

# Mobility-aware Handover Strategies in Smart Cities

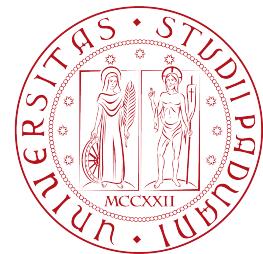
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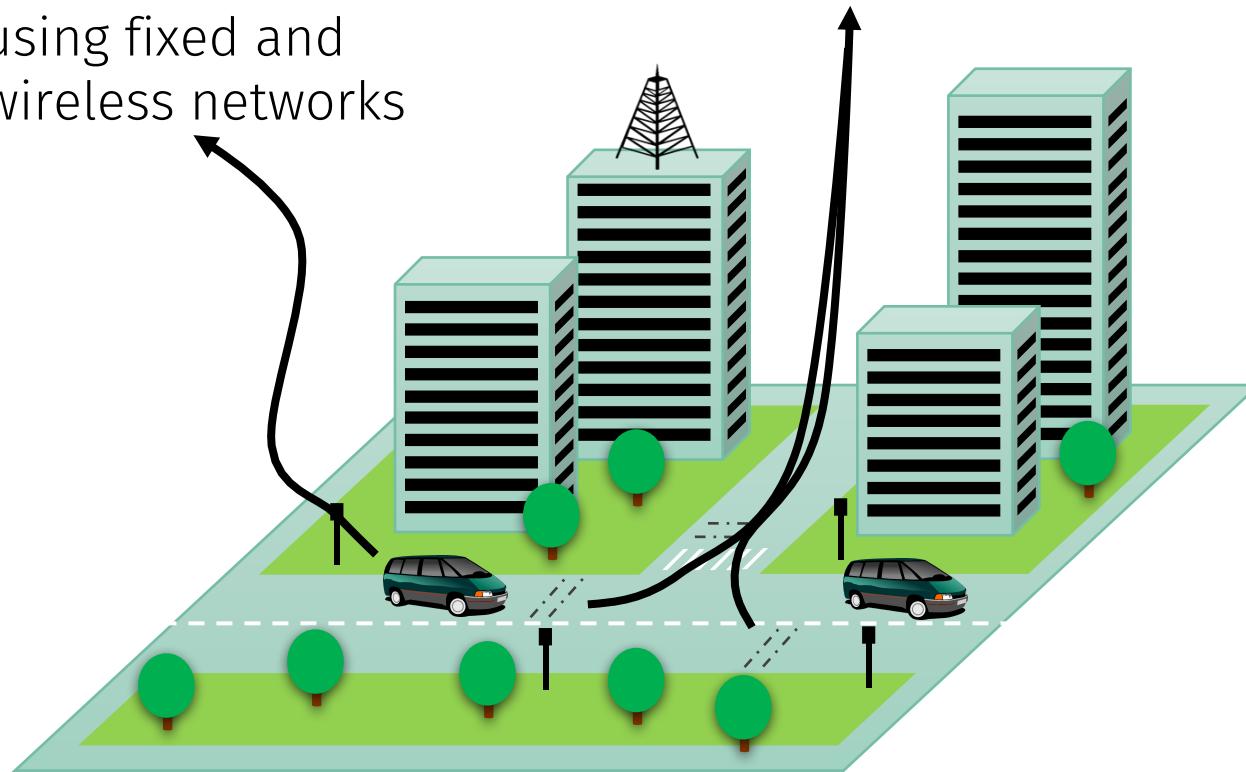
# Outline

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- Introduction
- Handover in HetNets
- Mobility data
- Asymmetrical Handover Bias
- Conclusions

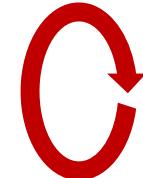
# IoT impacts the network

A Smart City gathers data from IoT sensors using fixed and wireless networks

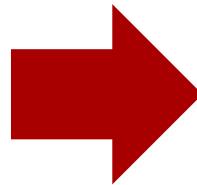


→ IoT introduces an additional load in the network

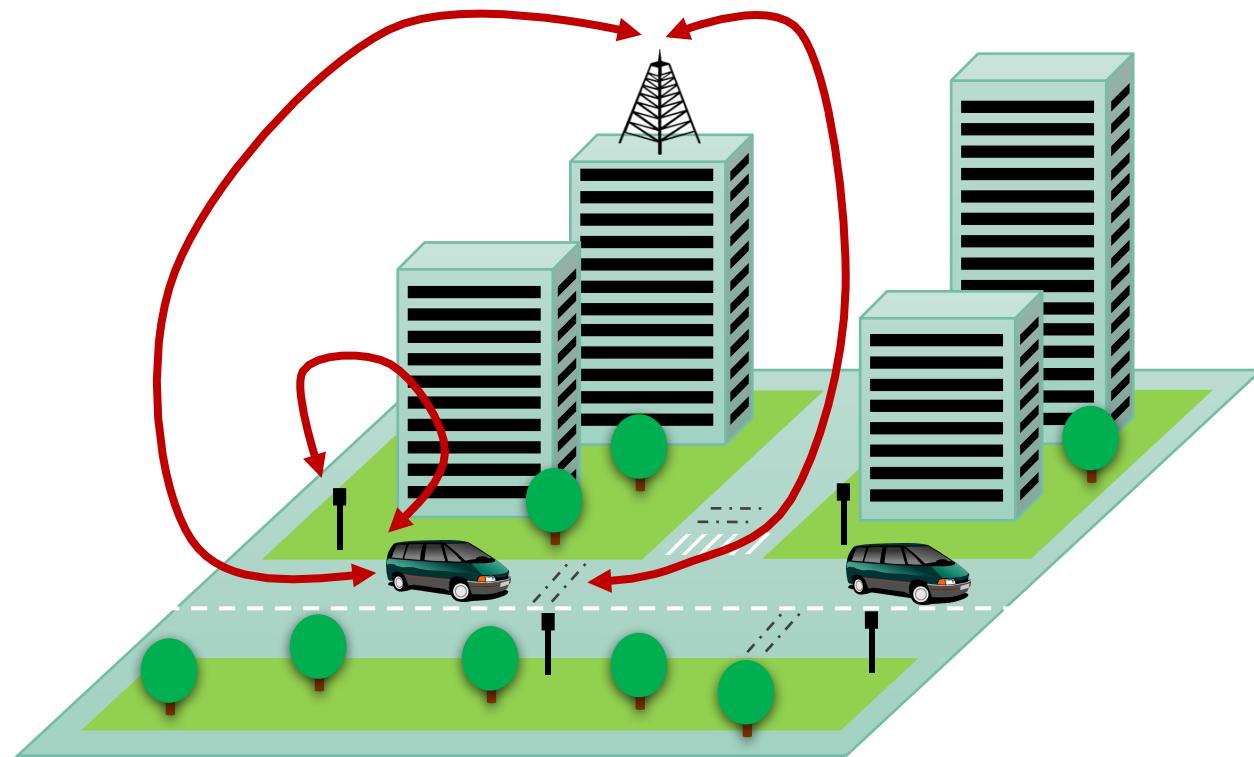
# SymbioCity



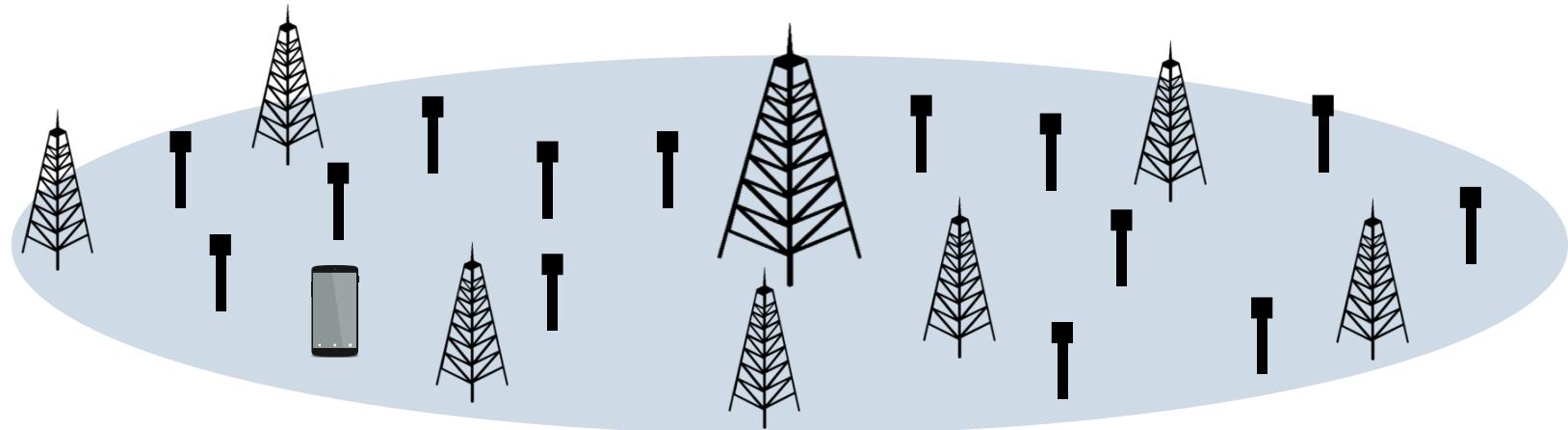
Smart City data can be used by the network to increase its awareness



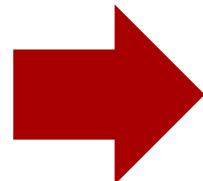
Smarter Self-Organizing Networks



# Heterogeneous Networks



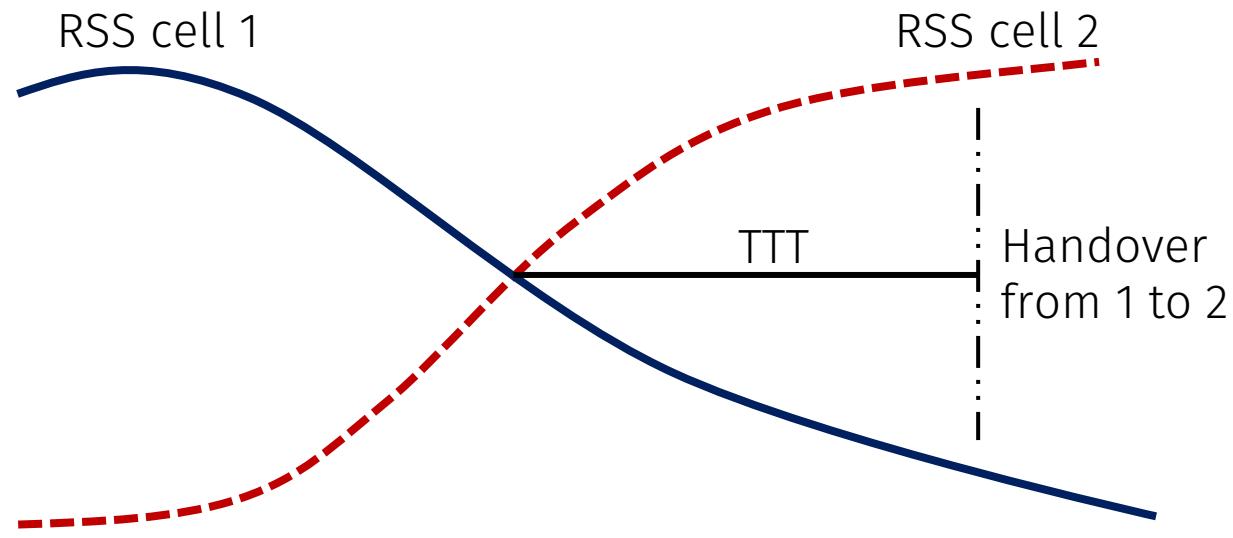
- Cells of different sizes and capabilities
- Tons of parameters to tune
- Perfect fit for SONs



Mobility procedures  
(handovers) are an issue

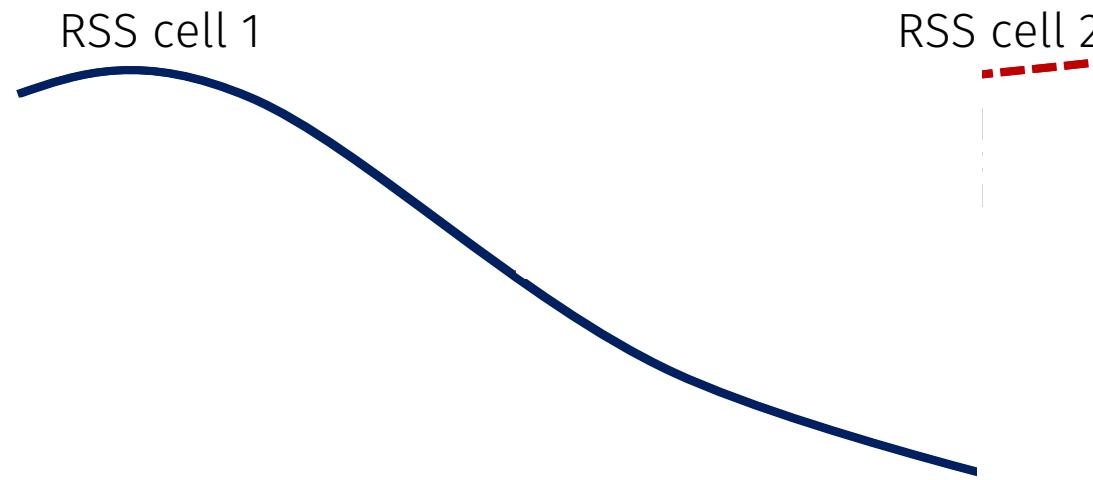
# Handover in HetNets

- Frequent HO
  - Dense cells
  - Small cells
- RSS-based



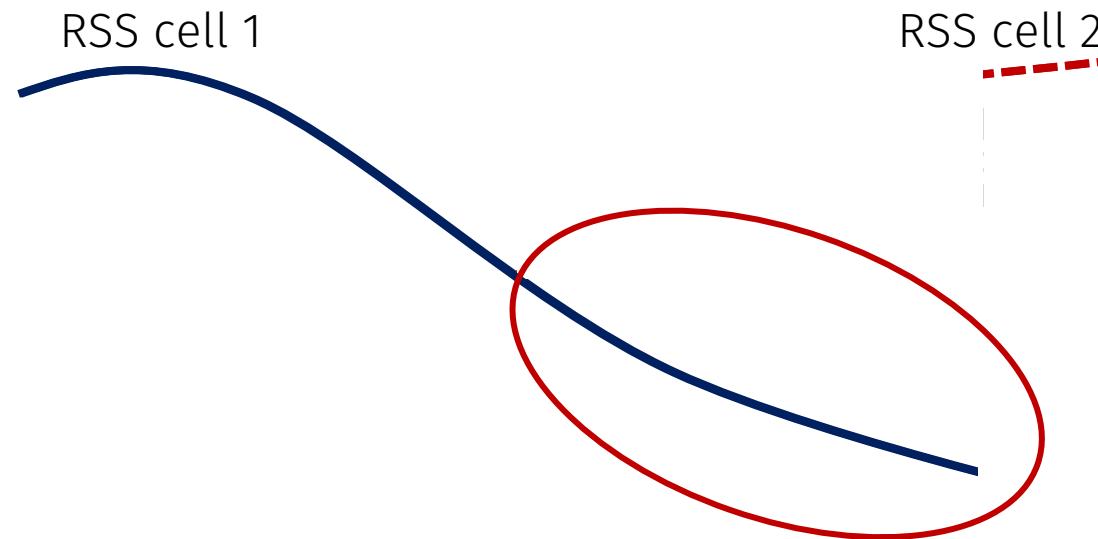
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# Handover in HetNets

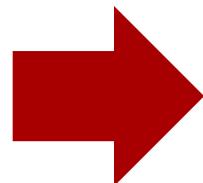
- Frequent HO
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# Handover in HetNets

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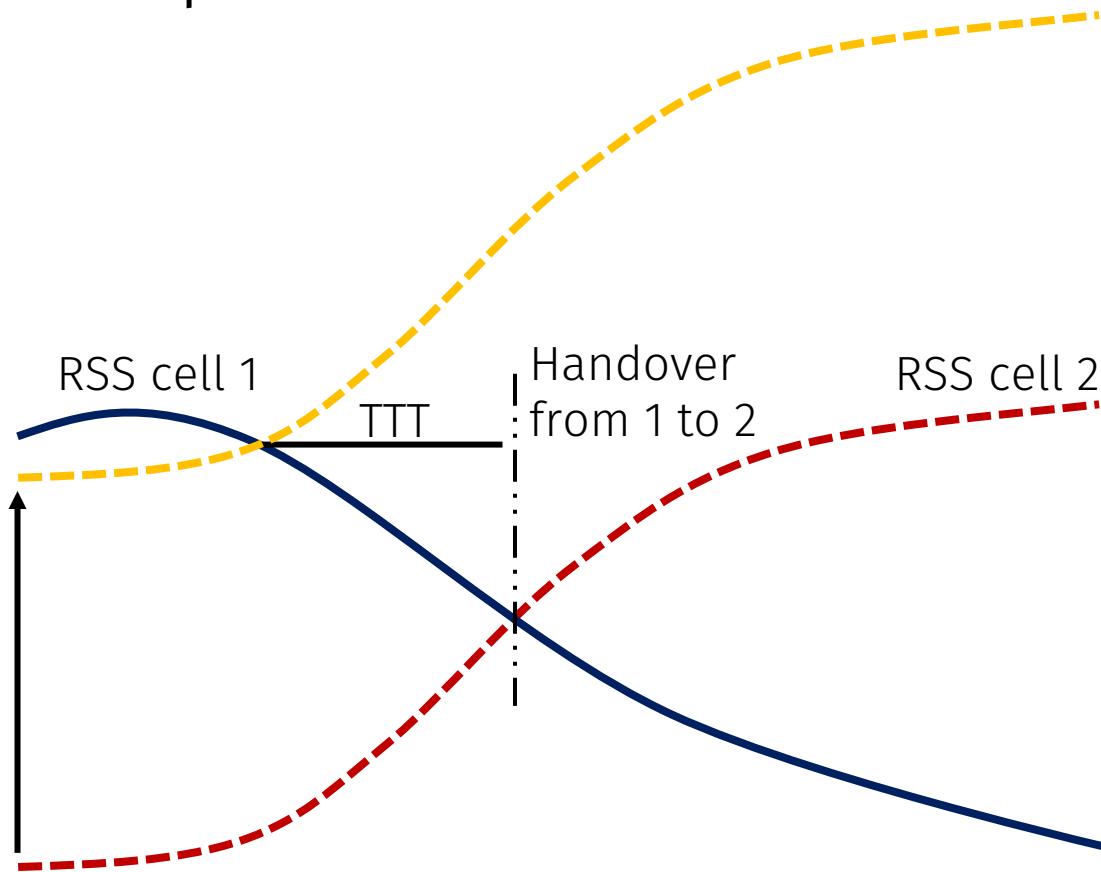
- Frequent HO
  - Dense cells
  - Small cells
- RSS-based



SymbioCity application: avoid ping-pong & increase capacity

# Range Expansion Bias

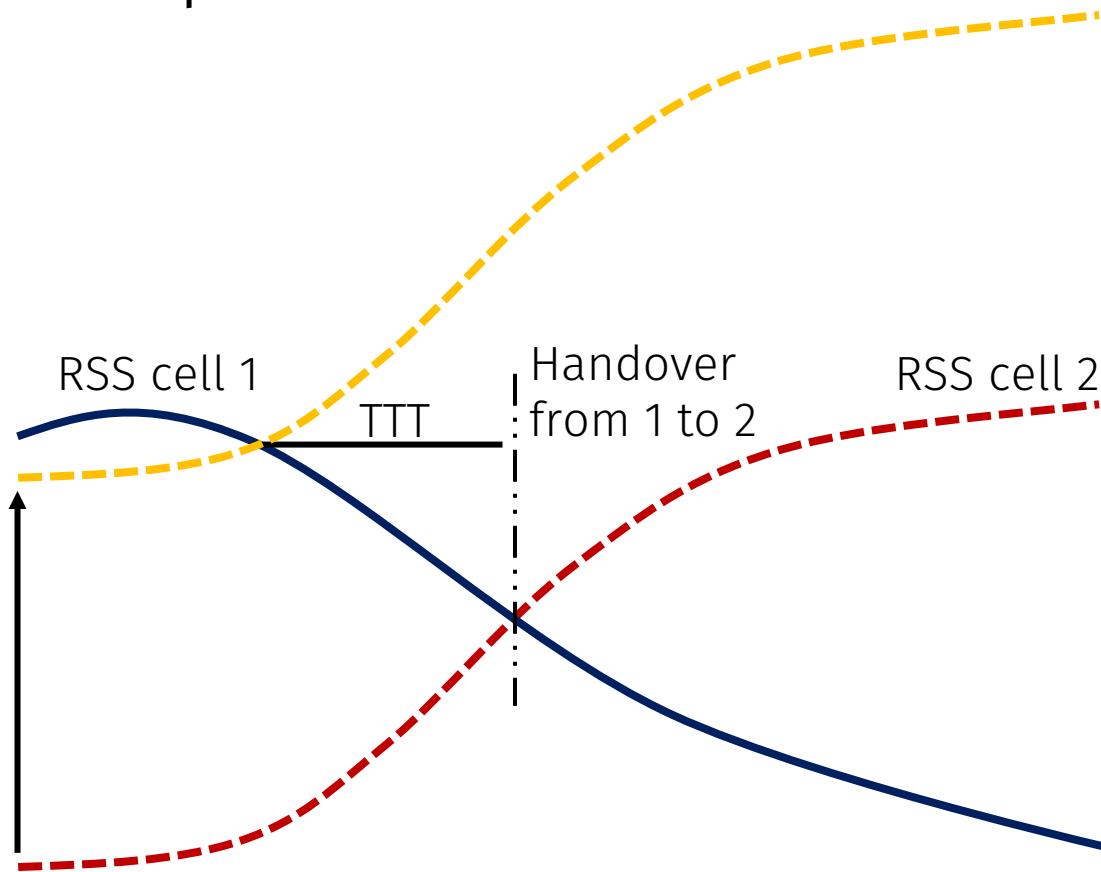
- Additive parameter



How to tune it?

# Range Expansion Bias

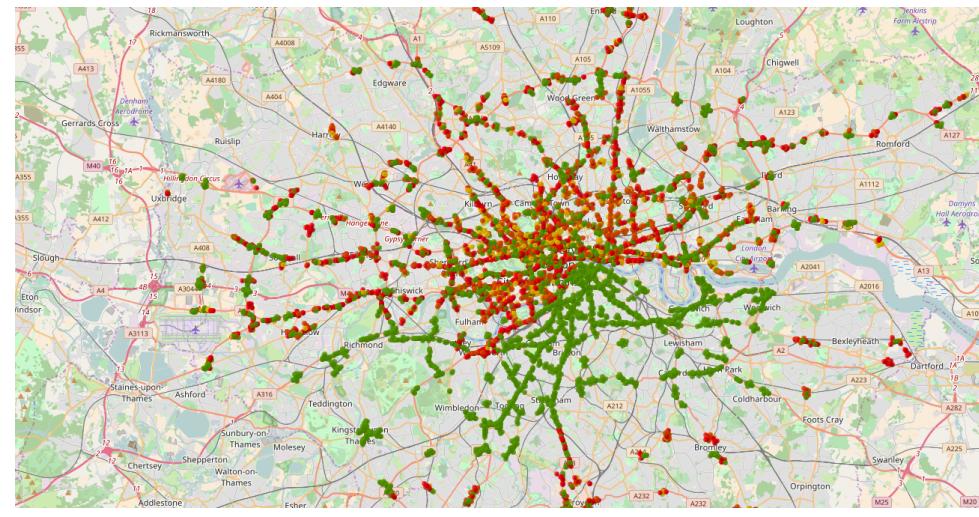
- Additive parameter



... we need data!

# London UTC network

Traffic light timing optimization to reduce congestion



From: Traffic for London (TfL) – first quarter 2015

# Speed estimation

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Detector

00000011111000000000000000000000000000111100000000



Headway = count of 0's



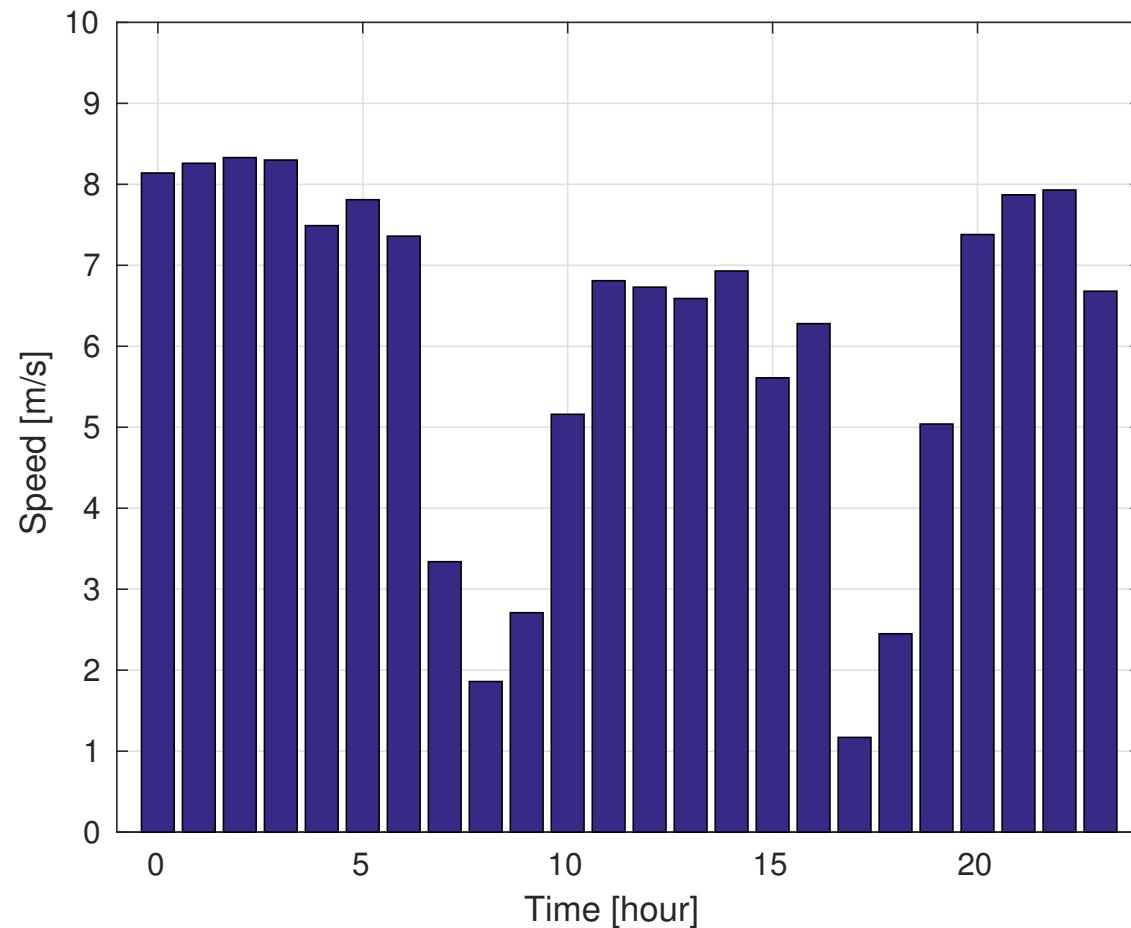
Vehicle Length = count of 1's

Next Vehicle Length = count of 1's

- $T_s = 250$  ms
- Vehicle length  $L$  (we assume  $L = 4$  m)
- Number of 1's  $n$

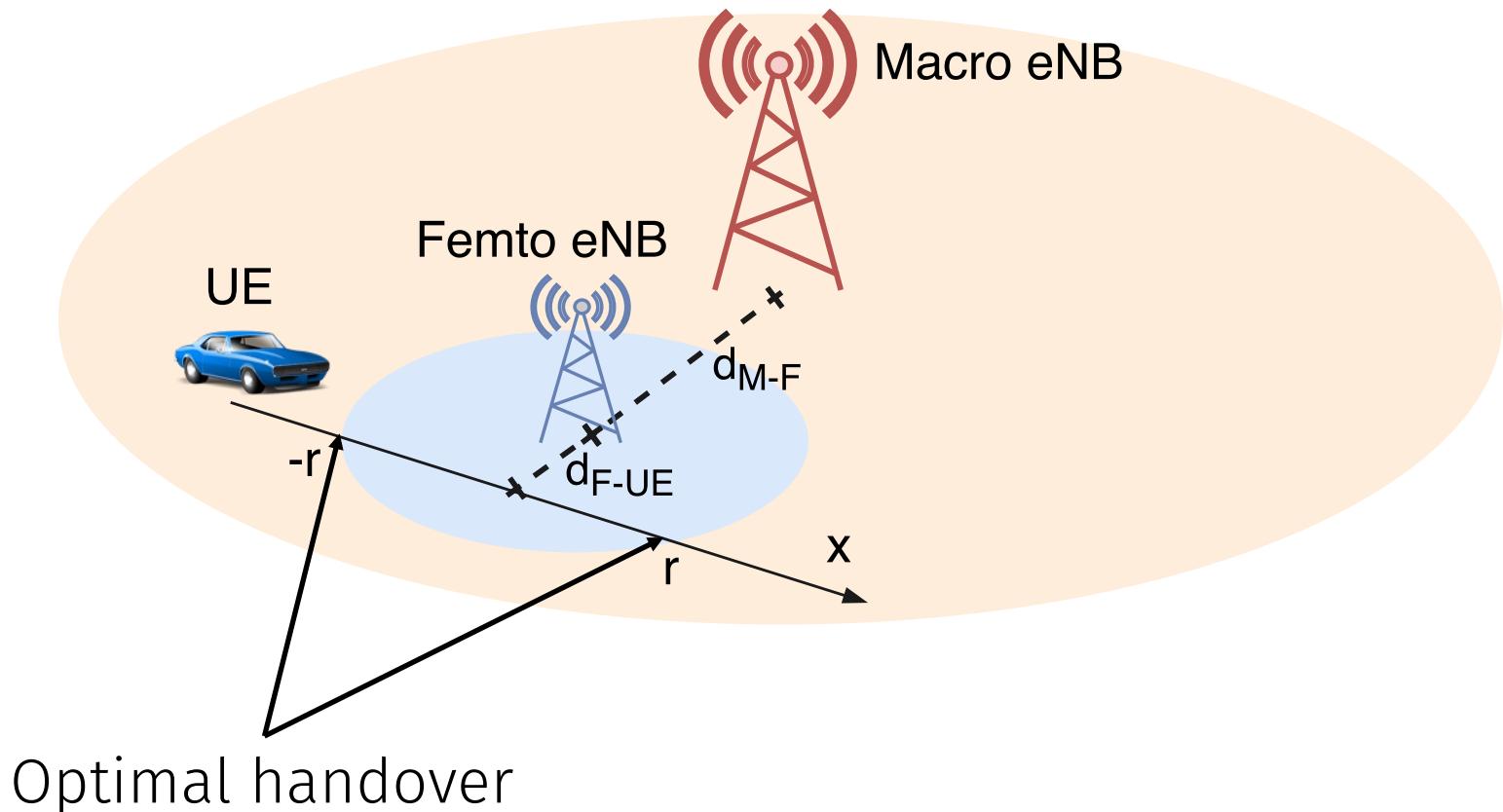
$$v = \frac{L}{nT_s}$$

# Example



Hourly average, January 23, 2015. Intersection between Homerton High St. and Daubeny Rd.

# Scenario



- Speed-based
- Asymmetrical

# Channel model

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$$P_{RX}^H(t) = P_{TX}^H(t)\Psi_{SH}\alpha(t)h(f_0, \beta, d)$$

# Channel model

$$P_{RX}^H(t) = P_{TX}^H(t) \Psi_{SH} \alpha(t) h(f_0, \beta, d)$$

$H \in \{F, M\}$       Shadowing      Fading      Pathloss

The diagram illustrates the components of a channel model equation. The equation is:

$$P_{RX}^H(t) = P_{TX}^H(t) \Psi_{SH} \alpha(t) h(f_0, \beta, d)$$

The components are labeled as follows:

- $H \in \{F, M\}$  (above  $\Psi_{SH}$ )
- Shadowing (pointing to  $\Psi_{SH}$ )
- Fading (pointing to  $\alpha(t)$ )
- Pathloss (pointing to  $h(f_0, \beta, d)$ )

# Channel model

$$P_{RX}^H(t) = P_{TX}^H(t) \Psi_{SH} \alpha(t) h(f_0, \beta, d)$$

$$h(f_0, \beta, d) = A \left( \frac{c}{4\pi f_0} \right)^2 \left( \frac{d}{d_0} \right)^{-\beta}$$

# Channel model

$$P_{RX}^H(t) = P_{TX}^H(t) \Psi_{SH} \alpha(t) h(f_0, \beta, d)$$

$$h(f_0, \beta, d) = A \left( \frac{c}{4\pi f_0} \right)^2 \left( \frac{d}{d_0} \right)^{-\beta}$$

Diagram illustrating the components of the channel coefficient:

- $\frac{c}{4\pi f_0}$  is labeled "Carrier Frequency".
- $\left( \frac{d}{d_0} \right)^{-\beta}$  is labeled "Distance".
- $-\beta$  is labeled "Pathloss exponent".

# Channel model

---

$$P_{RX}^H(t) = P_{TX}^H(t)\Psi_{SH}\alpha(t)h(f_0, \beta, d)$$

$$h(f_0, \beta, d) = A \left( \frac{c}{4\pi f_0} \right)^2 \left( \frac{d}{d_0} \right)^{-\beta}$$

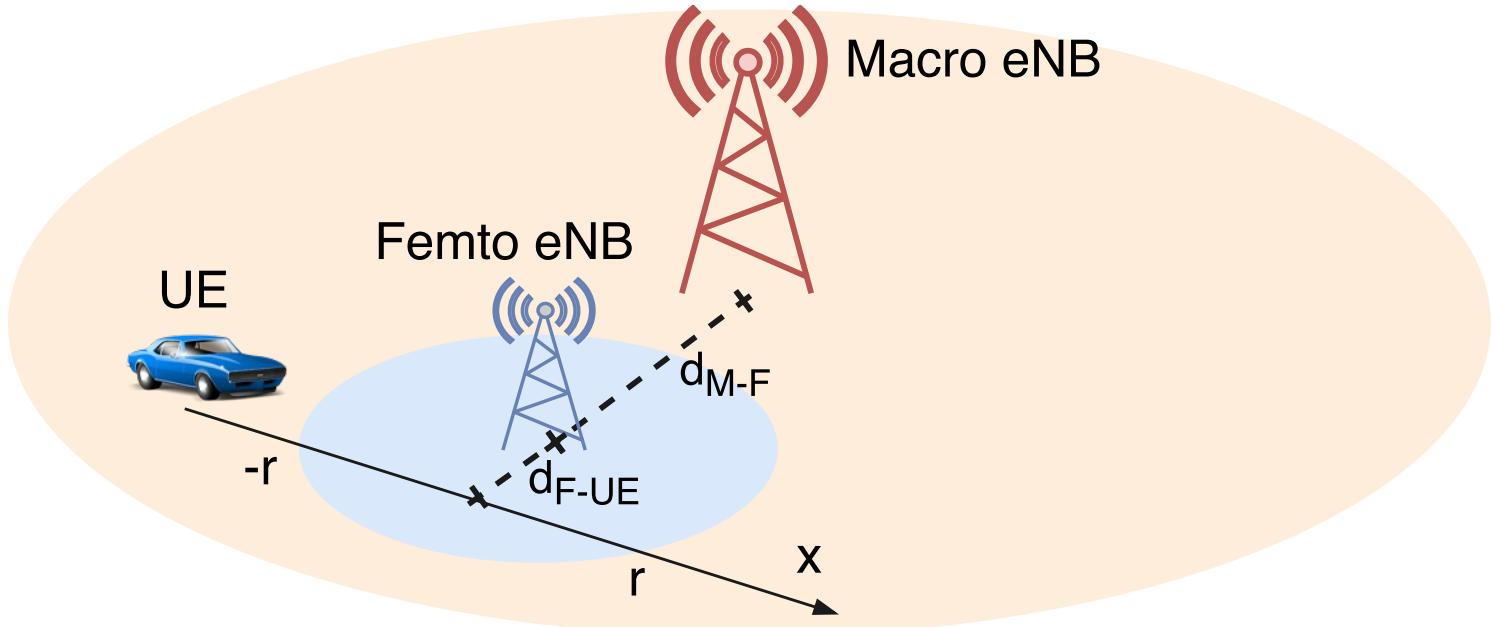
$$\gamma_H(t) = \frac{P_{RX}^H}{N_0 B}$$

# Parameters

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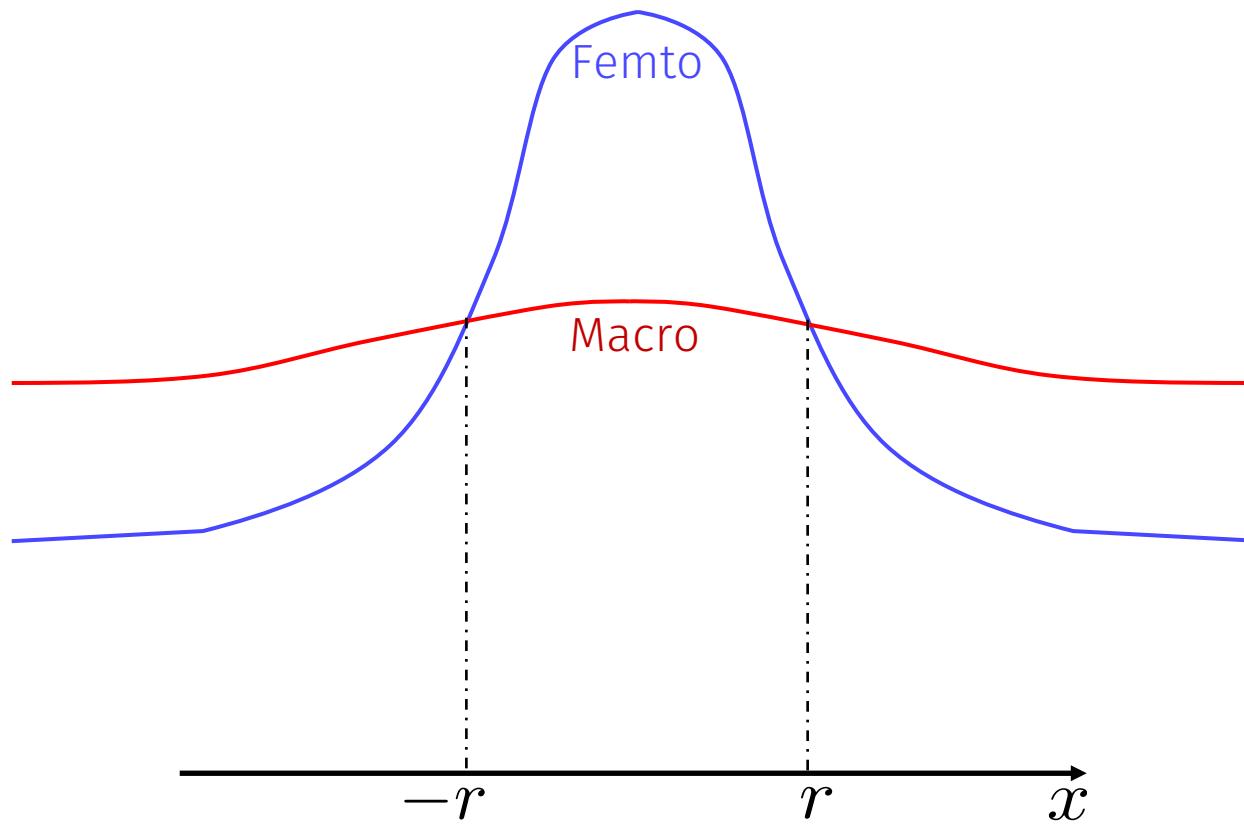
Parameter	Value	Description
$P_{TX}^M$	46	MeNB transmission power [dBm]
$P_{TX}^F$	26	FeNB transmission power [dBm]
$f_0^M$	900	MeNB carrier frequency [MHz]
$f_0^F$	1800	FeNB carrier frequency [MHz]
$B$	20	Bandwidth [MHz]
$d_{M-F}$	40	Distance between MeNB and FeNB [m]
$d_{F-UE}$	10	Distance between FeNB and UE [m]
$\sigma_M^2$	8	MeNB log-normal shadowing variance
$\sigma_F^2$	4	FeNB log-normal shadowing variance
$\beta_M$	4.28	MeNB pathloss exponent (NLOS)
$\beta_F$	3.76	FeNB pathloss exponent (LOS)

# Asymmetrical Bias



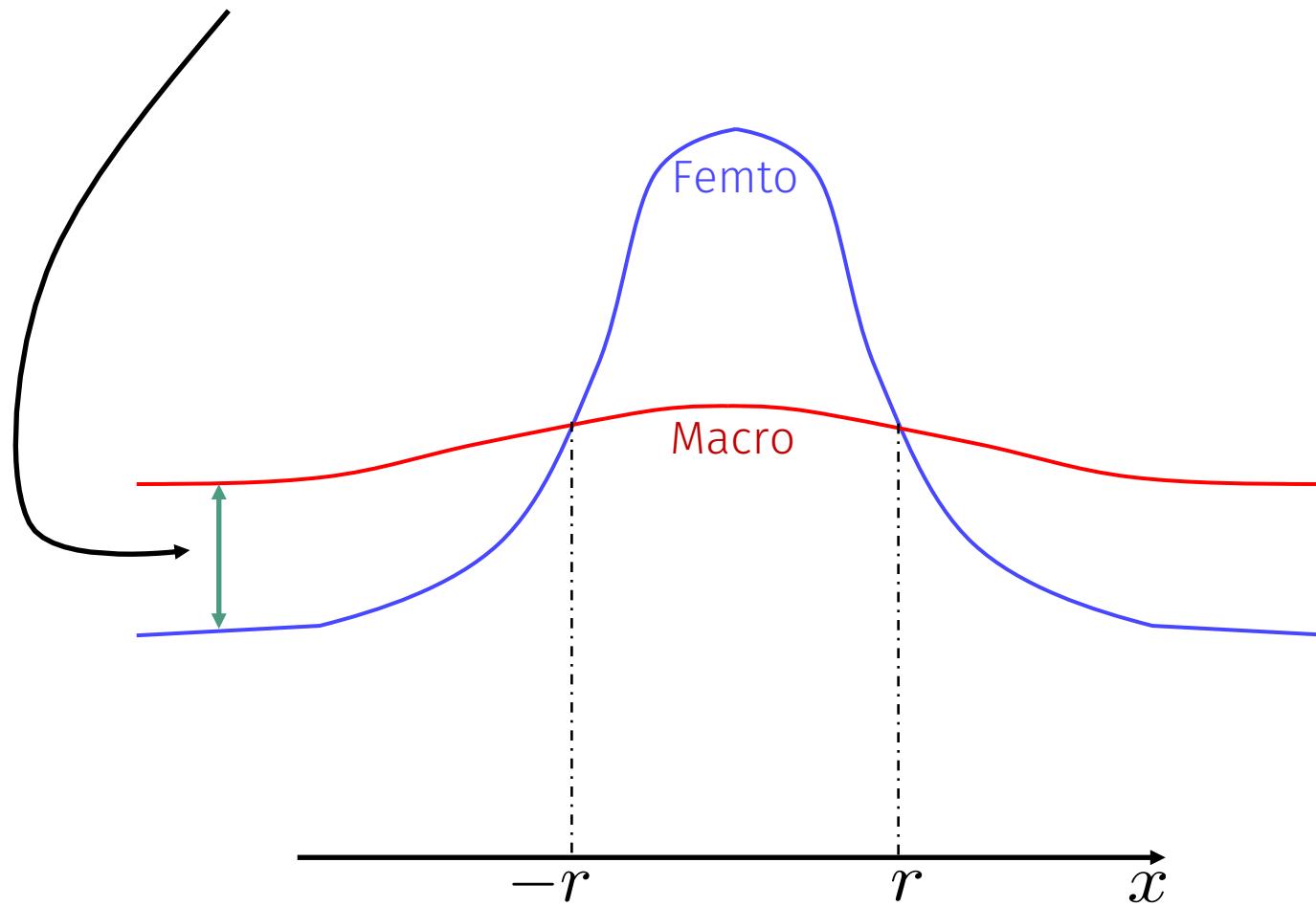
# Asymmetrical Bias

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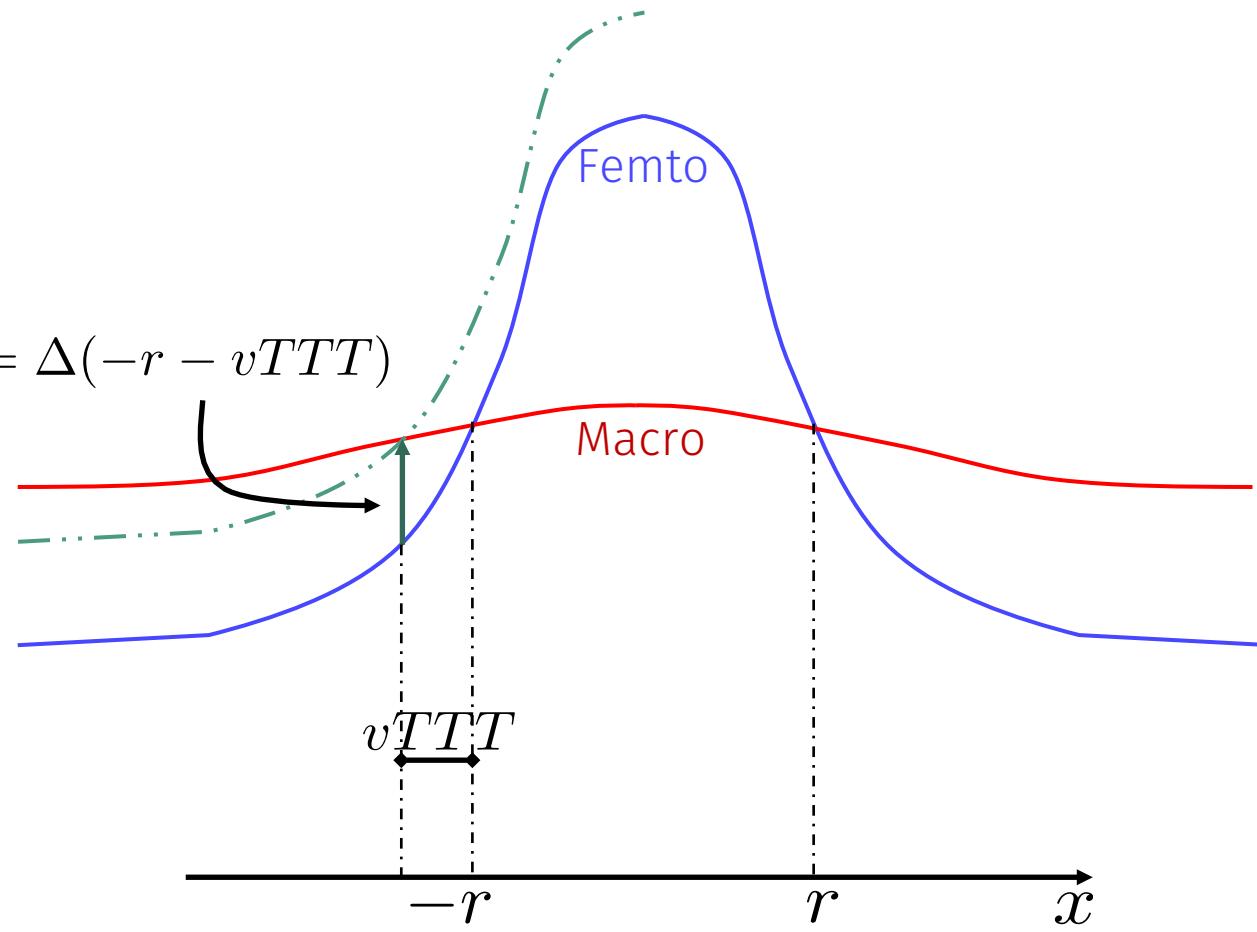
# Asymmetrical Bias

$$\Delta(x) = \bar{\gamma}_F(x) - \bar{\gamma}_M(x)$$

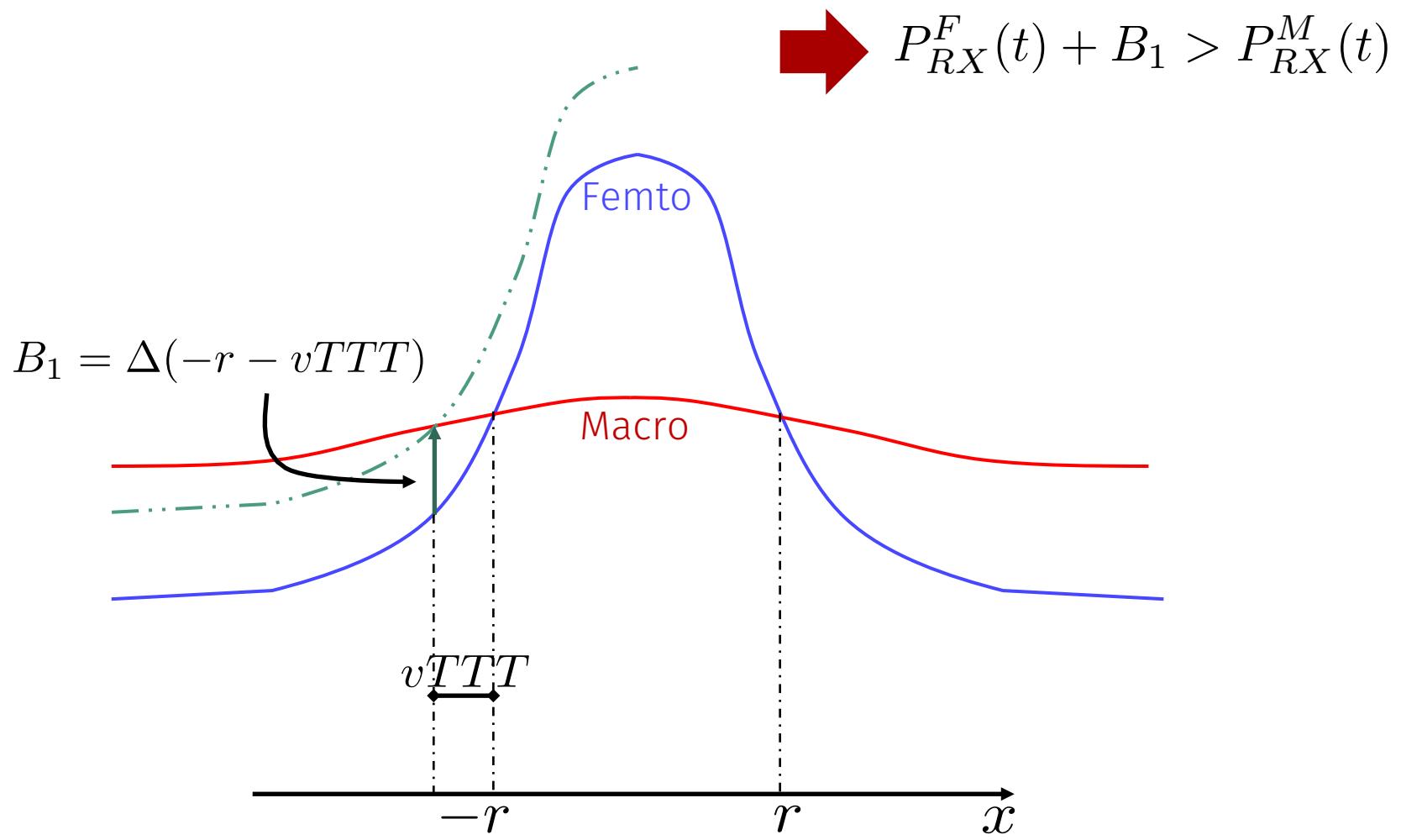


# Asymmetrical Bias

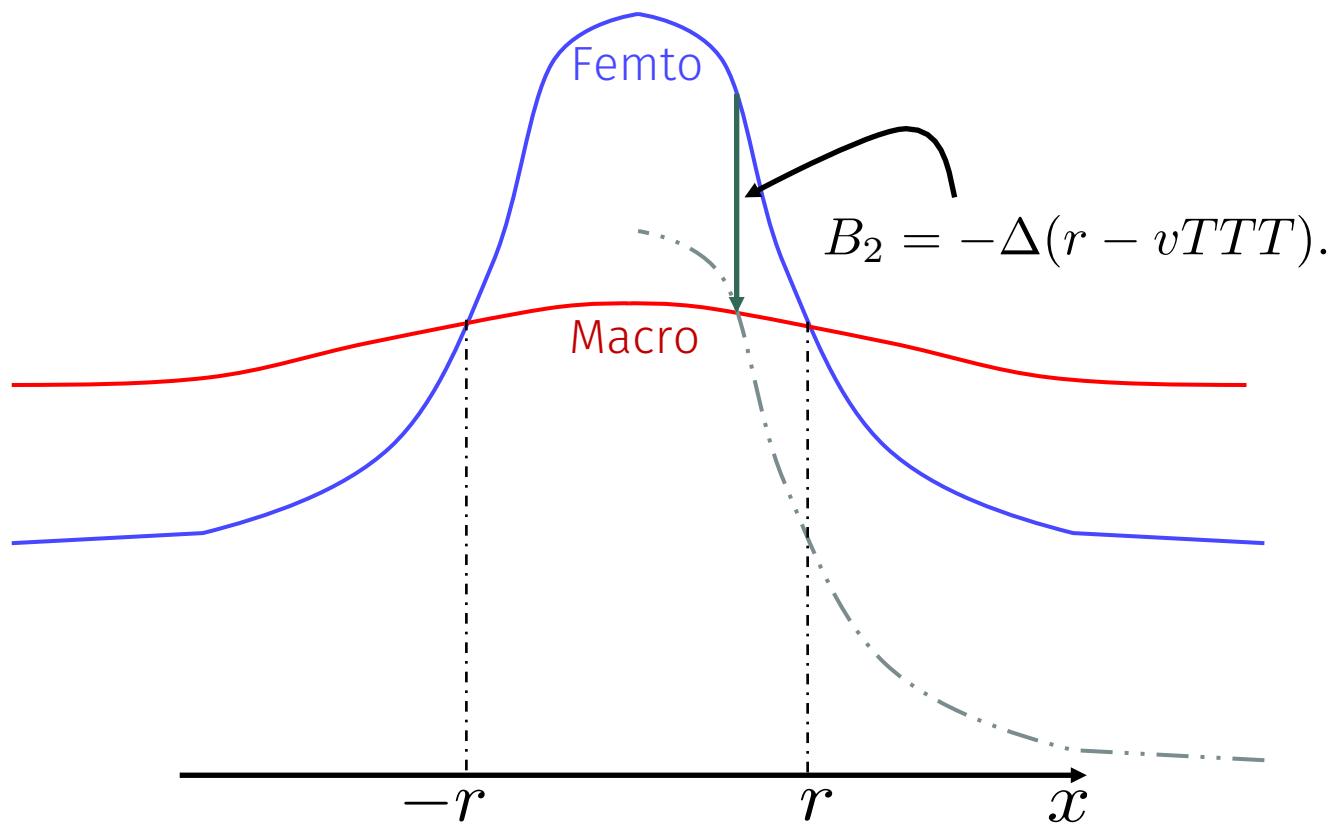
$$B_1 = \Delta(-r - vTTT)$$



# Asymmetrical Bias

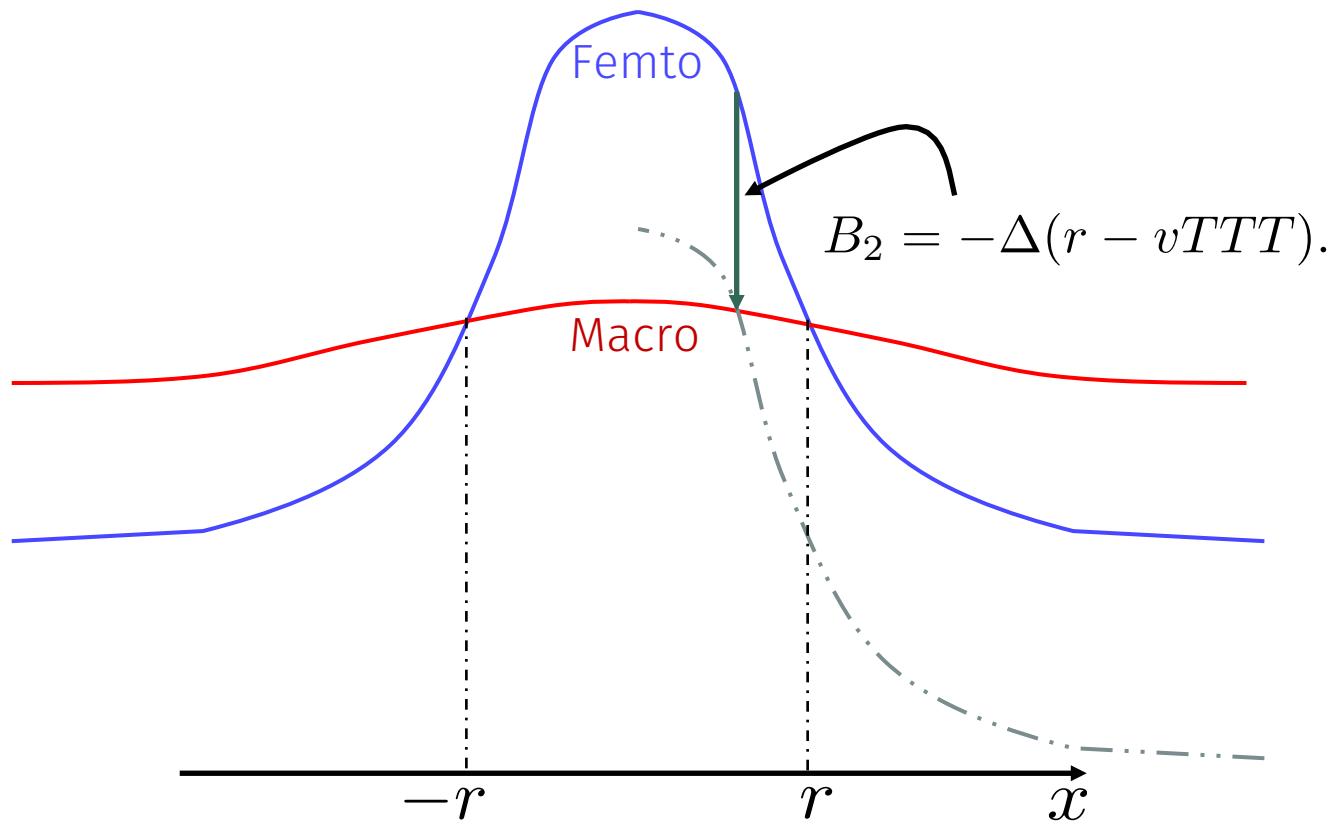


# Asymmetrical Bias



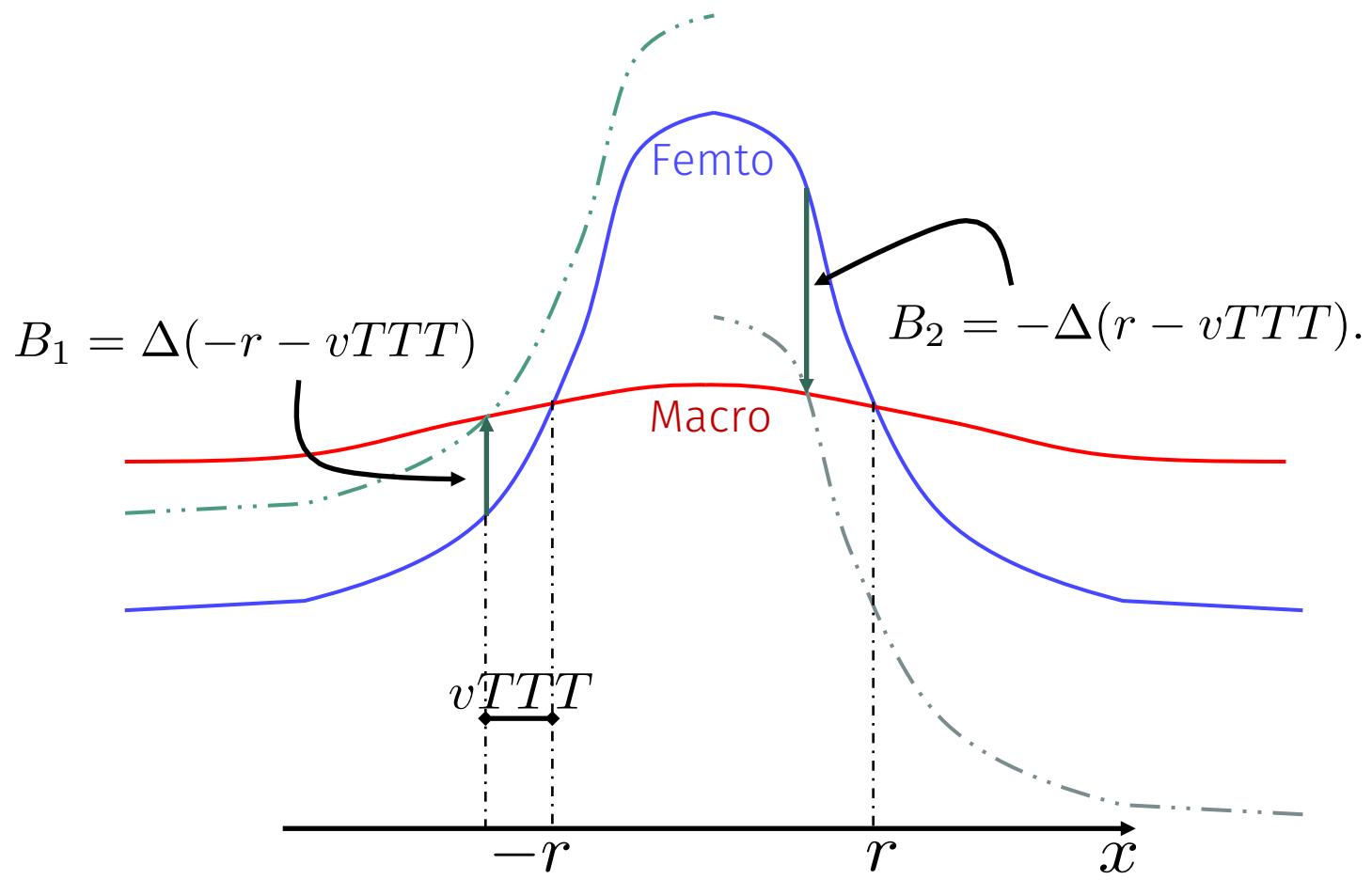
# Asymmetrical Bias

$$\rightarrow P_{RX}^M(t) + B_2 > P_{RX}^F(t)$$



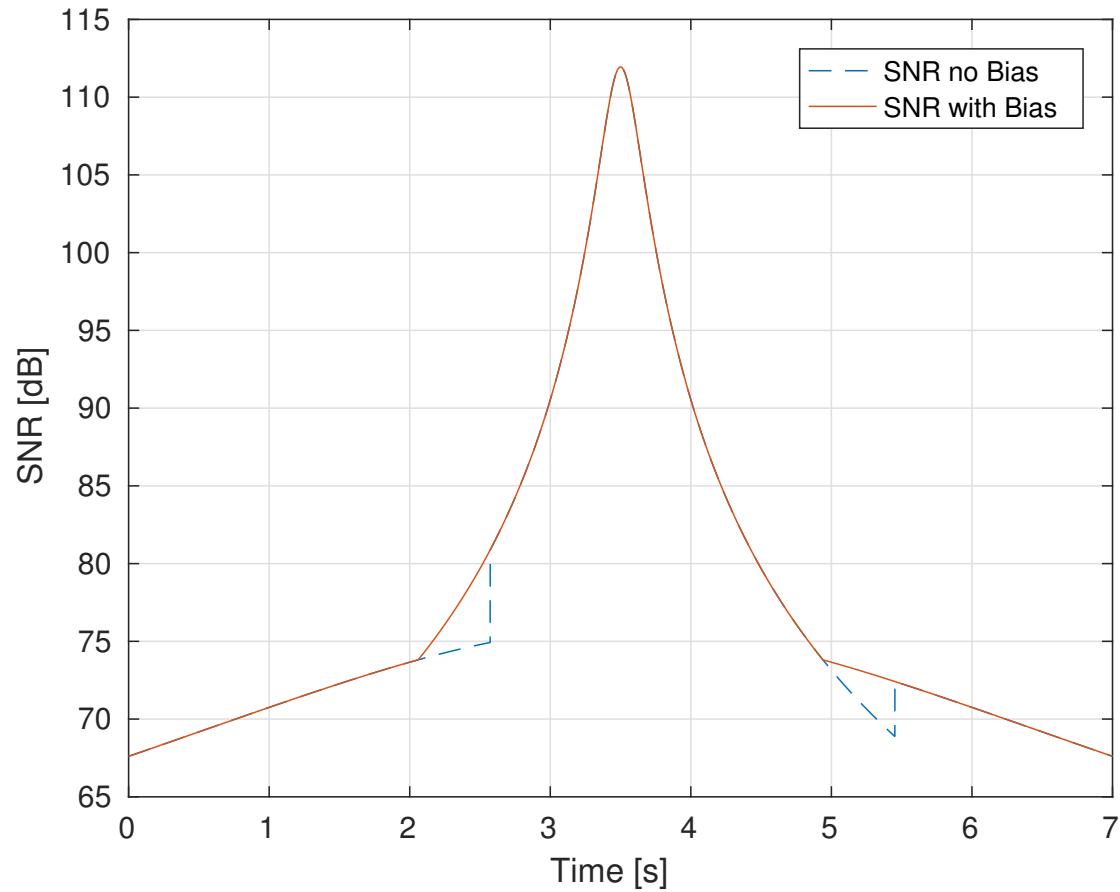
# Asymmetrical Bias

$$\Delta(x) = \bar{\gamma}_F(x) - \bar{\gamma}_M(x)$$

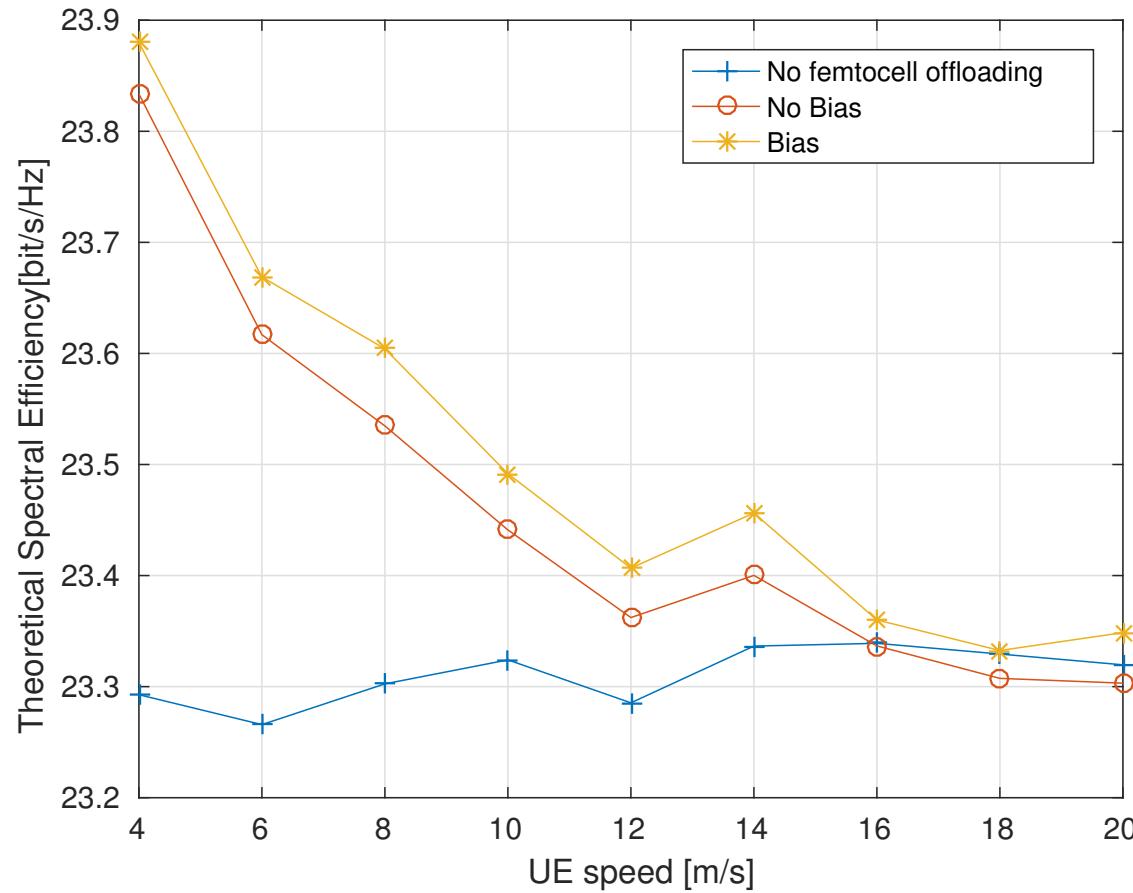


# SNR with and without bias

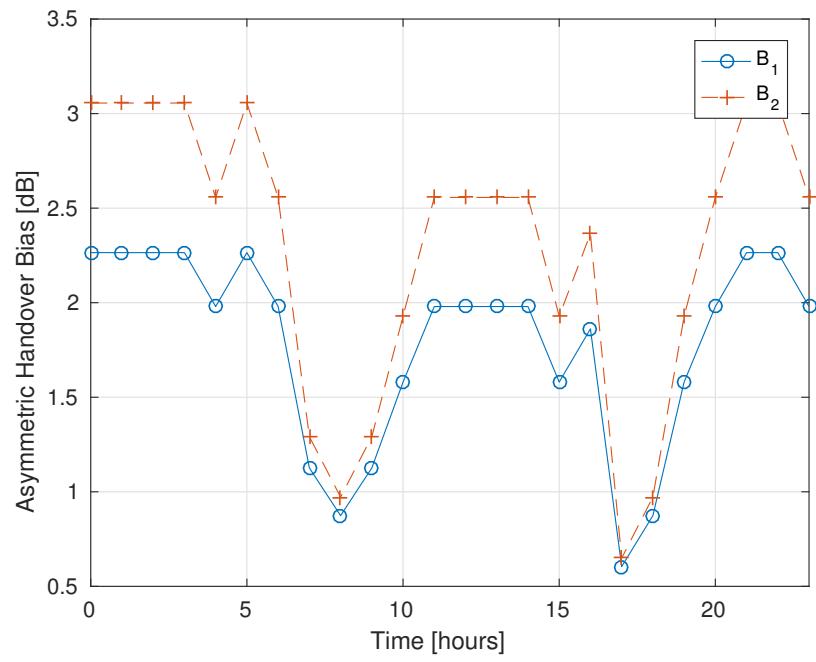
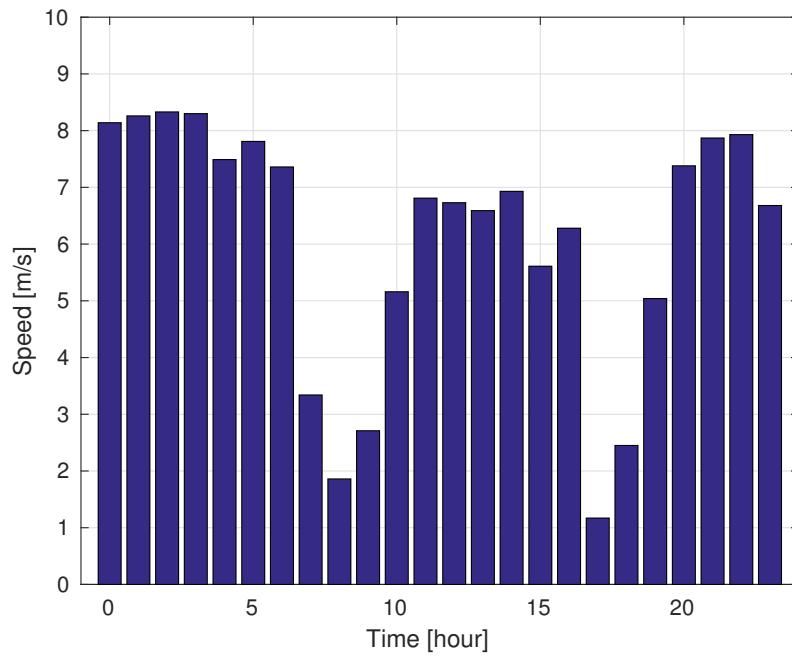
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# Spectral Efficiency



# Bias values



# Conclusions

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- SymbioCity idea: smarter networks with Smart City data
- Application: handover in HetNets
  - Data from TfL
  - Range expansion bias
  - Increased efficiency without *ping pong*
- Future works
  - Dynamic vMME allocation
  - Integration with more data

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