Directional Maximum Likelihood Self-Estimation of the Path-Loss Exponent

Objectives

1 Estimation of PLE in case of clustered nodes. 2 Maximum likelihood (ML) solution, using the distribution of RSS.

Oerive Cramér-Rao lower bound (CRLB).

Introduction

- The path-loss exponent (PLE) is very **crucial** for efficiently designing wireless communications and networking systems.
- Most existing methods for estimating the PLE:
- require nodes with **known locations**;
- assume **an omni-directional PLE**;
- change some **configurations** of the receiver;
- require some **information of the network**, e.g., the node density.
- Our previous work [1]:
- two (weighted) total least squares solutions;
- Simple, pervasive, local, sole, collective, secure and **directional** estimation for the PLE.
- However, the remaining problems:
- can **NOT** cope with **clustered nodes**.
- **NOT** the **ML** solution.
- obtaining the **CRLB** is **NOT** possible.
- In \mathbb{R}^m , the RSS with only geometric path loss

$$P_r = Cr^{-\gamma},$$

- C is the constant including the transmit power P_t
- r is the nodal distance
- γ is the PLE



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Pareto distribution for RSS

RSS follows a truncated $Pareto$ distribution Type	In I
I $ \mathbb{P}(P_r m, \gamma, P_{r,min}, P_{r,max}) = \begin{cases} \frac{m}{\gamma} \frac{P_{r,min}^{m/\gamma} P_r^{-m/\gamma-1}}{1 - (P_{r,min}/P_{r,max})^{m/\gamma}}, & P_r \in [P_{r,min}, P_{r,max}], \\ 0, & \text{otherwise}, \end{cases} $	ın square error (MSE)
• $P_{r,min} \triangleq Cr_{max}^{-\gamma}$	Mea
• $P_{r,max} \triangleq \min\{Cr_{min}^{-\gamma}, P_t\}$	
Accordingly, the CRLB is $CRLB(\gamma) = \frac{1}{\mathcal{I}(\gamma)}$.	
$\mathcal{I}(\gamma) = -\frac{n}{\gamma^2} - \frac{2mnln(P_{r,min})}{\gamma^3} + \frac{2n[(\gamma + m)]}{\gamma^3}$	$\frac{nln(F)}{r}$
$+ \frac{nm(\frac{P_{r,min}}{P_{r,max}})}{2}$	$\frac{m}{\gamma}ln(\frac{H}{H})$

Important Result

• The RSS based on only a geometric path-loss is first found to follow a truncated *Pareto* distribution. ² The CRLB is introduced and decreases with more samples, a small PLE and a close node cluster. ³ Two ML self-estimators of the PLE are derived and yield a good performance close to the CRLB.

ML solutions

• $P_{r,min}$ and $P_{r,max}$ are known: solve

$$\frac{n\gamma}{m} - \sum_{i=1}^{n} \left(ln \frac{P_i}{P_{r,min}} \right) + \frac{n \left(\frac{P_{r,min}}{P_{r,max}}\right)^{m/\gamma} ln \left(\frac{P_{r,min}}{P_{r,max}}\right)}{1 - \left(\frac{P_{r,min}}{P_{r,max}}\right)^{m/\gamma}} = 0.$$

• $P_{r,min}$ and $P_{r,max}$ are unknown: rank the RSSs as $P_{(1)} < \cdots < P_{(n)}$ and solve

$$\frac{n\gamma}{m} - \sum_{i=1}^{n} \left(ln \frac{P_i}{P_{(1)}} \right) + \frac{n \left(\frac{P_{(1)}}{P_{(n)}}\right)^{m/\gamma} ln \left(\frac{P_{(1)}}{P_{(n)}}\right)}{1 - \left(\frac{P_{(1)}}{P_{(n)}}\right)^{m/\gamma}} = 0.$$

- Solve by a simple bisection method.
- Shadowing effect, if considered, has an insignificant influence.







Node Cluster





Homogeneous random networks

[1] Yongchang Hu and G. Leus.

Self-estimation of path-loss exponent in wireless networks and applications.

IEEE Transactions on Vehicular Technology, 64(11):5091-5102, Nov 2015.

[2] Sunil Srinivasa and Martin Haenggi.

Path loss exponent estimation in large wireless networks. In Information Theory and Applications Workshop, 2009, pages 124–129. IEEE, 2009.

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