
3rd Dimension : building heights

Urban Environment 2D Model



$\{x_i\}$ building locations in chosen direction
 $\{h_i\}$ building heights
 θ visibility angle

SYSTEM MODEL (*continued*)■ **M/M Model**

- $\{\mathbf{x}_i\}_{i \in \mathbb{Z}}$: Homogeneous Poisson Point Process (λ) on \mathbb{R}
- $\{h_i\}$ i.i.d. exponential RVs with parameter μ

■ **M/D Model**

- $\{\mathbf{x}_i\}_{i \in \mathbb{Z}}$: Homogeneous Poisson Point Process (λ) on \mathbb{R}
- $\{h_i\}$ constant with common value μ^{-1}

■ **D/M Model**

- **Extensions** : All other models of queueing theory (e.g. GI/GI.....)

DISTRIBUTION OF VISIBILITY ANGLE

- **THEOREM** In the M/M case, the CDF of $\tan \theta$ is

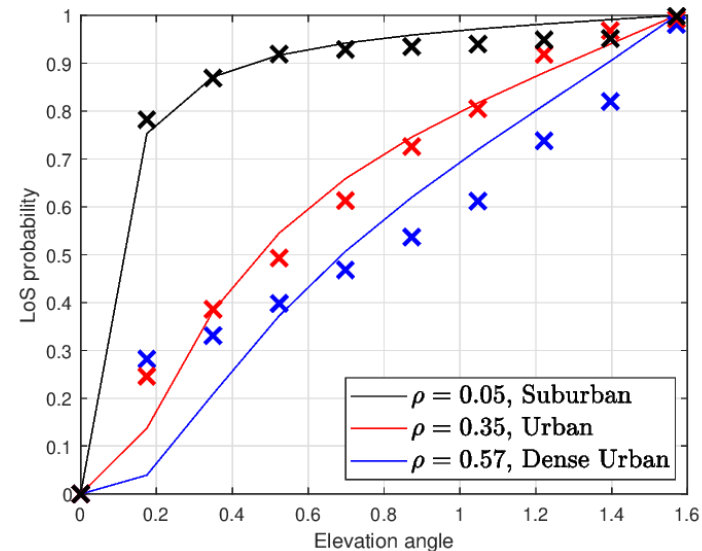
$$\mathbb{P}[\tan \theta \leq t] = e^{-\frac{\rho}{t}}, \quad t \geq 0,$$

which is a **Fréchet distribution** with

- shape parameter $\alpha = 1$
- scale parameter $s = \rho = \frac{\lambda}{\mu}$

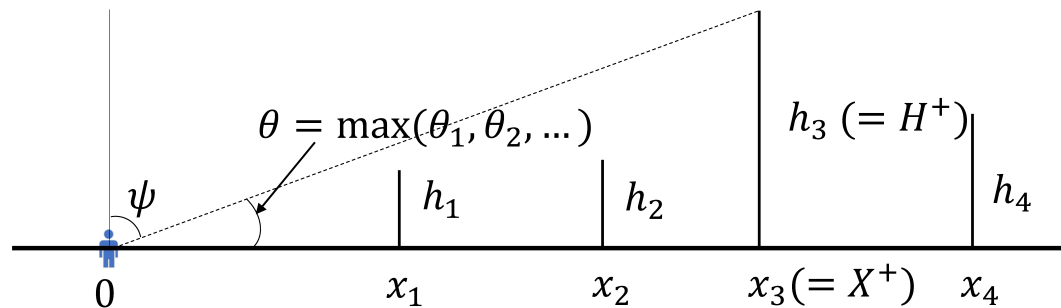
- Proof obtained from the Laplace Functional of the PPP

Closed form expressions for M/D and M/W as well



Comparison to 3GPP data

JOINT DISTRIBUTION



- **THEOREM** The joint density of (X^+, H^+) at (x, h) is

$$\mathbf{j}(\mathbf{x}, \mathbf{h}) = \lambda \mu e^{-\mu h - \frac{\lambda x}{\mu h}}, \quad \mathbf{h} \geq \mathbf{0}, \mathbf{x} \geq \mathbf{0}$$

JOINT DISTRIBUTION (*continued*)

- The density of H^+ at h is the Gamma distribution

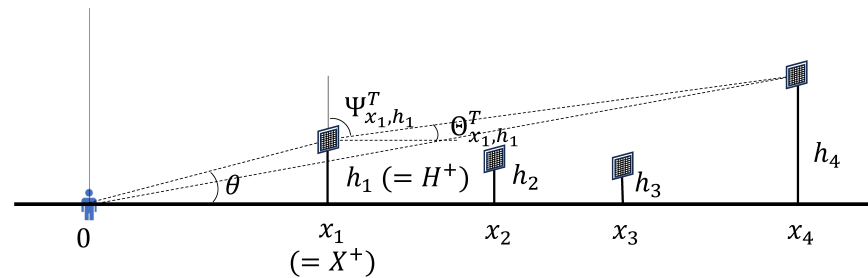
$$g(\mathbf{h}) = \mu^2 \mathbf{h} e^{-\mu \mathbf{h}}$$

- The density of X^+ at \mathbf{x} is

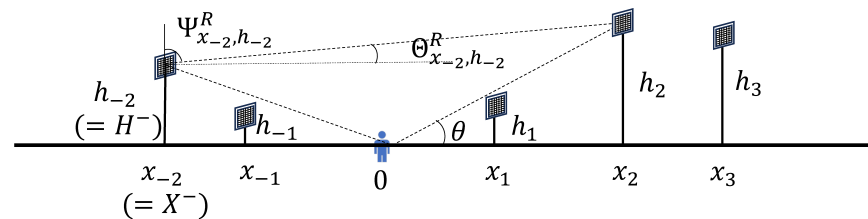
$$k(\mathbf{x}) = 2\lambda \sqrt{\lambda \mathbf{x}} K_1(2\sqrt{\lambda \mathbf{x}})$$

where $K_n(\cdot)$ is the modified Bessel function of the second kind

DISTRIBUTION OF RIS AUGMENTED VISIBILITY ANGLE



Transmissive mode : improved visibility angle $\Theta_{x,h}^T$



Reflective mode : similar definition

DISTRIBUTION

■ THEOREM

The conditional CDF of $\tan \Theta^T$ given that $(\mathbf{X}^+, \mathbf{H}^+) = (\mathbf{x}, \mathbf{h})$ is

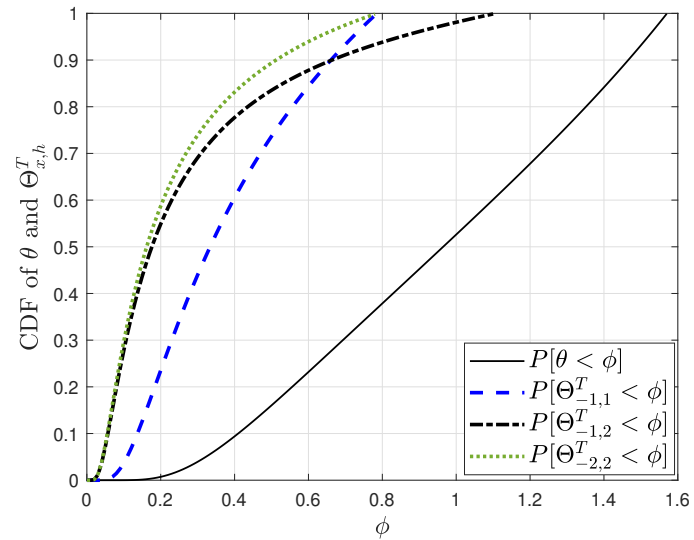
$$\begin{aligned} \mathbb{P}[\tan \Theta_{\mathbf{x}, \mathbf{h}}^T \leq \mathbf{t}] &= \mathbb{P}[\tan \Theta^T \leq \mathbf{t} | \mathbf{X}^+ = \mathbf{x}, \mathbf{H}^+ = \mathbf{h}] \\ &= \begin{cases} \exp\left(-\rho \left[\frac{1}{\mathbf{t}} - \frac{\mathbf{x}}{\mathbf{h}}\right] e^{-\mu \mathbf{h}}\right), & \text{for } 0 < \mathbf{t} \leq \frac{\mathbf{h}}{\mathbf{x}}, \\ 1, & \text{for } \mathbf{t} > \frac{\mathbf{h}}{\mathbf{x}} \end{cases} \end{aligned}$$

■ Closed form expressions for conditional

- density
- moments

■ Closed form expressions for conditional distribution of $\tan \Theta^R$

DISTRIBUTION (continued)



Visibility enhancement by transmissive RISs with $\lambda = 1$ and $\mu = 1$

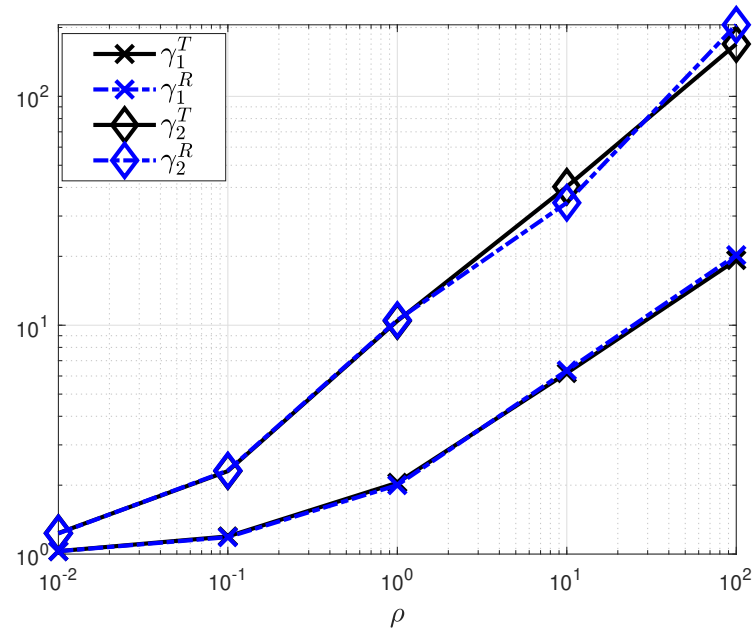
ANGULAR METRICS

- **Angular metrics** to quantify the visibility enhancement by RISs

$$\gamma_1^{\text{T}} \triangleq \frac{\mathbb{E}[\Psi^{\text{T}}]}{\mathbb{E}[\psi]}, \quad \gamma_1^{\text{R}} \triangleq \frac{\mathbb{E}[\Psi^{\text{R}}]}{\mathbb{E}[\psi]}$$

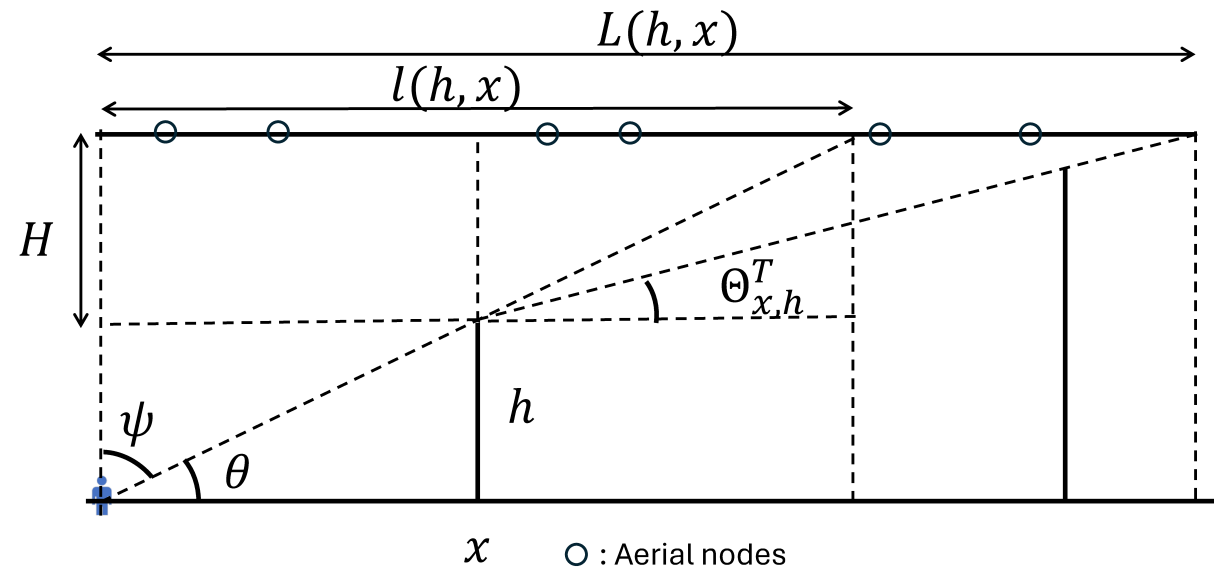
$$\gamma_2^{\text{T}} \triangleq \mathbb{E} \left[\frac{\Psi^{\text{T}}}{\psi} \right], \quad \gamma_2^{\text{R}} \triangleq \mathbb{E} \left[\frac{\Psi^{\text{R}}}{\psi} \right]$$

- Measure how much the visibility angle is increased by using the RISs
- Can be evaluated in integral form thanks to the analytical formulas

ANGULAR METRICS (*continued*)

Visibility enhancement with respect to ρ

LINEAR METRICS



NTN nodes are deployed at the same altitude
Extension of visible regions at this altitude by transmissive RISs

RESULTS ON LINEAR METRICS

- M/M Model, Transmissive mode
- Given $(\mathbf{X}^+, \mathbf{H}^+) = (\mathbf{x}, \mathbf{h})$

$$|\mathbf{l}(\mathbf{x}, \mathbf{h})| = \mathbf{x} + \mathbf{H} \frac{\mathbf{x}}{\mathbf{h}}$$

$$|\mathbf{L}(\mathbf{x}, \mathbf{h})| = \mathbf{x} + \frac{\mathbf{H}}{\tan \Theta_{\mathbf{x}, \mathbf{h}}^{\mathbf{T}}}$$

- Conditional Means

$$|\mathbf{l}(\mathbf{x}, \mathbf{h})| = \mathbf{x} + \mathbf{H} \frac{\mathbf{x}}{\mathbf{h}}, \quad \mathbb{E}[|\mathbf{L}(\mathbf{x}, \mathbf{h})|] = \mathbf{x} + \frac{e^{\mathbf{h}\mu} \mathbf{h} + \mathbf{x}\rho}{\mathbf{h}\rho} \mathbf{H}$$

- Means

$$\mathbb{E}[|\mathbf{l}|] = \frac{2 + \mathbf{H}\mu}{\lambda}, \quad \mathbb{E}[|\mathbf{L}|] = \infty$$

PROBABILITY OF COVERAGE

- UAVs assumed to be distributed as a homogeneous PPP Φ_u with intensity ν at altitude $h + H$
- $\tau(\mathbf{x}, \mathbf{h})$: conditional probability of coverage given (\mathbf{x}, \mathbf{h}) and given no initial coverage

$$\tau(\mathbf{x}, \mathbf{h}) = \mathbb{P}[\Phi_u(\mathbf{L}(\mathbf{x}, \mathbf{h})) > \mathbf{0} | \Phi_u(\ell(\mathbf{x}, \mathbf{h})) = \mathbf{0}] = 1 - \frac{\rho}{e^{h\mu} \mathbf{H}\nu + \rho}$$

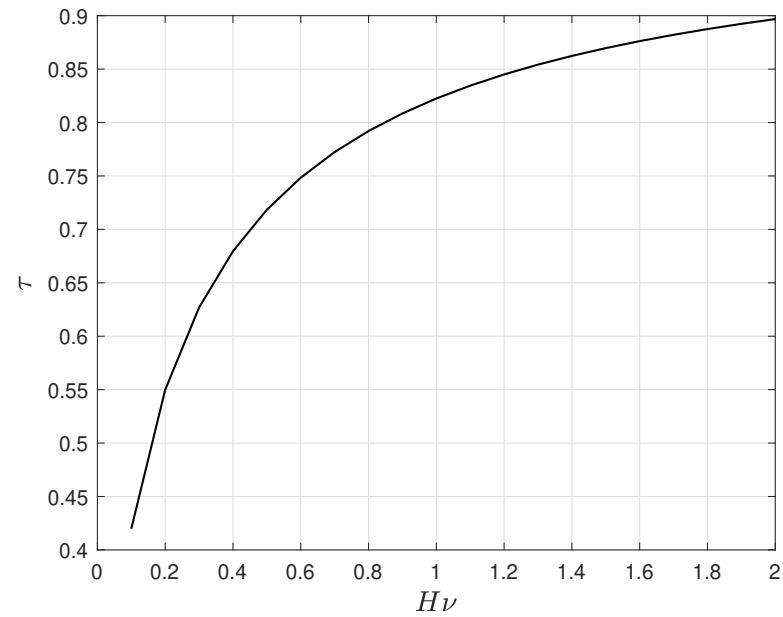
- Unconditioning

$$\tau = \frac{\mathbf{H}\nu}{\mathbf{6}\rho} \left(\pi^2 + \mathbf{6} \log\left(\frac{\rho}{\mathbf{H}\nu}\right) \log\left(1 + \frac{\rho}{\mathbf{H}\nu}\right) - \mathbf{3} \left(\log\left(1 + \frac{\rho}{\mathbf{H}\nu}\right) \right)^2 - \mathbf{6} \text{Li}_2\left(\frac{\mathbf{H}\nu}{\mathbf{H}\nu + \rho}\right) \right)$$

where $\text{Li}_n(\mathbf{z})$ is the polylogarithm function

$$\text{Li}_n(\mathbf{z}) = \sum_{k=1}^{\infty} \frac{\mathbf{z}^k}{k^n}$$

NUMERICAL RESULTS



τ as a function of $H\nu$ for $\lambda, \mu = 1$

EXAMPLES: REAL ENVIRONMENTS

	Case 1 (Dense urban)	Case 2 (Urban)	Case 3 (Suburban)
$\lambda(m^{-1})$	0.012	0.007	0.001
$\mu(m^{-1})$	0.02	0.02	0.02
$\mathbb{E}[\theta]$ (rad)	0.7732	0.5935	0.1695
$\mathbb{E}[\theta^I]$ (rad)	0.1956	0.1256	0.0195
$\mathbb{E}[\theta^R]$ (rad)	0.2214	0.1453	0.0201
$\mathbb{E}[l]$ (HAP) (m)	1.68×10^4	2.89×10^4	2.02×10^5
$\mathbb{E}[l]$ (Satellite) (m)	8.34×10^5	1.43×10^6	1.00×10^7
τ_H (HAP)	0.797838	0.864512	0.976052
τ_H (Satellite)	0.893742	0.935147	0.989409

Numerical values of the visibility for two cities

ONGOING/FUTURE RESEARCH

- 3D
- Multi Hop
- Refined statistics
- SINR coverage
- Economic model

RELATED PUBLICATIONS

- **Performance Analysis of RIS-assisted MIMO-OFDM Cellular Networks Based on Matern Cluster Processes**
G. Sun, F. Baccelli, K. Feng, L. G. Uzeda Garcia, S. Paris
[CoRR abs/2310.06754](#), 2024
- **Cox Point Processes for Multi-Altitude LEO Satellite Networks**
C.S. Choi, F. Baccelli, [IEEE Trans. Veh. Technol.](#), 2024.
- **How Much Can Reconfigurable Intelligent Surfaces Augment Sky Visibility: A Stochastic Geometry Approach**
J. Lee, F. Baccelli, to appear in [Trans. Wir. Comm.](#), 2024.

THANKS FOR YOUR ATTENTION

