



Modeling and Validation of Automotive RADAR MMIC Impairments by using the Standardized Interfaces for Closed-Loop Simulation

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MMIC: Monolithic Microwave Integrated Circuit

Overview



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Introduction

02

Physical RADAR Sensor Model

03

Modeling of MMIC Impairments

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Conclusion and Outlook

Automotive Sensors

Advanced driver assistance systems (ADAS) sensors and example applications

Adaptive Cruise Control



<https://www.bosch-mobility-solutions.com/>

Camera



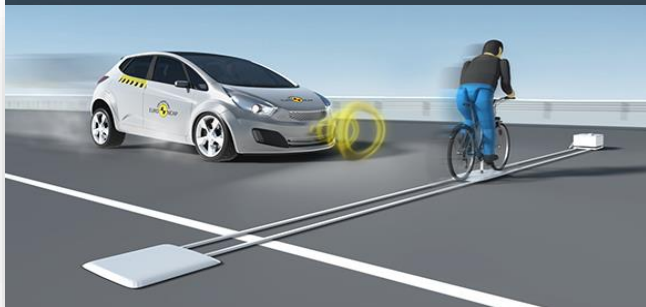
<https://www.kostal-automobil-elektrik.com/>

LiDAR



<https://www.blickfeld.com/>

Automated Emergency Braking



<https://www.openpr.com/>

Ultrasonic Sensors



<https://www.bosch-mobility-solutions.com/>

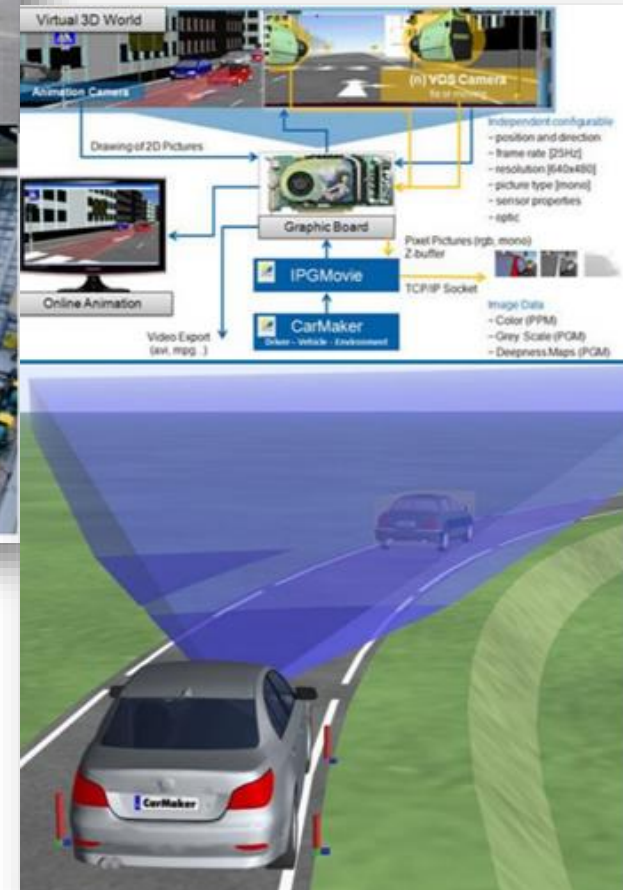
RADAR



<https://www.everythingrf.com/News/details>

Problem Statement

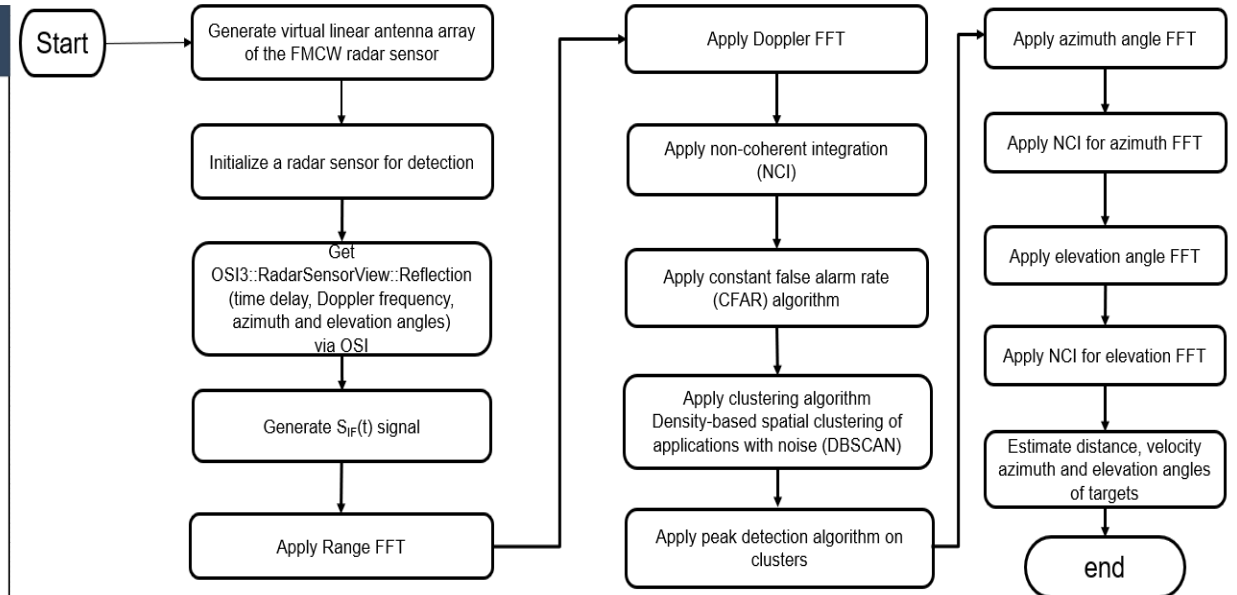
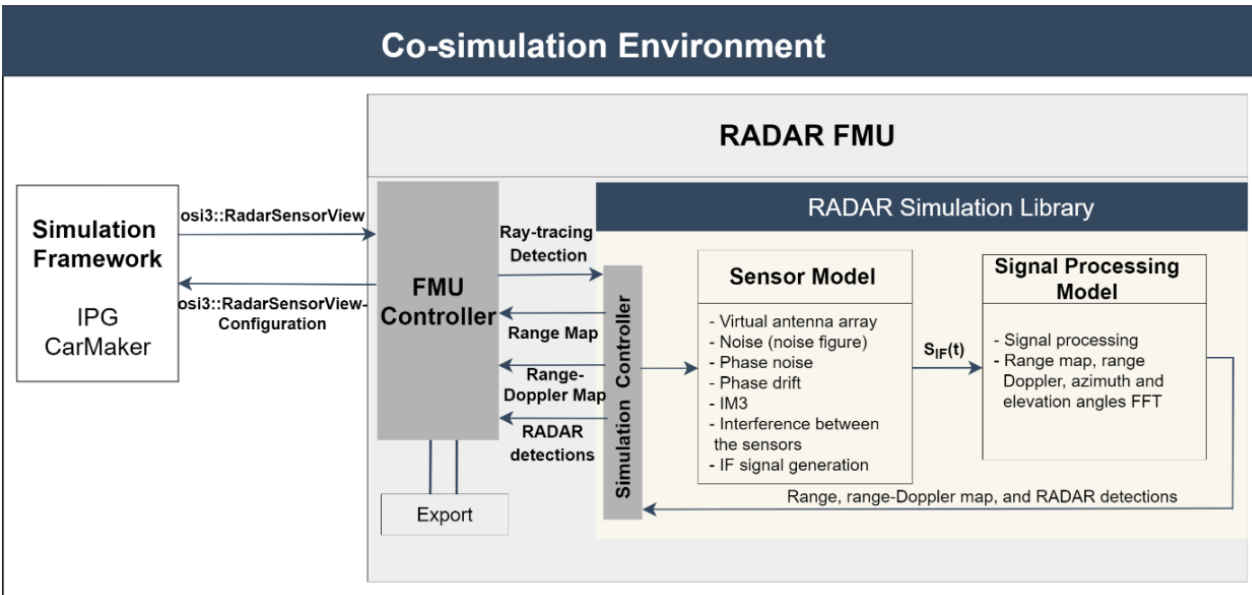
- Validation of these systems is done with real test drives which are expensive, time consuming, safety critical
- ADAS Safety functions require a proof distance of about 240 million km*
- Methods for ADAS Validation
 - Prototypes and road trials
 - Model-in-the-Loop Testing (driving simulator)
 - Hardware-in-the-Loop Testing (senor test benches)
 - **Combination of simulation & real-world test: hybrid strategy**
- **Required:** Development and validation of physical ADAS sensor models



Sources:

- MAGNA Steyr, IPG, Toyota, FTG
- *Handbook of Driver Assistance Systems, Editors: Winner, H., Hakuli, S., Lotz, F., Singer, C.

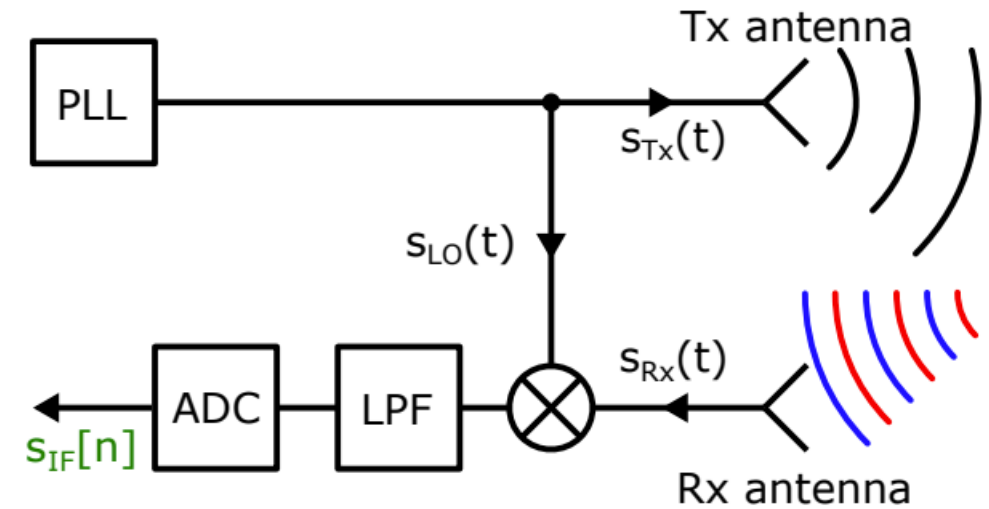
RADAR FMU Model Block Diagram



- “Radar_FMU” model is developed that retrieved the raytracing data via OSI
- The input of the “Radar_FMU” model is OSI3::SensorView->OSI::RadarSensorView::Reflection
- “Radar_FMU” also output distance, velocity, azimuth and elevation angle for each reflection/target
- Implemented RADAR MMIC impairments: Phase noise, **third order intermodulation**

IF: intermediate frequency

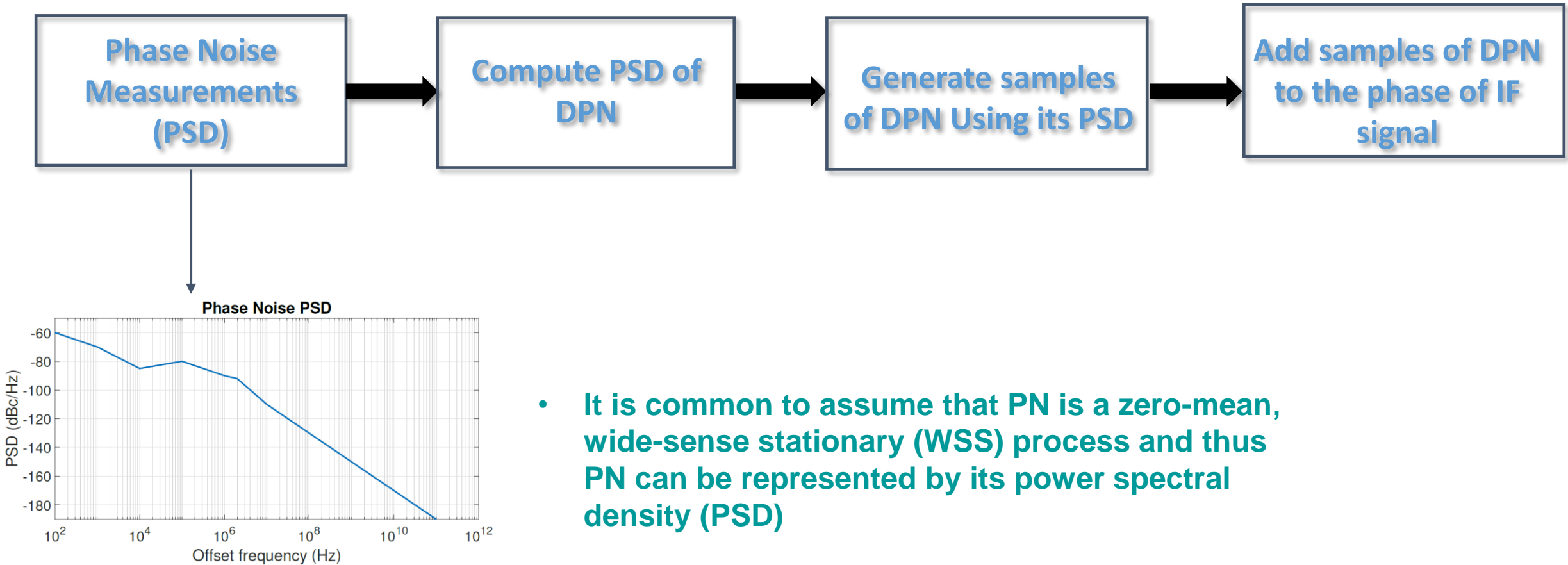
Phase Noise



- Phase Noise (PN) are random fluctuations in the phase of a signal due to non-idealities of the oscillators and phase-locked loop (PLL)
- **PN limits received SNR: Weak object signal is buried** under the PN of an adjacent strong object; it also has a strong impact on the velocity domain
- Spur in PN generate the ghost objects

Simulate Decorrelated Phase Noise (DPN) in IF level

- Phase noise after mixer is referred as decorrelated phase noise (DPN)



Mathematical Model for PN In IF Level

- Transmit signal can be written as $S_{Tx}(t) = A_{Tx} \cos(2\pi f_0 t + \pi k t^2 + \varphi(t))$

Where k is the slope of the chips calculated as $k = \frac{B}{T}$ where B is the bandwidth of the transmit signal and $\varphi(t)$ is the Phase Noise (PN)

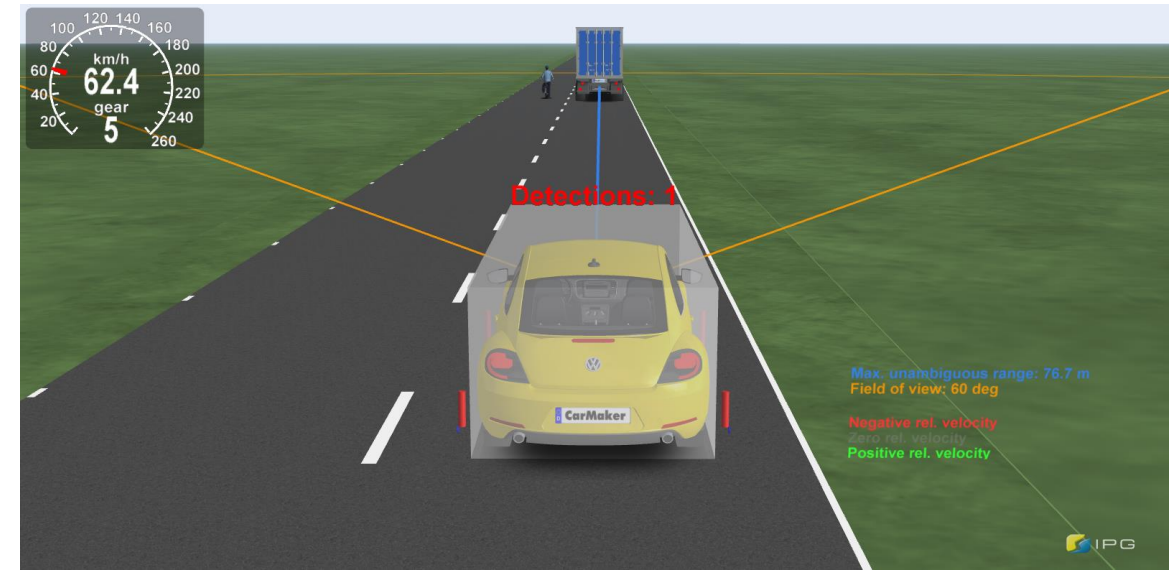
- The received signal $S_{Rx}(t) = A_{Tx} \cos(2\pi f_0(t - \tau) + \pi k(t - \tau)^2 + \varphi(t - \tau))$
Where τ is round trip delay time (RTDT) of target
- IF signal $S_{IF}(t) = [S_{Tx}(t) \cdot S_{Rx}(t)] * h_{LPF}(t)$

$$S_{IF}(t) = \left[\frac{A_{Tx}^2 \alpha}{2} \cos(2\pi f_B t + \Phi + \Delta\varphi(t)) \right] * h_{LPF}(t)$$

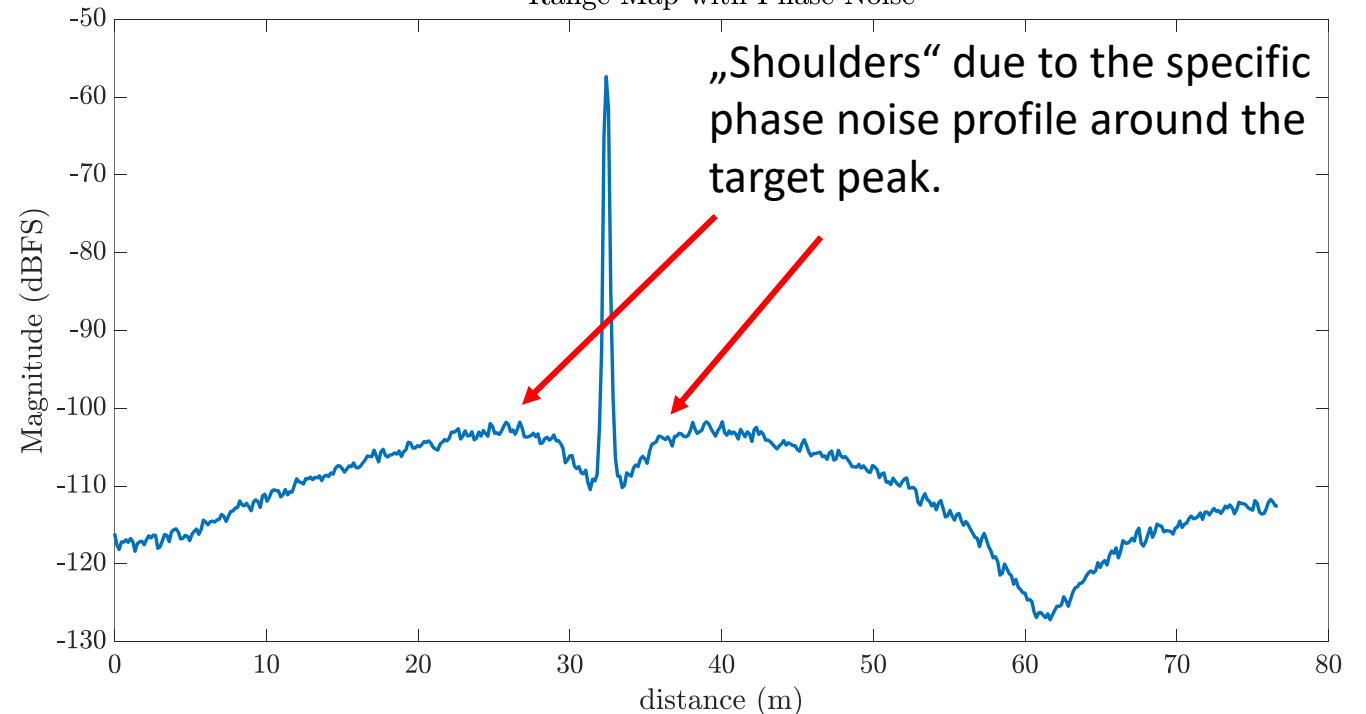
where $f_B = k\tau$ is the IF beat frequency and $\Phi = 2\pi f_0 \tau - \pi k \tau^2$ is a constant phase term and $\Delta\varphi(t) = \varphi(t) - \varphi(t - \tau)$ is decorrelated phase noise (DPN)

Range Map With Phase Noise Profile

- $f_0 = 77$ GHz
- $B = 1$ GHz
- Max. range = 74.94 m
- $T_{SW} = 40.96$ μ s
- $f_s = 25$ MHz
- Static Target at 30 m distance
- Azimuth angle: 0 deg
- 32 virtual receive antennas
- RX antenna spacing: $\frac{\lambda}{2}$
- RCS = 70 dBm²

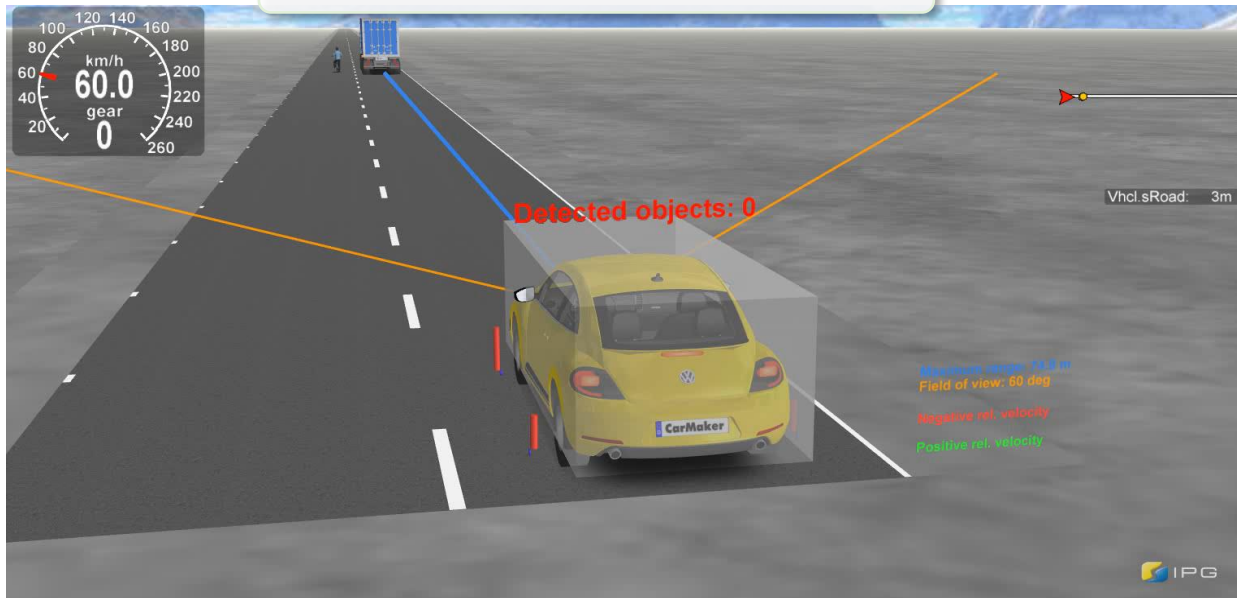


Range Map with Phase Noise

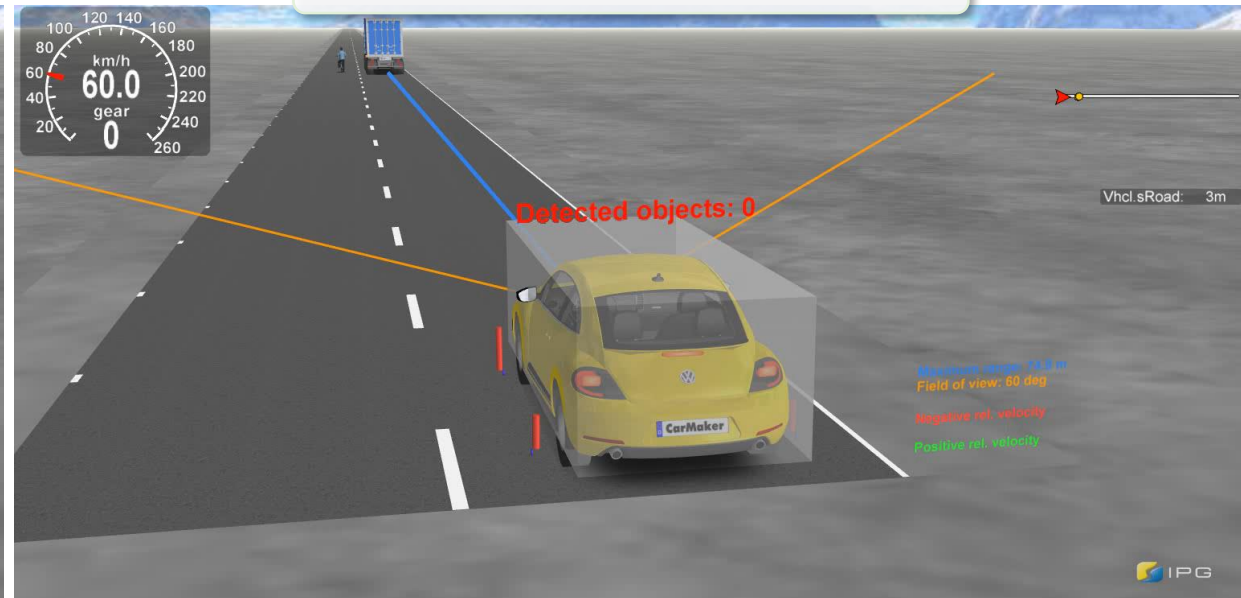


Phase Noise Visualisation in CarMaker

Simulation without phase noise



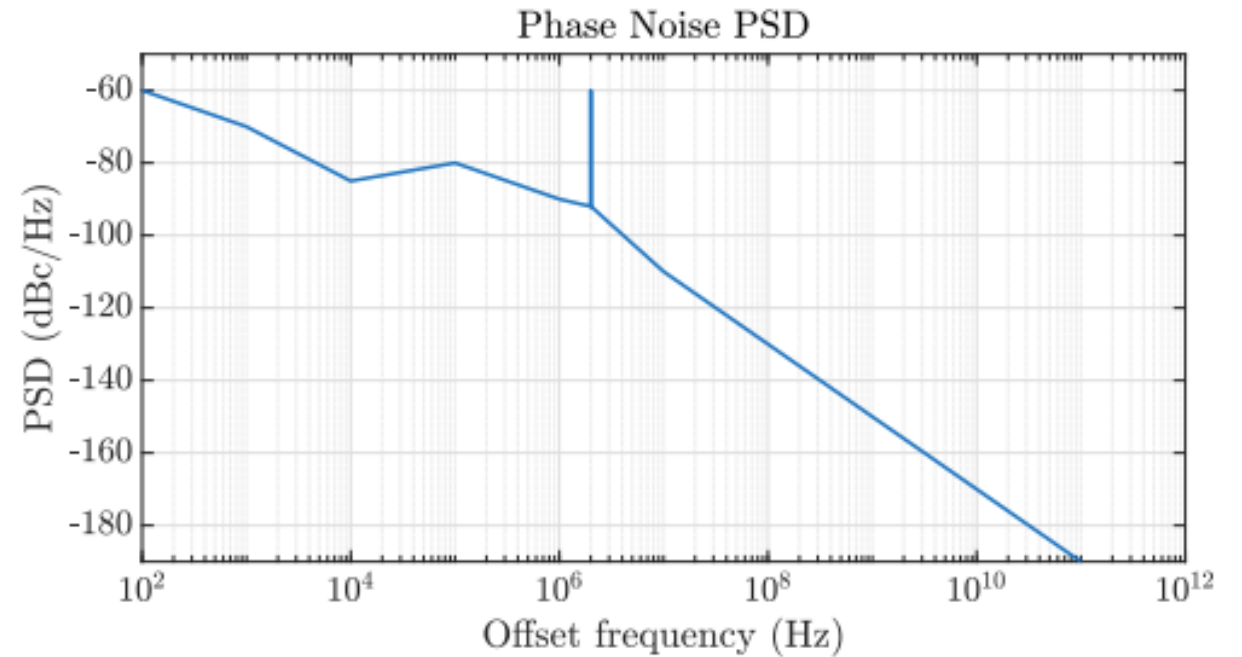
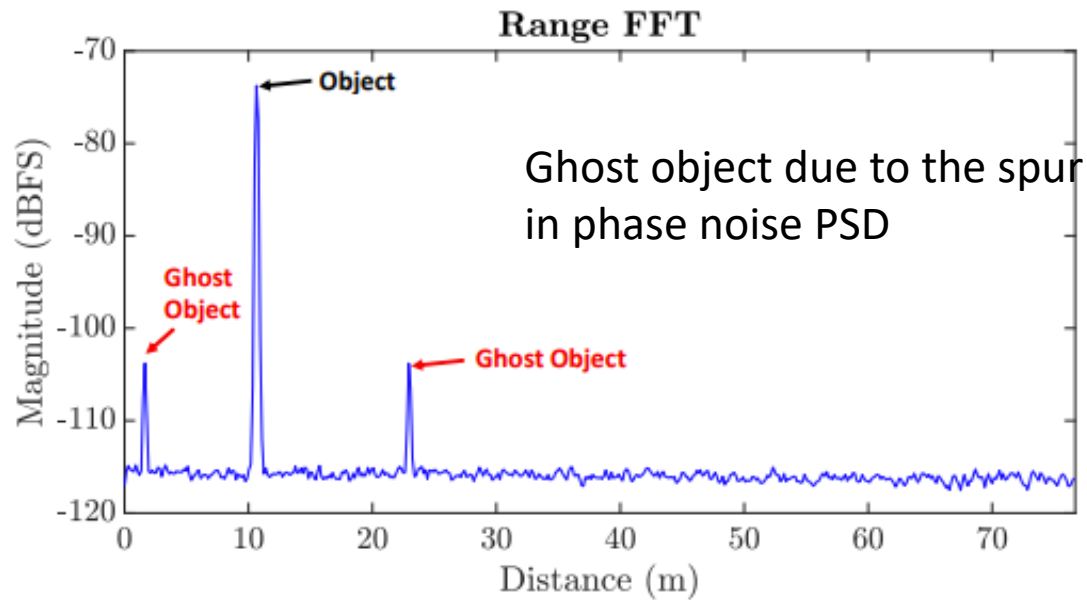
Simulation with phase noise



- Radar FMU simulation with thermal noise and without any other impairments
- Targets get detected as soon as they come into the field of view and the maximum detectable range
- RCS of truck: 70 dBm², RCS of bicycle: 7 dBm²

- Same scenario, but with phase noise
- RCS of truck: 70 dBm², RCS of bicycle: 7 dBm²
- Bicycle gets detected only at very close distance, at larger distances it gets masked by the truck.

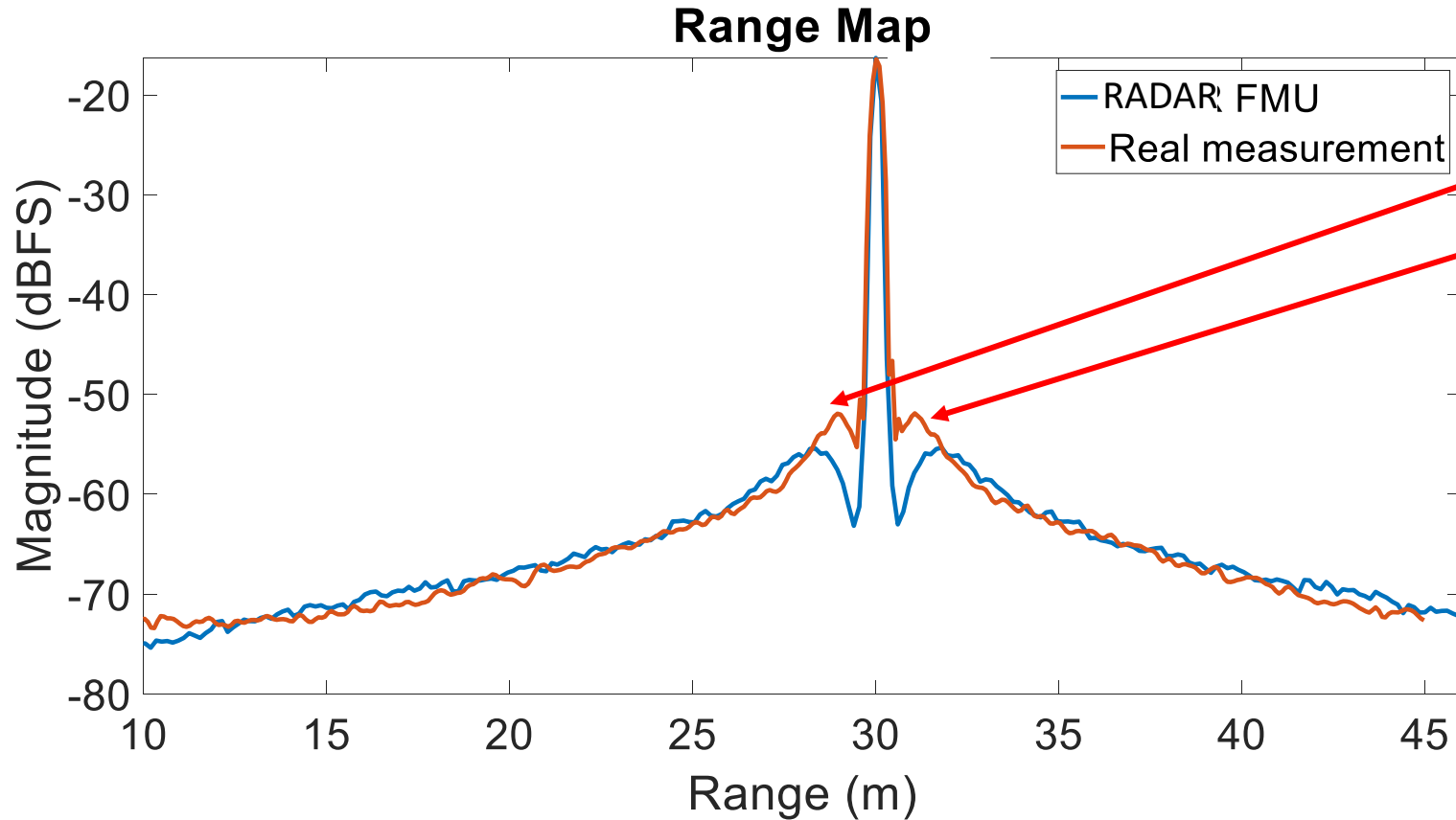
Ghost Objects Due to Phase Noise



- $f_{Ghost\ 1,2} = f_{spur} \pm f_B$

f_{spur} is the frequency of phase noise spur and f_B is the beat frequency of target

RADAR FMU Model: Validation of Phase Noise



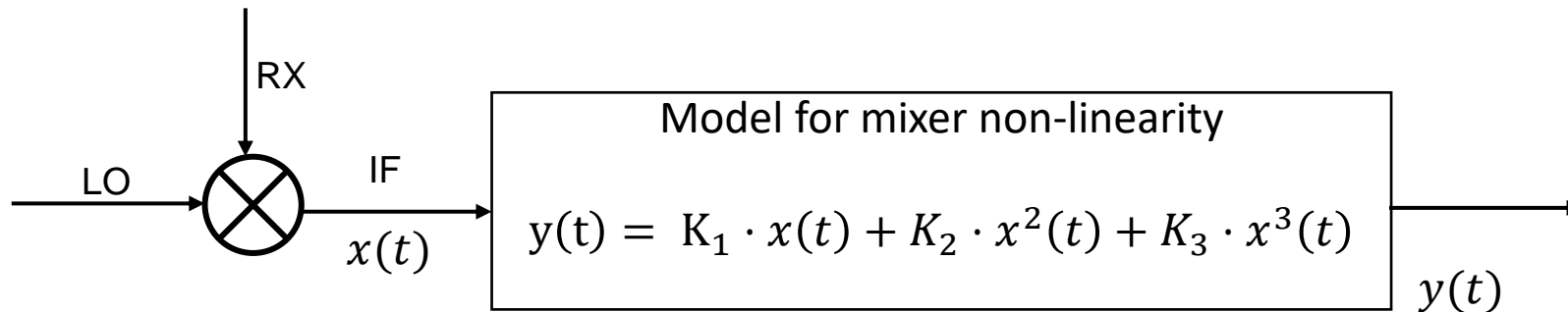
„Shoulders“ due to the specific phase noise profile around the target peak.

RCS = 70 dBm²

- Peak shape, noise level are matching well
- The difference in shoulder height appears because simulation and real measured phase noise profile are different

Mixer Non-Linearity: third-order intermodulation (IM3)

- Two-tone sinusoidal test signal with frequencies $\omega_{IF,1}$ and $\omega_{IF,2}$ is applied to the input of the non-ideal mixer will output (generally undesired) third-order intermodulation (IM3) components with frequencies $2\omega_{IF,1} \pm \omega_{IF,2}$ and $2\omega_{IF,2} \pm \omega_{IF,1}$



- K_1 : mixer voltage gain
- Only the IM3 products $2\omega_{IF,1} - \omega_{IF,2}$ and $2\omega_{IF,2} - \omega_{IF,1}$ are modelled
- Local oscillator (LO) leakage is not modeled here

Mixer Non-Linearity: third-order intermodulation (IM3)

- Conversion of mixer power gain to linear coefficient:

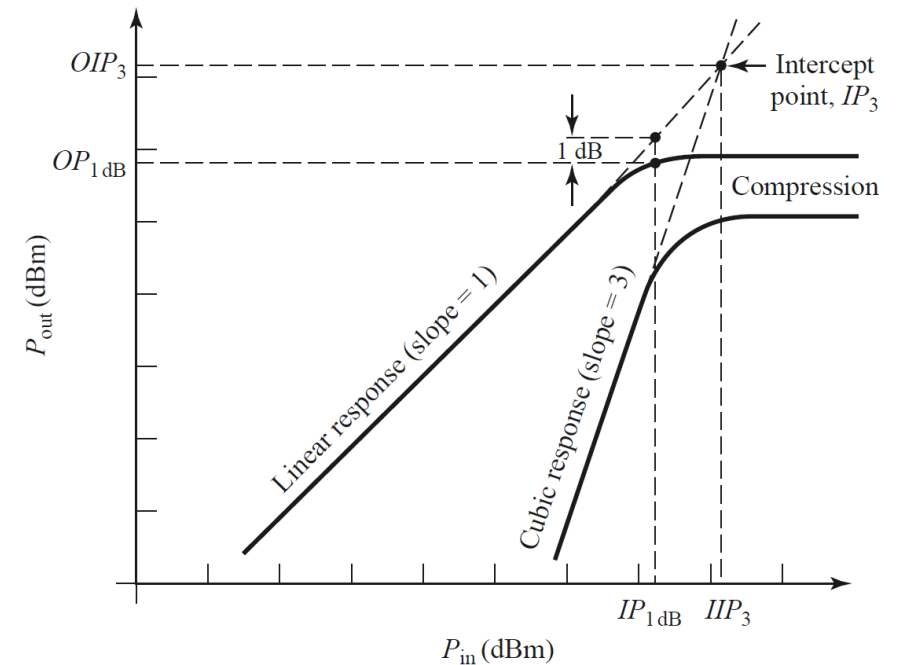
$$K_1 = 10^{\frac{G}{20}}$$

- All IF signals are scaled with K_1 , because the linear part of $y(t)$ models the ideal mixer operation
- K_3 can be computed from the third-order

intercept point (IIP3):

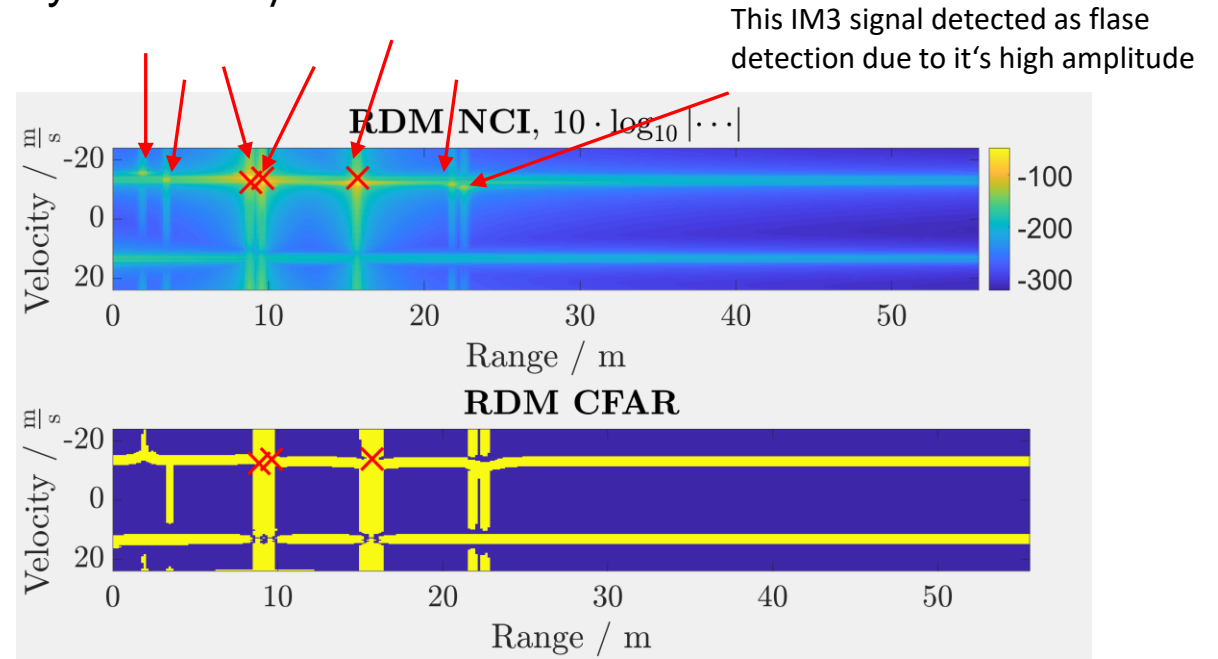
$$IIP3 = 20 \cdot \log_{10} \left(\frac{\sqrt{4 \cdot K_1}}{\sqrt{3 \cdot K_3}} \right)$$

$$K_3 = \frac{4 \cdot K_1}{3 \cdot 10^{-10} \cdot IIP3}$$



Mixer Non-Linearity: third-order intermodulation (IM3)

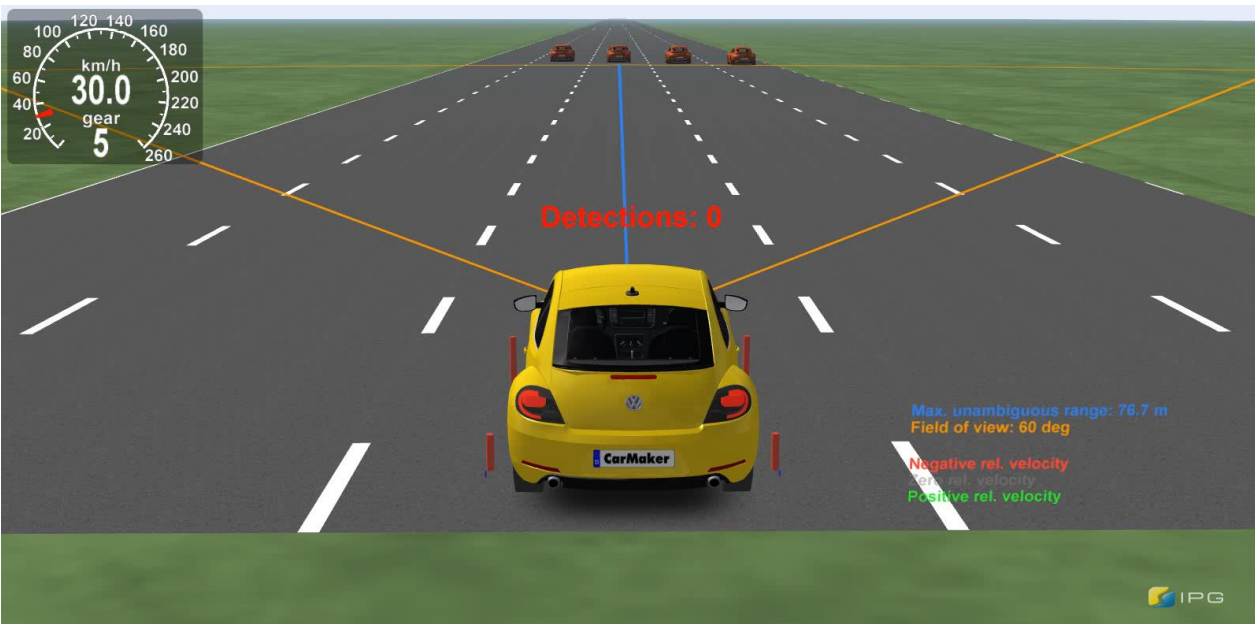
- In this scenario 3 trucks are placed at a distance of 8.9 m, 9.7 m and 15.8 m
- Ego vehicle approaching the targets with the velocity of 50 km/h



- $2 \cdot R_1 - R_2 = 7.65 \text{ m}$ (overlapped in RDM by R1 peak), $2 \cdot R_2 - R_1 = 10.62 \text{ m}$ (overlapping with target peak R2)
- $2 \cdot R_1 - R_3 = 1.55 \text{ m}$ (visible in RDM), $2 \cdot R_3 - R_2 = 22.82 \text{ m}$ (visible in RDM)
- $2 \cdot R_2 - R_3 = 3.53 \text{ m}$ (visible in RDM), $2 \cdot R_3 - R_2 = 21.83 \text{ m}$ (visible in RDM)
- Red crosses indicate VGT values

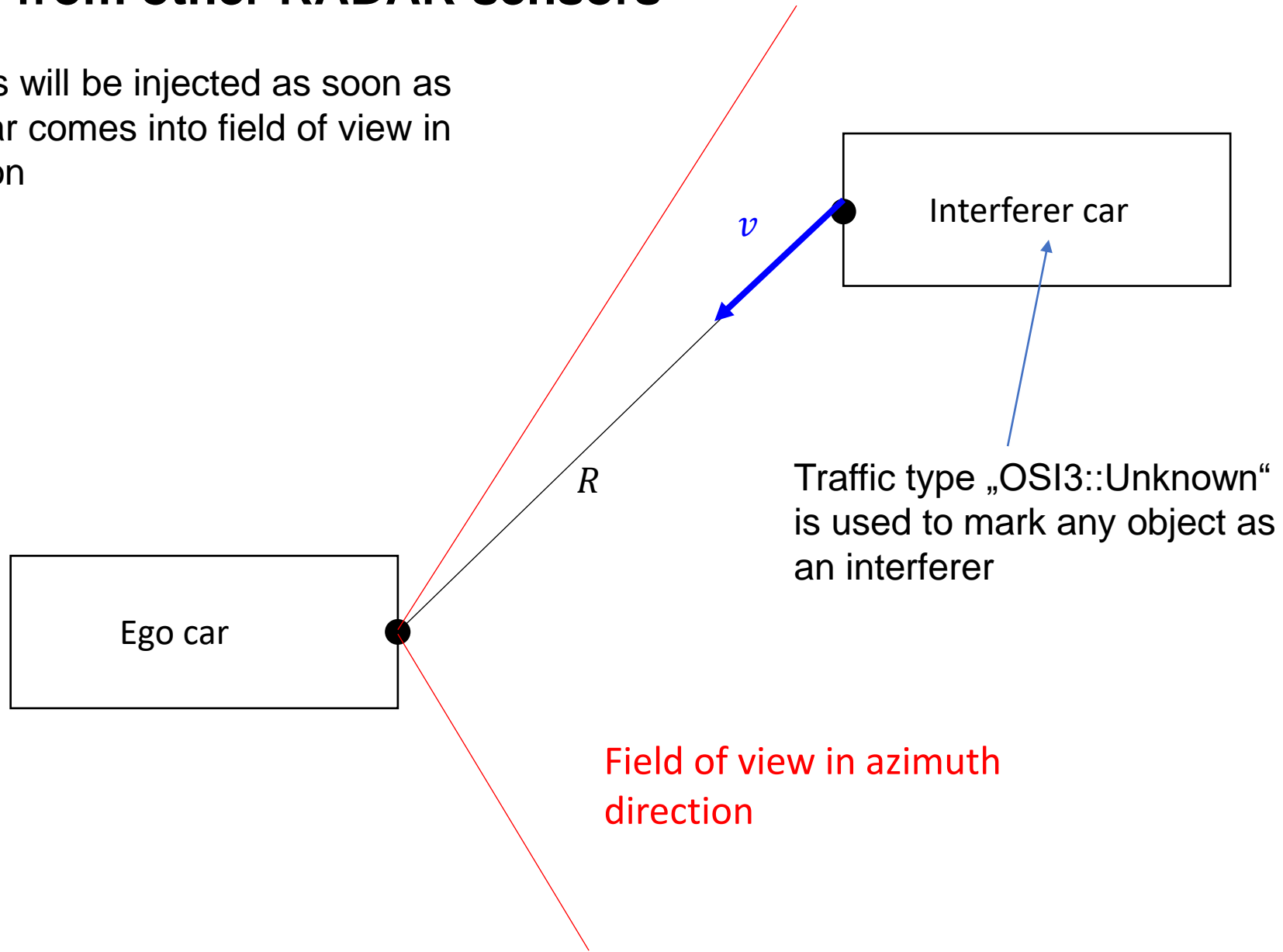
Mixer Non-Linearity: third-order intermodulation (IM3)

- Ego vehicle is approaching the targets with 30 km/h



Interference from other RADAR sensors

- Interferer ramps will be injected as soon as the interferer car comes into field of view in azimuth direction



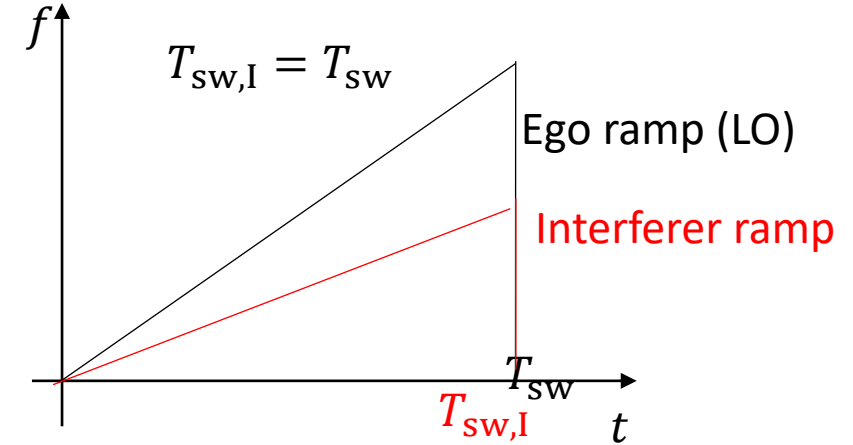
Traffic type „OSI3::Unknown“
is used to mark any object as
an interferer

Field of view in azimuth
direction

Interference from other RADAR sensors

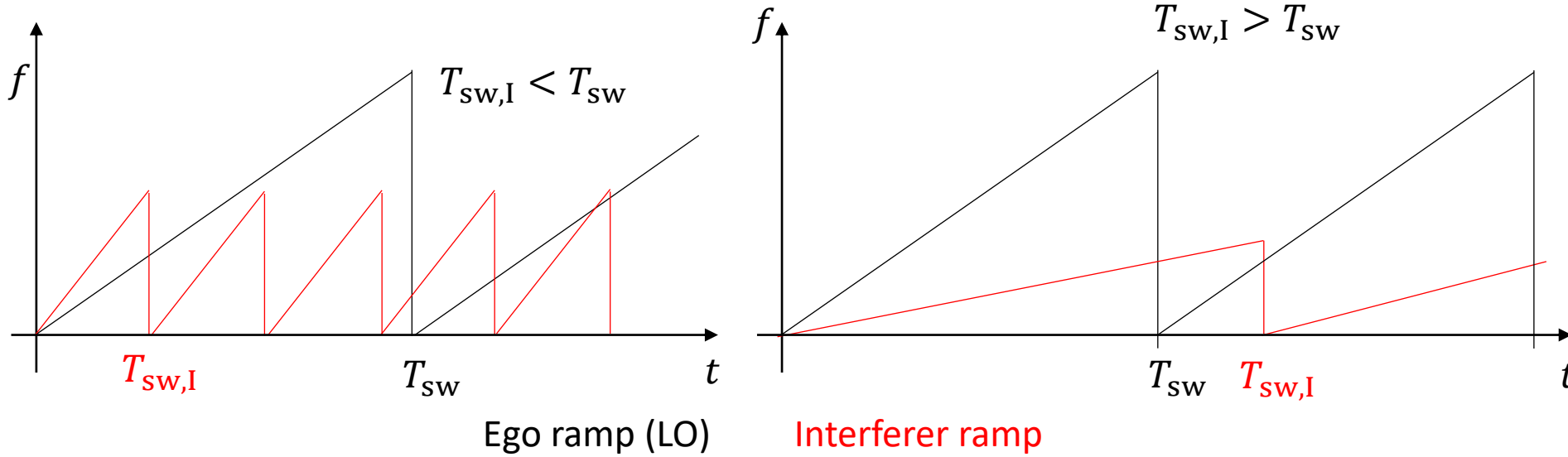
Same ramp duration for interferer and ego sensor

- $y_{IF}(t) = \frac{1}{2} A_{TX} A_{RX} \cdot \cos[2\pi(f_0 - f_I) \cdot t + \pi(k - k_I) \cdot t^2 + 2\pi f_I \tau_I + 2\pi k_I t \tau_I - \pi k_I \tau_I^2]$
- $\tau_I = \tau_{0,I} + \frac{v}{c} \cdot (i \cdot T_S + l \cdot T_{sw}) = \tau_{0,I} + \frac{v \cdot \Delta t}{c} = \tau_{0,I} + \frac{\Delta x}{c} = \tau_{0,I} + \Delta \tau$
- $\tau_{0,I} = \frac{R'_{n,m}}{c}$: time delay from interferer to ego car (one way)
- $i = 0, 1, \dots$: current sample in ego ramp, $l = 0, 1, \dots$: current ego ramp
- Instantaneous frequency: $f_{inst}(t) = \frac{1}{2\pi} \varphi'_I(t) = (f_0 - f_I) + (k - k_I) \cdot t + k_I \tau_I$
- Low pass filter: set samples of $y_I(t) \equiv 0$ if $f(t) \notin \left[-\frac{f_s}{2}, \frac{f_s}{2}\right]$, f_s : sample rate



Interference from other RADAR sensors

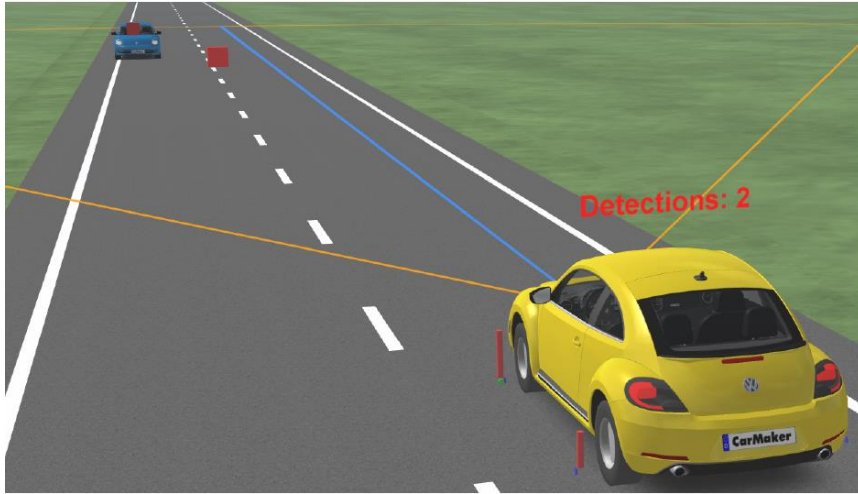
Different ramp duration for interferer and ego sensor



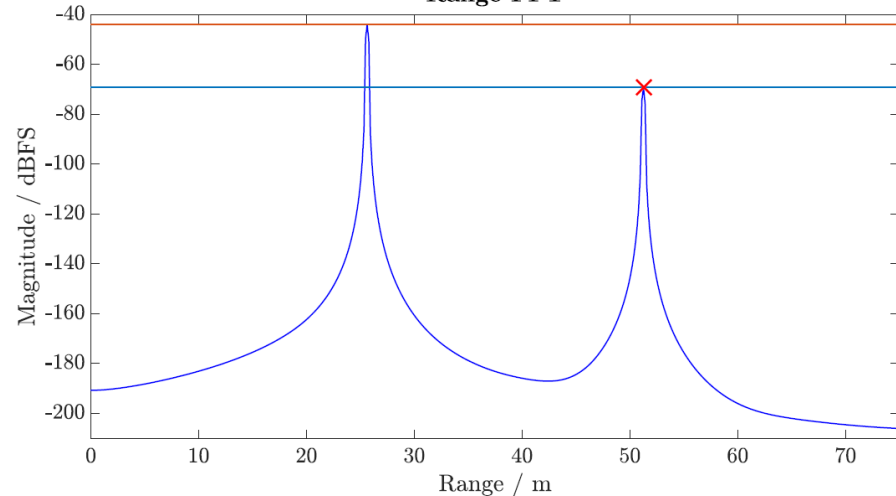
- Ego ramp (LO): $\varphi(t) = 2\pi f_0 \cdot t + \pi k \cdot t^2$, $0 \leq t \leq T_{sw}$, $t = i \cdot T_s$ i :
 current sample ego ramp $i = 0, 1, \dots, N_{\text{samples,ramp}} - 1$
- Interferer: $\varphi_I(t' - \tau) = 2\pi f_I \cdot (t' - \tau) + \pi k_I \cdot (t' - \tau)^2$, $0 \leq t' \leq T_{sw,I}$ $t' =$
 $(t + l \cdot T_{sw}) \bmod T_{sw,I} \equiv t + l \cdot T_{sw} - \left\lfloor \frac{t + l \cdot T_{sw}}{T_{sw,I}} \right\rfloor \cdot T_{sw,I}$, l : current ego ramp

Interference from other RADAR sensors

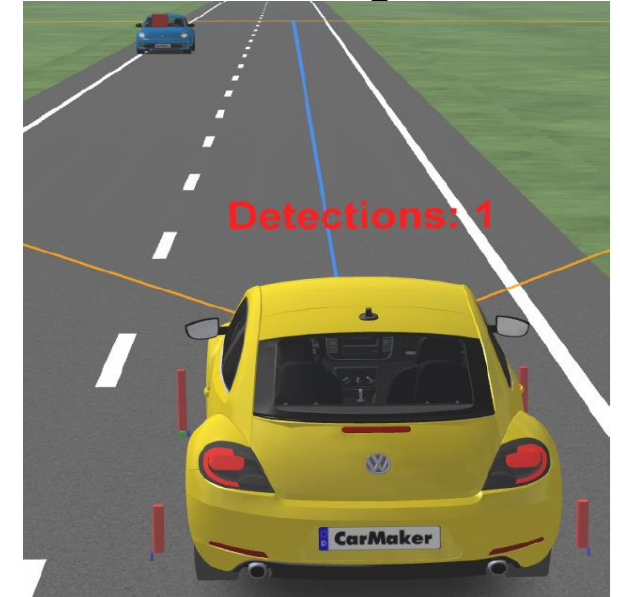
Same ramp duration for interferer and ego sensor



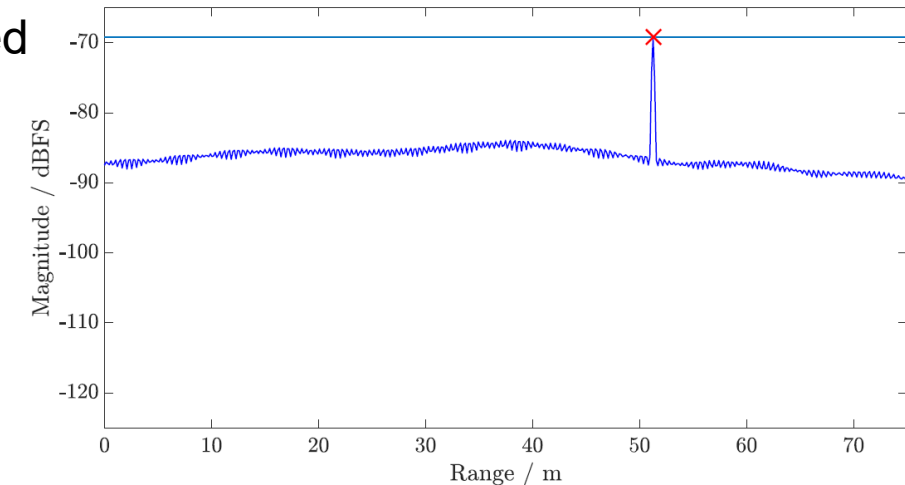
Range FFT



Different ramp duration for interferer and ego sensor



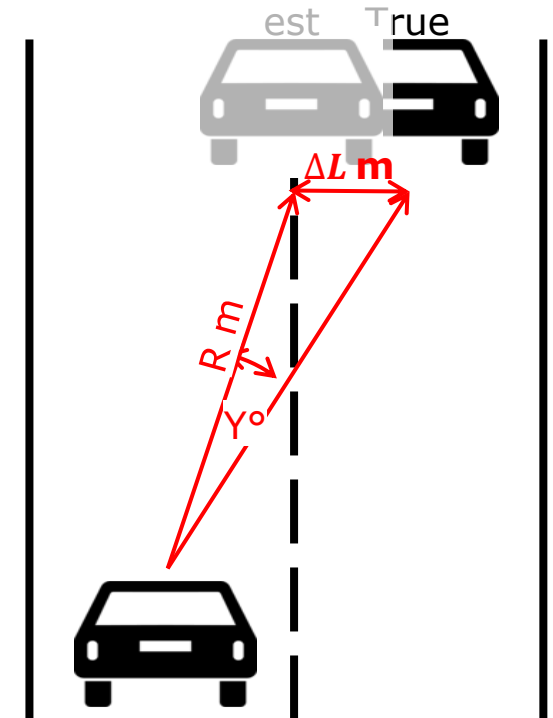
Range FFT



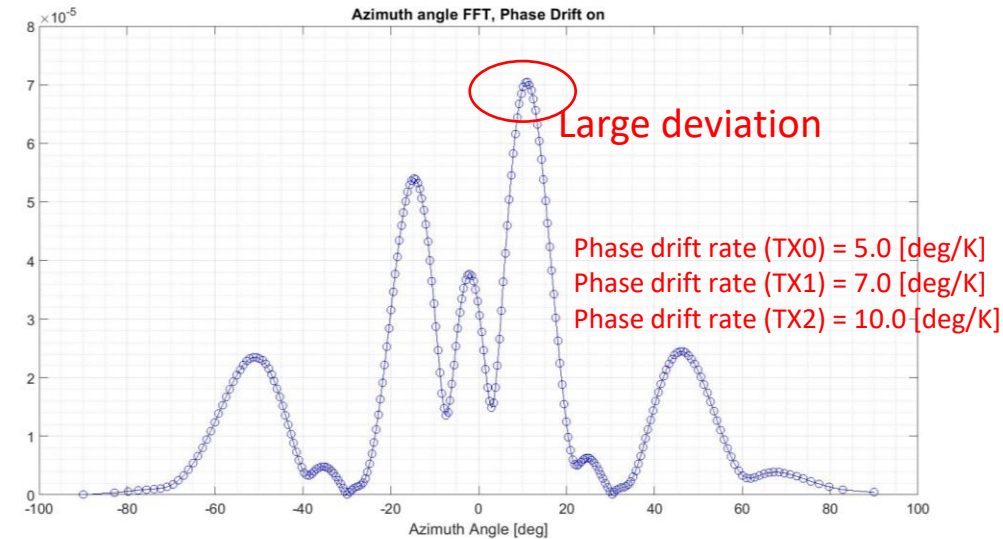
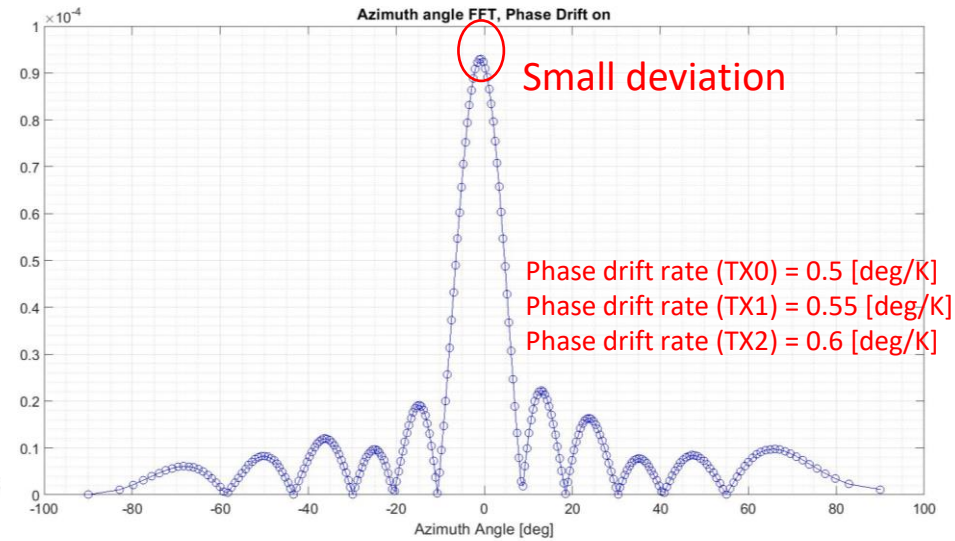
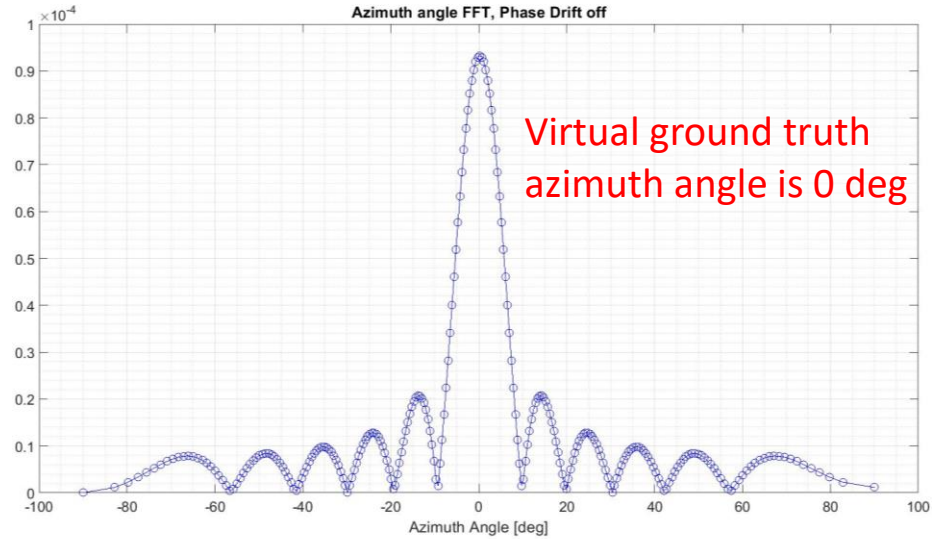
- **Same ramp duration:** ghost target is created at half the distance
- Red cross marks VGT distance of actual target
- Ghost object magnitude is higher since the power decays only with $1/R^2$ instead of $1/R^4$
- **Different ramp duration:** Non-uniform „noise“ floor is created

Phase Drift of TX and RX Channels

- **Phase Drift:** describes the change of the output phase of one TX channel (affecting also the phase balance), mainly over temperature
- **Phase balance** is the phase difference between the phases of two TX channels
- **Phase Drift Causes**
 - Angular estimation error,
 - **Sensitivity / SNR degradation in angular domain** (increased sidelobe level)



Phase Drift of TX and RX Channels



- $s_{IF}(t) = A_{IF} \cos(2\pi k\tau t + 2\pi f_c \tau - \pi k\tau^2 -)$
- Typical values of phase drift rate: 2.5°, 1.5°, 0.5°
- Typical values of initial phase imbalance: $\Delta\varphi_{initial, TX1-TX2}$: 3°, 5°, 10°

Conclusion

- The ray tracing-based RADAR sensor model is developed by using standardized interfaces OSI and FMI
- The sensor model includes the RADAR MMIC impairments, including phase noise, IM3, sensor interference, and phase drift
- These effects need to be considered to obtain realistic sensor model output
- The modeling of these impairments on the IF level makes simulation faster