

# Generalised trade-off model for energy-efficient WSN synchronisation

P. Briff<sup>✉</sup>, A. Lutenberg, L. Rey Vega, F. Vargas and M. Patwary

A mathematical framework to obtain a generalised energy-efficient trade-off model for generic wireless sensor networks (WSNs) while attaining sensing node synchronisation at a given network-wide estimation error threshold is presented. The model outputs both a theoretical optimal solution as well as a set of sub-optimal solutions to cater for real-world WSN designs. The robustness of the proposed framework is examined with an example in which randomly deployed sensors are affected by path-loss and uncorrelated Rayleigh fading effects.

**Introduction:** Time synchronisation has become a major feature within application-specific wireless sensor networks (WSNs) to meet application demands such as data fusion, energy management and collision avoidance mechanisms. The need for energy-efficient time synchronisation has attracted intense research focus of recent years. A recent survey published in [1] details the challenges in unattended WSNs and the inevitable need for energy-efficient routing and data processing algorithms, although without mentioning the existence of a network-wide energy-efficient solution with regard to estimation error. Basagni *et al.* [2] highlight the underlying trade-off between consumed energy and data latency, and propose an energy harvesting technique to alleviate the application constraints. Lenzen *et al.* [3] have updated the PulseSync Protocol with further considerations of energy efficiency, message latency and estimation error. However, the solutions proposed in [2, 3] have not considered optimal solutions for application-specific resource trade-off while attaining a given estimation error. In this Letter, we present a generalised model providing optimal solutions with a wide range of tunability for the energy-efficient synchronisation within WSN nodes, which has not been yet studied to the best our knowledge.

**Model statement:** Consider the WSN depicted in Fig. 1 consisting of  $N$  number of sensor nodes randomly deployed within a given area. Assume that all nodes in the WSN under consideration are identical in their properties and affected by equal noise power levels within the network, which is a widely adopted assumption in the literature [4, p. 37]. Let two nodes  $u_i$  and  $u_j$  be located at spatial positions  $\{\mathbf{x}_i, \mathbf{x}_j\} \in \mathbb{R}^3$ , respectively, separated by the Euclidean distance  $d_{ij} \triangleq \|\mathbf{x}_i - \mathbf{x}_j\|_2$ . Let  $D_{ij}$  denote node  $u_i$ 's maximum coverage radius for communicating with node  $u_j$ . Furthermore, let  $S_{ij}$  denote the transmit power emitted from node  $u_i$  when communicating with node  $u_j$ , and let  $\gamma_0$  denote the receiver sensitivity of node  $u_j$ . The measure of  $D_{ij}$  is determined by  $S_{ij}$  and  $\gamma_0$ , as well as the channel condition between nodes  $u_i$  and  $u_j$ . Node  $u_i$  is said to be 'connected with' node  $u_j$  if  $D_i \geq d_{ij}$ . If two nodes  $u_i$  and  $u_j$  are connected, throughout the rest of this Letter they are to be known as 'neighbours'. A WSN can be represented by a directed graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  with vertices  $\mathcal{V} = \{1, 2, \dots, N\}$ , a set of edges  $\mathcal{E} \subset \mathcal{V} \times \mathcal{V}$  and weighted adjacency matrix  $\mathbf{A} \in \mathbb{R}^{N \times N}$  with an amplitude set  $\mathbf{A} = \{a_{ij}\}$ , where the coefficient  $a_{ij}$  represents the connection weight between nodes  $u_i$  and  $u_j$ , with  $\{i, j\} \in \mathcal{V}$ . To account for the spatial attenuation of transmitted signals, coefficient  $a_{ij}$  is defined as follows:  $a_{ij} = 1$  if  $i = j$ ,  $a_{ij} = (d_0/d_{ij})^n$  if  $d_{ij} \leq D_{ij}$  or  $a_{ij} = 0$  otherwise. Parameter  $d_0 (< d_{ij})$  is a reference distance, whereas  $n$  denotes the path-loss exponent, which ranges from  $n = 2-6$  [5, p. 41].

Consider the situation in which node  $u_j$  estimates  $u_i$ 's clock offset, denoted by  $\theta_{ij}$ , by means of one-way message exchange, as in the Flooding Time Synchronization Protocol [6]. At a given time instant, node  $u_i$  transmits a message modulated in signal  $q_{ij}(t)$  over a flat-fading channel with gain  $g_{ij}(t)$ , and node  $u_j$  receives  $y_{ij}(t) = g_{ij}(t)q_{ij}(t) + w(t)$ , where  $w(t)$  is an additive white Gaussian noise process with power  $\sigma^2$ . Signal  $q_{ij}$  is received by node  $u_j$  with a probability of success determined by the channel's outage probability. More precisely, node  $u_i$  sends  $m_{ij}$  number of messages to  $u_j$ , which successfully receives  $M_{ij}$  messages, where  $M_{ij} \leq m_{ij}$ . The expectation of  $M_{ij}$ , denoted as  $\tilde{m}_{ij}$ , can be obtained as  $\tilde{m}_{ij} = m_{ij} \cdot (1 - P_{out_{ij}})$  [7], where  $P_{out_{ij}}$  denotes the outage probability of the wireless link. In general, the outage probability is a function  $S_{ij}, g_{ij}, \gamma_0, \sigma^2$  at the receiver,  $a_{ij}$  and the relative velocity between sensors  $v_{ij}$ . For Cramer-Rao efficient estimators of  $\theta_{ij}$ , the estimation error incurred by node  $u_j$  when estimating  $u_i$ 's clock offset,

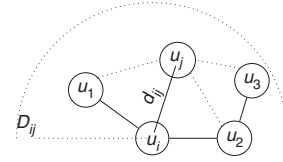
denoted as  $\epsilon_{ij}$ , is given by  $\epsilon_{ij} = \sigma_V^2 / \tilde{m}_{ij}$  [7], where  $\sigma_V^2$  is the variance of the measured  $\theta_{ij}$ . Thus, the total local estimation error on node  $u_i$  when estimating its neighbours' clock offsets, denoted as  $\epsilon_i$ , is

$$\epsilon_i \triangleq \sum_j \epsilon_{ij} = \sum_j \frac{\sigma_V^2}{m_{ij}(1 - P_{out_{ij}})} \quad \forall j \in \mathcal{V}, \quad j \neq i \quad (1)$$

**Energy optimisation against network-wide estimation error:** Node  $u_i$ 's pairwise synchronisation energy, defined as the local average energy function of node  $u_i$  when synchronising with  $u_j$ , is dictated by  $E_{ij} = S_{ij}m_{ij}\delta_{ij}$  [7], where  $\delta_{ij} = T_m/(1 - P_{out_{ij}})$  represents each message's average delivery time and  $T_m$  is each message's time duration. In general, node  $u_i$  shall synchronise with  $(N-1)$  nodes, consuming a total synchronisation energy, denoted as  $E_i$ , equal to  $E_i \triangleq \sum_j E_{ij} \quad \forall j \in \mathcal{V}, \quad j \neq i$ . Energy being a scarce resource,  $E_i$  is to be minimised. Moreover, it is required to guarantee that network-wide synchronisation is achieved within a maximum tolerable threshold, denoted by  $\epsilon_{max}$ . Hence, the following problem can be stated:

$$\text{Minimise } E_i \text{ s.t. } \sum_i \epsilon_i \leq \epsilon_{max} \quad \forall i \in \mathcal{V} \quad (2)$$

The objective is then to find the optimal pairs  $\{S_{ij}, \epsilon_{ij}\} \forall i, j \in \mathcal{V}$  that solve for the optimisation problem expressed in (2).



**Fig. 1** Node  $u_i$  is connected to node  $u_j$  since  $d_{ij} < D_{ij}$

**Tight, optimal solution:** An optimal solution for equality in (2) can be found using the Lagrange multipliers method, by stating

$$\nabla E_i = \lambda \nabla \left( \sum_i \epsilon_i \right) \quad \text{s.t.} \quad \sum_i \epsilon_i = \epsilon_{max} \quad \forall i \in \mathcal{V} \quad (3)$$

where  $\lambda \in \mathbb{R}$  is the Lagrange multiplier and  $\nabla(\cdot)$  is calculated with respect to all the variables composing the energy function and for the  $(N-1)$  neighbours of node  $u_i$ . Parameters  $a_{ij}, \gamma_0, \sigma^2$  and  $v_{ij}$  are assumed constant and known. Therefore, it holds that  $\epsilon_{ij} \triangleq \epsilon_{ij}(S_{ij}), P_{out_{ij}} \triangleq P_{out_{ij}}(S_{ij}), P'_{out_{ij}} \triangleq dP_{out_{ij}}/dS_{ij}$  and  $E'_{ij} \triangleq dE_{ij}/dS_{ij}$ . Thus, (3) implies minimisation of each component of  $E_i$ , leading to

$$E'_{ij} = \lambda \epsilon'_{ij} \quad (4)$$

$$E'_{ij} = \frac{\sigma_V^2 T_m}{\epsilon_{ij}(1 - P_{out_{ij}})^2} \left( 1 - \frac{\epsilon'_{ij} S_{ij}}{\epsilon_{ij}} + \frac{2S_{ij} P'_{out_{ij}}}{1 - P_{out_{ij}}} \right) \quad (5)$$

Equation (4) implies terms in (5) not containing  $\epsilon'_{ij}$  vanish, hence

$$2S_{ij} P'_{out_{ij}} = 1 - P_{out_{ij}} \quad (6)$$

Let  $S_{ij, opt}$  denote the optimal transmit power that solves (6). Hence, from (4) and (5) it follows that the optimal pairwise estimation error for the link between nodes  $u_i$  and  $u_j$ , denoted by  $\epsilon_{ij, opt}$ , is given by:

$$\epsilon_{ij, opt} = \sqrt{-\frac{S_{ij, opt}}{\lambda(1 - P_{out_{ij}}(S_{ij, opt}))^2 \sigma_V^2 T_m}} \quad (7)$$

The value of  $\lambda$  in (7) is obtained by plugging each  $\epsilon_{ij, opt}$  into the summation constraint in (3) and solving for  $\lambda$ , namely

$$\lambda = -\frac{1}{\epsilon_{max}^2} \left( \sum_i \sum_{j, j \neq i} \sqrt{\frac{S_{ij, opt}}{(1 - P_{out_{ij}}(S_{ij, opt}))^2 \sigma_V^2 T_m}} \right)^2 \quad (8)$$

Note in (8) that  $\lambda < 0$  for all non-zero  $S_{ij, opt}$ , which indicates that the optimal solutions of (3) can only be energy minima.

**Flexible, sub-optimal solution:** In practice, WSN sensors may be unable to tune their transceivers to the exact value of  $S_{ij, opt}$ . Therefore, it is useful to find flexible solutions suitable for different realistic scenarios. Namely, inequality constraints in (2) can be attained at a

predefined constant  $E_{ij}$ , which implies  $\nabla E_i = 0$  over the set of solutions  $\{S_{ij}, \epsilon_{ij}(S_{ij})\}$  for all  $i, j \in \mathcal{V}$ , or

$$E'_{ij} = \frac{\partial E_{ij}}{\partial S_{ij}} + \frac{\partial E_{ij}}{\partial \epsilon_{ij}} \epsilon'_{ij} = 0 \quad (9)$$

After simple mathematical operations, (9) leads to

$$\epsilon'_{ij} = \frac{1}{S_{ij}} + \frac{2P'_{\text{out}_{ij}}}{1 - P_{\text{out}_{ij}}} \epsilon_{ij}, \quad \epsilon_{ij}(S_{ij_{\text{opt}}}) = \xi_{ij} \epsilon_{ij_{\text{opt}}} \quad (10)$$

where  $S_{ij_{\text{opt}}}$  is a reference transmit power that sets the initial condition of the estimation error, given by  $\epsilon_{ij_{\text{opt}}} \xi_{ij}$ . Using separation of variables, system (10) solves to

$$\epsilon_{ij}(S_{ij}) = \epsilon_{ij_{\text{opt}}} \xi_{ij} \frac{S_{ij}}{S_{ij_{\text{opt}}}} \left[ \frac{1 - P_{\text{out}_{ij}}(S_{ij_{\text{opt}}})}{1 - P_{\text{out}_{ij}}(S_{ij})} \right]^2 \quad (11)$$

where  $\xi_{ij}$  is a degree of freedom of the energy-efficient pairwise estimation error, in the range  $\xi \in (0, 1]$ . To ensure the fulfilment of constraint (2),  $\epsilon_{ij}(S_{ij})$  must be capped to  $\epsilon_{ij_{\text{opt}}}$ . Note that  $\xi_{ij} = 1$  corresponds to the optimal solution found by the Lagrange method, whereas decreasing  $\xi_{ij}$  allows variation of  $S_{ij}$  and  $\epsilon_{ij}$  in an amplitude range determined by  $\xi_{ij}$ , at fixed synchronisation energy.

*Application example:* Consider the WSN in Fig. 1 with  $N=5$  nodes designated  $u_i$ , with  $i \in \mathcal{V} = \{1, 2, \dots, N\}$ , randomly deployed in a sphere of radius  $r = 6$  m. Nodes' position tolerances are modelled by a zero-mean normal distribution with standard deviation  $\sigma_d = 0.03$  m. Assume that nodes communicate over a channel that experiences combined path-loss and uncorrelated Rayleigh fading effects. Using the simplified path-loss model, the power received by sensor  $u_j$  from transmitter sensor  $u_i$ , denoted as  $S_{ijR}$ , is equal to  $S_{ijR} = Ka_{ij}S_{ij}$ , where  $K$  is a unitless constant, and the outage probability is given by [5, p. 169]

$$P_{\text{out}_{ij}}(S_{ij}) = 1 - \exp\left(-\frac{\gamma_0}{\gamma S_{ij}}\right) \approx 1 - \exp\left(-\frac{\gamma_0 \sigma^2}{Ka_{ij}S_{ij}}\right) \quad (12)$$

where  $\gamma_0$  represents the minimum acceptable signal-to-noise ratio. The transmit power  $S_{ij}$  and the noise power  $\sigma^2$  in (12) are expressed in watts. Power from interferer sources is within parameter  $\sigma^2$ . Plugging (12) into (6) leads to  $S_{ij_{\text{opt}}} = 2\sigma^2 \gamma_0 / (Ka_{ij})$ . Hence, the value of each  $\epsilon_{ij_{\text{opt}}}$  is obtained by plugging  $S_{ij_{\text{opt}}}$  into (7) and (8).

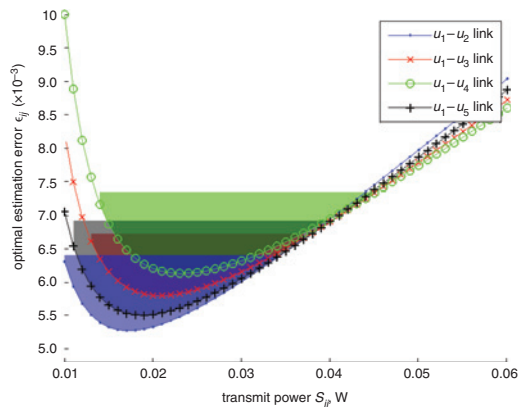


Fig. 2 Optimal estimation error between node  $u_1$  and its neighbours

Fig. 2 shows the Monte Carlo simulation outputs of pairwise estimation errors for node  $u_1$  when synchronising nodes  $u_2$  to  $u_5$  using MATLAB<sup>®</sup>. The shaded areas show the region in which each pairwise estimation error can be freely varied, through parameter  $\xi_{ij}$ , without compromising the network-wide estimation error. The region of freedom is upper bounded by  $\epsilon_{ij_{\text{opt}}}$ . Also, note that the pairwise estimation error reaches a minimum at  $S_{ij_{\text{opt}}}$ . For constant  $S_{ij}$ , the required synchronisation energy can be further reduced at the expense of  $\epsilon_{ij}$ . Fig. 3 depicts the pairwise synchronisation energy against transmit

power and estimation error. For the sake of clarity, and without loss of generality, only the energy for pair  $u_1 - u_3$  is drawn. Note that, for a given  $\xi_{ij}$ , the synchronisation energy remains constant along the solutions of (11), which is observed as level curves of the synchronisation energy.

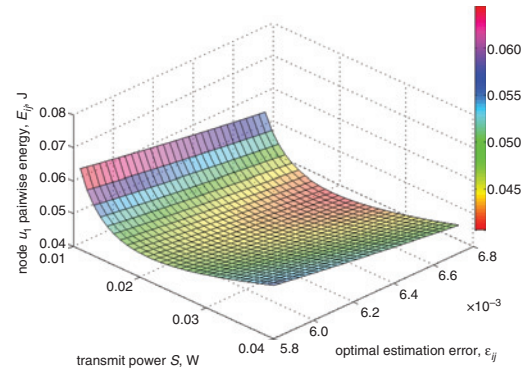


Fig. 3 Synchronisation energy level curves for link  $u_1 - u_3$

*Conclusion:* The generalised model presented in this Letter contemplates the main parameters that drive energy-efficient synchronisation within a WSN. The proposed solutions, supported by simulation results, show how to fine tune both pairwise transmit power and local estimation error for each node in the network, in order to meet the expected network-wide estimation error requirement in an energy-efficient manner. More importantly, the system model is based on a generic WSN, without being constrained by network size or topology, which makes it a powerful tool for energy-aware WSN design.

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One or more of the Figures in this Letter are available in colour online.

P. Briff (*LSE, FIUBA, University of Buenos Aires, Buenos Aires, Argentina*)

✉ E-mail: pbriff@fi.uba.ar

A. Lutenberg (*LSE, FIUBA and CONICET, Buenos Aires, Argentina*)

L. Rey Vega (*LPSC, FIUBA, and CSC-CONICET, Buenos Aires, Argentina*)

F. Vargas (*Catholic University – PUCRS, Porto Alegre, Brazil*)

M. Patwary (*SPCR Laboratory, FCES, Staffordshire University, Stafford, United Kingdom*)

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