
Maximum Lifetime Routing in Wireless Ad Hoc networks I - Theory

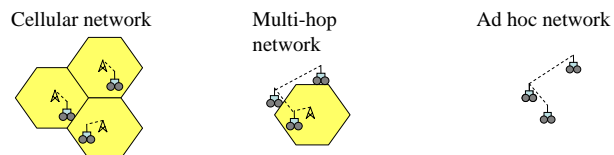
Riku Jäntti
University of Vaasa, Finland

Contents

- Introduction
- System model
- Maximum life-time routing problem
- Iterative algorithm
- Numerical example
- Effect of power sensitive topology
- Effect of SIR to rate mapping
- Conclusions

Wireless Ad Hoc Networks

- Some of the transceivers (nodes) act as routers
- The network is formed by means of self-configuration without any pre-existing infrastructure
- Network control is distributed among the nodes, no central control unit is needed

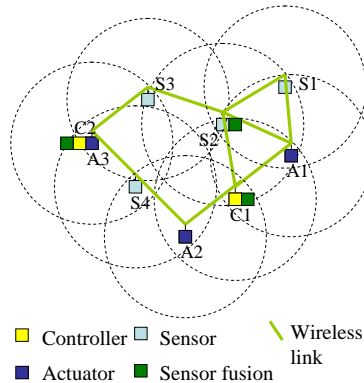


Wireless Ad Hoc Networks

Radio resource management in interference limited ad hoc networks (Academy of Finland)

- Research problems
 - Accurate modeling of co-channel interference
 - Power control, topology control
 - Data rate control, link adaptation
 - Routing
 - Multi-hop 'virtual' cellular networks

Sensor and actuator networks



ERHE Project (TEKES ÄLY)

- Communication
 - Radio technology
 - Medium access
 - Networking and routing
 - Security
- Sensor fusion
 - Filtering sensor data
 - Detecting sensor faults
 - Sensor calibration
- Control
 - Compensation of communication delays
 - Coordination of control loops (hierarchical control)



Ad Hoc Networks

- Benefits
 - No need for separate base stations or the number of them could at least be decreased
 - Easy to deploy, no wiring required
 - Reconfigurable: Network can quickly adapt to topology changes
 - Can be utilized in an areas where infrastructure doesn't exists
 - Robust: Break down of single network node doesn't prevent networking
- Drawbacks
 - Distributed operation => difficult to control
 - Lower capacity, higher packet delay and more severe jitter than in cellular/infrastructure networks
 - Nodes are either battery operated or use energy scavenging => Utilized protocols should be energy efficient
 - Network maintenance can be expensive (e.g. change of batteries etc.) => Redundancy is required



Sensor network

- Sensor network is an ad hoc network, in which the nodes are sensor devices
 - Low computation power, small memory space
 - Energy consumption is critical: If nodes are running on batteries their operation time depends heavily on the energy efficiency of the data transmission.
 - Node size and cost are critical: Nodes should be cheap enough, so that they could be deployed in large quantities to obtain both coverage and redundancy
 - Data rates are small: Size of single measurement data unit is only few bytes; data rates 1 ... 100 kbytes per second
 - Number of nodes is large: Ad hoc data network ~10 nodes, sensor network ~1000 nodes.



Applications

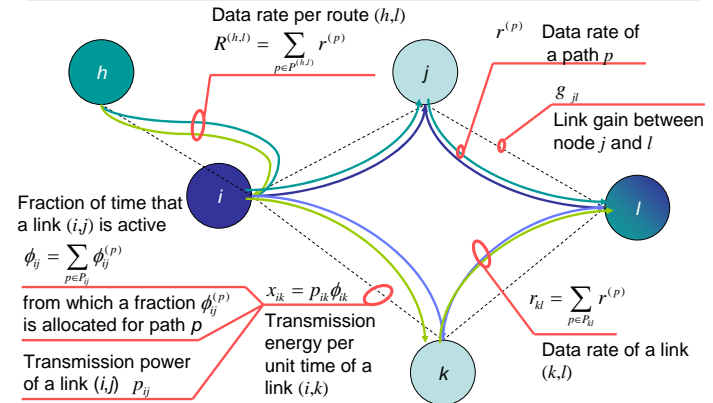
- Mobile ad hoc networks (MANETS)
 - Ambient intelligence
 - Military communications
 - Personal area networks
 - Peer-to-peer networking: CSCW, mobile gaming etc.
 - Multi-hop extensions to cellular systems
- Wireless sensor and actuator networks
 - Ambient intelligence
 - Military sensor networks
 - Healthcare and emergency response units
 - Logistics
 - Automation and control
 - Agriculture and forestry
 - Research



Objective

- Maximize the operation time of the worst node in the network by jointly controlling
 - Transmit power
 - Scheduling
 - Data rate
 - Routing

System model (1/3)



System model (2/3)

DSSS waveforms

- Instantaneous data rate

$$r_{ij}(t) = \frac{W}{\Gamma_{ij}} \frac{g_{ij}(t)p_{ij}(t)}{\sum_{k \neq i} \sum_{k \neq j} \left[g_{kj}(t) \sum_{l \neq k} \sum_{l \neq j} p_{kl}(t) \right] + \nu_j}$$

- Transmitter power constraint

$$p_{ij}(t) = \begin{cases} p_{ij} & \text{if the link } (i,j) \text{ is active} \\ 0 & \text{otherwise} \end{cases}$$

$$0 \leq p_{ij} \leq p^{\max}$$

- Peak rate constraint

$$0 \leq r_{ij}(t) \leq r^{\max}$$

Snapshot assumption

- Average data rate

$$r_{ij} = \phi_{ij} \min \left\{ r_{\max}, \frac{W}{\Gamma} \frac{g_{ij} p_{ij}}{\sum_{k \neq i} g_{kj} \sum_{l \neq k} \sum_{l \neq j} \phi_{kl} p_{kl} + \nu_j} \right\}$$

- Transmission time constraint

$$\phi_{ij} \geq \frac{r_{ij}}{r_{\max}} \stackrel{\text{def}}{=} \phi_{ij}^{\min}$$

ν_j Noise power
 W Chip rate
 Γ_{ij} SIR-target

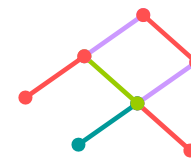
System model (3/3)

- Node cannot transmit and receive simultaneously

$$\sum_j \left(\sum_{p \in P_{ij}} \phi_{ij}^{(p)} + \sum_{p \in P_{ji}} \phi_{ji}^{(p)} \right) \leq 1$$

- "Edge coloring" cost

$$\sum_j \left(\sum_{p \in P_{ij}} \phi_{ij}^{(p)} + \sum_{p \in P_{ji}} \phi_{ji}^{(p)} \right) \leq \delta < 1$$



Maximum lifetime routing problem (1/2)

- Lifetime of a node i

$$\tau_i = \frac{E_i}{\sum_{j \neq i} \phi_{ij}(P_t + p_{ij}) + \sum_{j \neq i} \phi_{ji}P_r + P_s}$$

E_i Battery energy
 P_t Transmitter base band power consumption
 p_{ij} Transmitter RF-power
 P_r Receiver power consumption
 P_s Terminal CPU power consumption

- Objective function

$$f(\phi, p) = \min_i \{\tau_i\} \Leftrightarrow F(\phi, p) = \max_i \left\{ \frac{1}{\tau_i} \right\}$$



Maximum lifetime routing problem (2/2)

$$\min_{\{\phi, p\}} \max_i \left\{ \frac{1}{E_i} \left(\sum_{j \neq i} \left(\sum_{p \in P_{ij}} \phi_{ij}^{(p)} (P_t + p_{ij}) + \sum_{p \in P_{ji}} \phi_{ji}^{(p)} P_r \right) + P_s \right) \right\}$$

Subject to

- Data rate constraints

$$r^{(p)} = \frac{W}{\Gamma^{(p)}} \frac{g_{ij} P_{ij} \phi_{ij}^{(p)}}{\sum_{\substack{k \neq i \\ k \neq j}} g_{kj} \left[\sum_{\substack{l \neq k \\ l \neq j}} \sum_{q \in P_{kl}} P_{kl} \phi_{kl}^{(q)} \right] + \nu_j} \quad \sum_{p \in P^{(i,j)}} r^{(p)} \geq R^{(i,j)}$$

- Transmission time constraints

$$\sum_j \left(\sum_{p \in P_{ij}} \phi_{ij}^{(p)} + \sum_{p \in P_{ji}} \phi_{ji}^{(p)} \right) \leq 1 \quad \phi_{ij}^{(p)} \geq 0 \quad \phi_{ij} = \sum_{p \in P_{ij}} \phi_{ij}^{(p)} \geq \phi_{ij}^{\min}$$

- Transmission power constraint

$$0 \leq p_{ij} \leq p_{\max}$$



Transmission energy allocation

- For a given rate to path allocation $R = (r^{(p)})$, the transmission energy can be written as

$$x_{ij} = \frac{\sum_{q \in P_{ij}} r^{(q)} \Gamma^{(q)}}{W} \left(\sum_{\substack{k \neq i \\ k \neq j}} \sum_{\substack{l \neq k \\ l \neq j}} \frac{g_{kl}}{g_{ij}} x_{kl} + \frac{\nu_j}{g_{ij}} \right), \quad x_{ij} = p_{ij} \phi_{ij}$$

- By ordering the links, the energy can be solved using a simple Jacobi iteration (distributed power control algorithm)

$$x_i^{(n+1)} = \xi_i(R) \left(\sum_j h_{ij} x_j^{(n)} + \eta_i \right) \stackrel{\text{def}}{=} I_i(R, X^{(n)}), \quad X = (x_{a,b})$$

$$\xi_i(R) = \frac{\sum_{q \in P_{a,b}} r^{(q)} \Gamma^{(q)}}{W}$$



Joint energy, time, and power allocation

- Power and peak rate constraints

$$p_{ij} \leq p^{\max} \Leftrightarrow \phi_{ij} \geq \frac{x_{ij}}{p^{\max}}$$

$$r_j(t) \leq r^{\max} \Rightarrow \phi_{ij} \geq \phi_{ij}^{\min}$$

- JETP algorithm

$$\begin{aligned}
 x_i^{(n+1)} &= I_i(R, X^{(n)}) \\
 \phi_i^{(n+1)} &= \max \left\{ \frac{x_i^{(n+1)}}{p^{\max}}, \phi_i^{\min} \right\} \\
 P_i^{(n+1)} &= \frac{x_i^{(n+1)}}{\phi_i^{(n+1)}}
 \end{aligned}$$

can be shown to be converge for any feasible rate $R = (r^{(p)})$ allocation from any nonnegative initial value $X^{(0)} \geq 0$ (Standard interference function in the sense of Yates)



Transmission rate allocation

- For fixed energy, time, and power allocation, the objective function can be written as

$$F(X, P) = \max_i \left\{ \frac{1}{E_i} \left(\sum_{j \neq i} \left(\left(1 + \frac{P_j}{P_i} \right) x_{ij} + \frac{P_j}{P_i} x_{ji} \right) + P_s \right) \right\}$$

By noting that

$$x_{ij} = \frac{\sum_{q \in \Gamma_i} r^{(q)} \Gamma^{(q)}}{W} \left(\sum_{\substack{k \neq j \\ k \neq i}} \frac{g_{kj}}{\theta_{ij}} x_{ki} + \frac{v_j}{\theta_{ij}} \right)$$

is a linear function of $R = (r^{(p)})$

=> Rate can be solved using linear programming



Iterative routing algorithm

- IR algorithm

1. Start with any feasible $(X^{(0)}, \Phi^{(0)}, P^{(0)}, R^{(0)})$.
2. Solve the following linear programming problem

$$R^{(n+1)} = \arg \max_{R \in \mathcal{R}} \{ F(\mathcal{I}(R, X^{(n)}), P^{(n)}) \}$$
3. For the obtained $R^{(n+1)}$, solve $(X^{(n+1)}, \Phi^{(n+1)}, P^{(n+1)})$ by applying JETP.
4. if $\|X^{(n+1)} - X^{(n)}\| < \epsilon$, then stop; otherwise, set $n = n + 1$ and goto Step 2

The above algorithm is a special case of nonlinear Gauss-Seidel algorithm. Since $F(X, P)$ can be shown to be concave for one of the variables at the time, the iteration is convergent.



Numerical example (1/4)

- System parameters

$W = 1$ MHz

$\Gamma = 5$ dB

$v_i = -174$ dBm/Hz

$P_t = 800$ mW

$2 \text{ mW} \leq p_{ij} \leq 1000$ mW

$P_r = 400$ mW

$P_s = 50$ mW

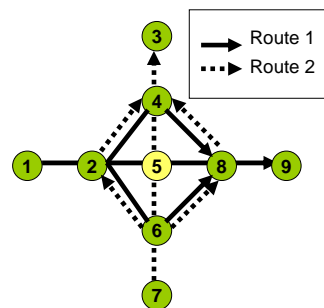
$E_i = E, i \neq 5 \quad E_5 = 0.1E$

User data rate:

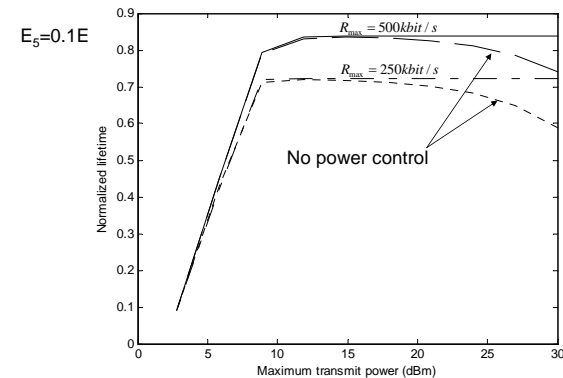
$R^{(1,9)} = R^{(7,3)} = 30$ kbit/s

Signalling data rate:

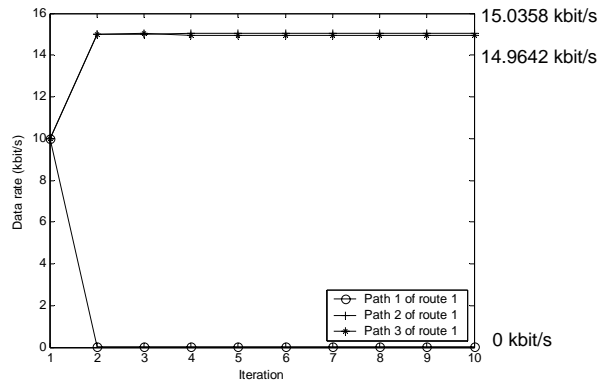
$R^{(i,j)} = 1$ kbit/s



Numerical example (2/4)



Numerical example (3/4)



Numerical example (4/4)

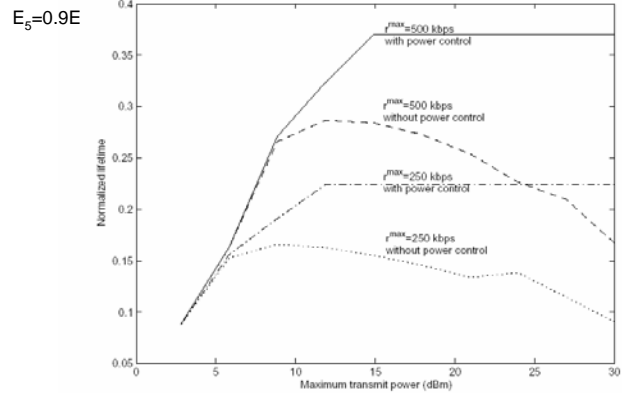
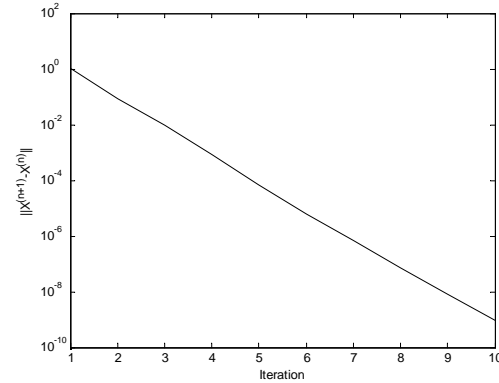


Figure 4: Normalized network lifetime
Normalized by $\min_i \{E_i\} / P_s$

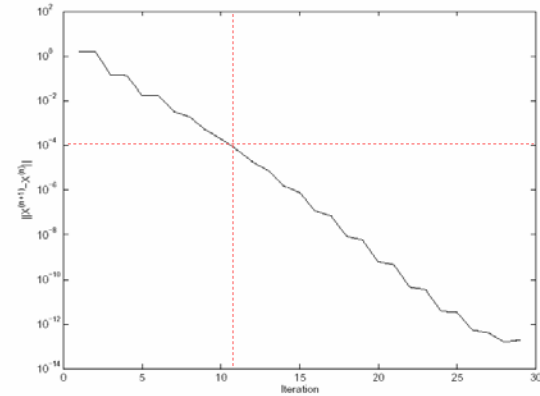


Figure 2: Convergence of X as a function of iterations

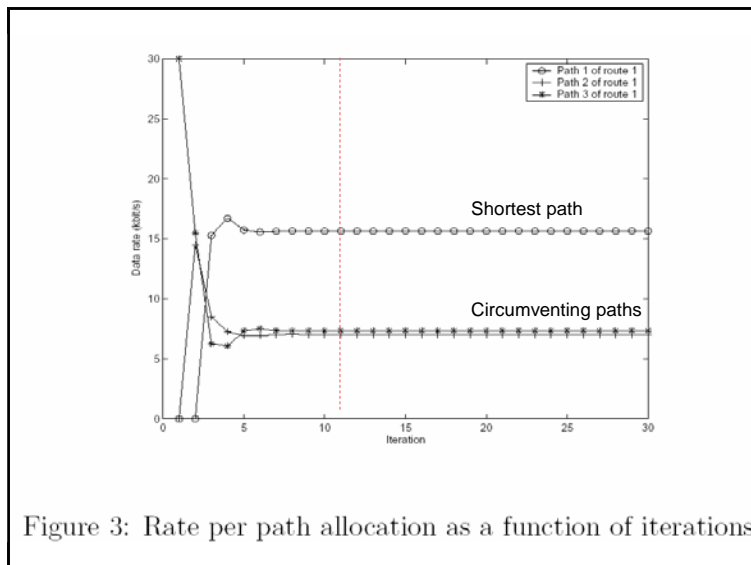


Figure 3: Rate per path allocation as a function of iterations

Power dependent topology

- Grid network
 - When power increases more links become available.
 - The longer the link, the higher the interference it causes and the lower the peak data rate.
 - Signaling over the long links consume considerable amount of resources even if no data is allocated to such links.

Signalling over the long links consume a lot of power even if they are not used for data transmission

University of Vaasa
Department of Computer Science

26

Logarithmic rate relation

- Data rate of a link

$$r_i = \phi_i W \log_2 \left(1 + \frac{1}{\sum_j h_{ij} \phi_j + \frac{\eta_i}{p_{max}}} \right)$$
- Rate control mapping

$$\phi_i^{(n+1)} = \zeta_i(R) \Psi \left(\sum_j h_{ij} \phi_j + \frac{\eta_i}{p_{max}} \right)$$

$$\zeta_i(R) = \frac{\sum_{q \in \mathcal{P}_{a_i b_i}} r^{(q)}}{W}$$

$$\Psi(x) = \frac{1}{\log_2(1 + x^{-1})}$$

$p_i = p_{max} \quad \forall i$ Fixed transmission power

University of Vaasa
Department of Computer Science

27

Logarithmic rate relation

- For sufficiently small R the mapping

$$J_i(R, \phi) = \zeta_i(R) \Psi \left(\sum_j h_{ij} \phi_j + \frac{\eta_i}{p_{max}} \right)$$
 is contraction.
 - It follows from the mean-value theorem that

$$|J_i(R, \phi) - J_i(R, \phi')| \leq \zeta_i(R) \Delta \left(\sum_j h_{ij} \phi_j - \phi'_j \right)$$

$$\Delta = \frac{d}{dx} \Psi(x) |_{x=x^*} = \Psi^2(x^*) \frac{x^*}{x^*(x^* + 1)}$$

$$x^* \in \left[\sum_j h_{ij} \phi_j + \frac{\eta_i}{p_{max}}, \sum_j h_{ij} \phi'_j + \frac{\eta_i}{p_{max}} \right]$$
 - Let $J=(J_i)$ and $Z(R)=\text{diag}(\zeta_i(R))$. It can be shown that there exists nonnegative diagonal weight matrix W such that

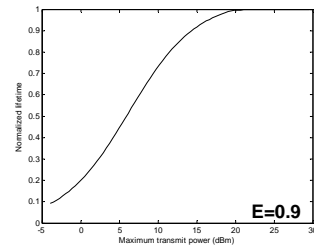
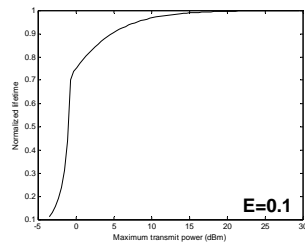
$$\|J(R, \phi) - J(R, \phi')\|_{\infty}^W \leq \Delta \rho(Z(R)H) \|\phi - \phi'\|_{\infty}^W$$

University of Vaasa
Department of Computer Science

28

Numerical example

- In both cases, the algorithm converges solutions that circumvent node 5.



Concluding remarks

- Our analysis suggest that if the rate is a linear function of the instantaneous SIR, then the lifetime of the worst node is a non-decreasing function of the maximum air interface power p^{max} .
- If link adaptation is used without power control, i.e. by setting $p_{ij} = p^{max} \forall i, j$, then there exists a threshold value after which any excess power affects the lifetime negatively.
- The routing problem can be solved using a simple iterative algorithm.
 - Energy, time, and power can be solved in a distributed fashion for each link separately using a scheme similar to traditional distributed power control
 - Rate allocation (routing) is based on linear programming and must be solved in centralized fashion.

