Measurement Based Spatial Characteristics of 
MPCs in Train-to-Train Propagation 

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Abstract—Railway operators for high speed trains, commuter trains or subways request for technologies to increase the density and efficiency on their rails. One of those future technologies is virtual coupling based on train-to-train (T2T) wireless communication links. For the design of a wireless communication system, the knowledge of wave propagation is crucial. To characterize the propagation effects between two moving trains a T2T channel sounding measurement campaign was performed. To post-process the data, we use the Kalman enhanced super resolution tracking (KEST) algorithm to detect and track multipath components (MPCs). Using the single-input single-output measurement data recorded while the transmitter and receiver were in motion, the complex amplitude, delay and the life time of individual MPCs can be estimated. First results of the post-processing step, the geometrical representation of MPCs using point scatterers and possibilities to identify scattering positions are presented in this paper.

Index Terms—train-to-train, high speed train, propagation, measurement, KEST, path tracking.

I. INTRODUCTION

Railway operators for high speed trains (HSTs), commuter trains or subways request for new technologies to increase the density and efficiency on their rails. One of those future technologies is virtual coupling. Nowadays, trains are connected with a mechanical coupler and electrical hard wired connectors. In comparison, virtual coupled trains will be connected only via a wireless communication link. Consequently, virtual coupling requires an ultra reliable low latency communication (URLLC) system between two or more trains. Low latency requirements can be fulfilled with direct links such as train-to-train (T2T), similar to car-to-car (C2C) communication. For design and testing, channel models are essential but hardly investigated for T2T. An overview of measurements and channel models is given in [1]. To characterize the propagation effects between two moving trains for frequencies above 1 GHz and large bandwidths, a T2T channel sounding measurement campaign was performed. A unique data set for different environments, maneuvers and speeds was collected.

In case of multiple-input multiple-output (MIMO) based channel sounding measurements as performed in [2], the complex amplitude, delay, angle of departure (AoD) and angle of arrival (AoA) can be estimated. Hence, a geometrical representation of multipath components (MPCs) by scatterers in a 3D space can be achieved. For single-input single-output (SISO) measurements, with a fixed transmitter (Tx) and moving receiver (Rx) a virtual array approach can be applied as described in [3]. Hence, the complex amplitude, delay and the AoA can be estimated.

For SISO measurements, with moving Tx and moving Rx as described in [4], the virtual array approach is not feasible. Therefore, neither AoD nor AoA information can be extracted. In this context we provide first results on the spectral characteristics of MPCs and explain how this information might be used to locate objects influencing wave propagation. The Kalman enhanced super resolution tracking (KEST) algorithm is applied to detect and track individual MPCs.

The paper is structured as follows: In Section II we present the parameter estimation done by KEST and in Section III possible geometrical relations between consecutive snapshots. The T2T measurement campaign is briefly introduced in Section IV. We provide first results in Section V and conclude the paper in Section VI.

II. PARAMETER ESTIMATION

The measured channel impulse response (CIR) \( h(t, \tau) \) at the time instant \( t \) and delay \( \tau \) can be modeled as a sum of Dirac impulses

\[
 h(t, \tau) = \sum_{i=0}^{N} \alpha_i(t) \cdot \delta(\tau - \tau_i(t)),
 \]

where \( \alpha_i \) stands for the complex amplitude and \( \tau_i \) for the delay of the \( i \)th path [5].

KEST is a super resolution algorithm to estimate and track the channel parameters as introduced in [6]. The estimated parameters are the complex amplitude \( \alpha_i(t) \) and the delay \( \tau_i(t) \) of each path \( i \) for each snapshot at time \( t \). The tracking of individual paths provides the behavior of MPCs over time. In other words, the life time and the fading of paths can be analyzed. The KEST algorithm itself uses a Kalman filter (KF) that utilizes the output of a maximum likelihood (ML) estimator as measurements for tracking the parameters over time. In our case we use the space-alternating generalized expectation-maximization (SAGE) algorithm introduced by [7] as ML estimator. The model order detection is done at time instants \( t \) using the posterior estimation results from a bank of KF using information criteria [6].
III. Geometrical Representation of MPCs

In case of a moving Tx and Rx as for T2T or C2C communication as presented in [8], the Doppler can be derived in two different ways. First, using the train velocities $\nu_T$ and $\nu_R$, the Doppler shift can be expressed as

$$ f_d(\varphi_T, \varphi_R) = \frac{\|\nu_T\| \cos(\varphi_T(t)) + \|\nu_R\| \cos(\varphi_R(t))}{c} f_c $$

(2)

AoD is represented by $\varphi_T$ and AoA by $\varphi_R$; the carrier frequency $f_c$ is divided by the speed of light $c$.

Second, the Doppler shift can be described as the derivation of the delay change as shown in (3), with $x_S$ as the scatterer position vector [8]:

$$ f_d(x_S, t) = -\frac{\partial}{\partial t} \tau_i(x_S, t) \cdot f_c. $$

(3)

As well known, propagation paths with a propagation length of $\tau_i(t) \cdot c$ results from a single interaction point at scatterer position $x_c$ are located on an ellipse with their focuses at Tx and Rx positions. The ellipse is defined as

$$ a_{\tau_i}(t) = \frac{\tau_i(t) \cdot c}{2} $$

(4)

and

$$ b_{\tau_i}(t) = \sqrt{\left(\frac{\tau_i(t) \cdot c}{2}\right)^2 - \left(\|x_T - x_R\| / 2\right)^2} $$

(5)

where $a_{\tau_i}(t)$ is the semi-major axis and $b_{\tau_i}(t)$ the semi-minor axis [9]. The Tx with the position vector $x_T$ and Rx with its position vector $x_R$ are in the focuses of the ellipse. To further determine $x_S$, geometrical relations for individual MPCs are considered.

A. Geometrical relations between consecutive snapshots

In the following, we will discriminate between scatterers that are fixed in their positions and time dependent scatterers. First, we consider a point scatterer like a tree or a mast of the catenary. The scattering position remains the same over time (i.e. $x_S$ time independent) while the Tx and Rx positions change. The resulting phase shift of one path from one snapshot to the next snapshot results in a Doppler shift. If we draw ellipses related to the delay of one MPC for each snapshot, the ellipses will intersect in the position of the point scatterer $x_S$ which is static.

Second, for a moving scatterer position like on a flat wall of a house or noise barrier, the triangular relation between Tx, Rx and the scatterer remains similar over time. If both trains drive with the same speed in the same direction, the Doppler will be close to 0 for this kind of MPC.

B. Geo-information from maps and pictures

An easy way to identify scatterer points is to use geo-information from e.g. Google-maps or from videos or pictures. For example, the possible scatterer points can be plotted on top of a map and correlated to objects on the map. The maps can be supplemented with pictures obtained from the measurement campaign.

IV. T2T Measurement Campaign

In April 2016 Trenitalia [10] and the German Aerospace Center (DLR) performed the world wide first T2T C-Band channel sounding measurement campaign with two HSTs. Within four nights we measured the wireless propagation in various environments for different maneuvers on the high speed railway (HSR) between Naples and Rome. The main measurement equipment was the DLR RUSK channel sounder. The channel sounder was set up at a center frequency $f_c = 5.2$ GHz with a bandwidth of 120 MHz.

In the third night, T2T propagation with omni-directional antennas was measured. Overtaking and crossing maneuvers in different environments like sub-urban, rural and tunnels were performed. One data set with an overtaking maneuver in a rural environment is used for the presented results in the later sections. In case of the overtaking maneuvers, the Tx train was driving with a speed of 50 km/h and the Rx train with a speed varying between 10 km/h to 70 km/h. Further information can be found in [4].

V. Results

The following results are based on the measurements from the third night with omni-directional antennas. Two blocks of an overtaking maneuver, with 1200 snapshots each, are picked out and presented. The whole maneuver took place in rural environment; the Rx train with a speed $\nu_R = 10$ km/h is in front of the Tx train with a speed of $\nu_T = 50$ km/h. In the first evaluation block the link distance is around 1250 m and in the second block of around 60 m as shown in Figure 2 and 3.

A. Velocity and Distance Estimation

To measure the time variant CIR, stationarity must be ensured within the measurement of one snapshot [11]. Therefore, the absolute velocity of the Tx train $\nu_T$ and the Rx train $\nu_R$ as well as the resulting relative velocity $\Delta \nu$ in (6) need to be chosen. In combination with the time period of the transmitted
<table>
<thead>
<tr>
<th>Setting</th>
<th>$t_r$ [µs]</th>
<th>$t_p$ [µs]</th>
<th>$\nu_{max}$ [m/s]</th>
<th>$\nu_{max}$ [km/h]</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1.024</td>
<td>12.8</td>
<td>28.15</td>
<td>101.3</td>
</tr>
<tr>
<td>2</td>
<td>2.048</td>
<td>25.6</td>
<td>14.08</td>
<td>50.7</td>
</tr>
</tbody>
</table>

signal $t_p$ and the time between two consecutive snapshots $t_r$, (7) and (8) should be fulfilled.

$$\Delta \nu = \max (|\nu_T - \nu_R|, |\nu_T + \nu_R|)$$  \hspace{1cm} (6)

$$t_p \cdot \Delta \nu < \lambda$$ \hspace{1cm} (7)

$$t_r \cdot \Delta \nu < \frac{\lambda}{2}$$ \hspace{1cm} (8)

where $\lambda$ stands for the wavelength of the transmitted signal. For the used carrier frequency $f_c = 5.2$ GHz, $\lambda$ equals to 5.77 cm.

The given values for $t_p$ and $t_r$ and resulting maximum velocities $\nu_{max}$ are listed in Table I. As mentioned before, the Tx train was driving at a constant speed of 50 km/h. The Rx train varied the speed from 10 km/h to fall behind, up to 70 km/h to overtake the Tx train. To obtain time variant speeds, the global navigation satellite system (GNSS) measurements were performed and post processed. With an update rate of 1 Hz of the GNSS information, every 976 snapshot the velocity and distance can be updated. Due to the large inertial mass of the trains, the change of velocity within 1 s is rather small.

The Rx velocity information is used to calculate the moved distance of the Rx between two consecutive measurement snapshots. The time $t_r$ is stable due to the use of rubidium clocks at Tx and Rx. Therefore, the main limiting factor for the distance estimation accuracy is the velocity estimation of the GNSS receivers.

The distance between Tx and Rx can be derived in two ways. First, the distance is calculated by the GNSS positions and, therefore, limited in accuracy. Second, the distance is estimated by the absolute delay of the line of sight (LOS) of each CIR. It is only possible when the LOS path can be accurately tracked by the KEST algorithm.

### B. Path Detection and Tracking

The results of the KEST algorithm for the far and short distance link are shown in Figure 5 and Figure 7 respectively. In comparison, the CIRs of both measurements are plotted in Figure 4 and Figure 6. In all four figures the delay $\tau_i$ is plotted over the moved distance of the Rx. The color of each point indicates the power in dB.

The measurement bandwidth is limited to 120 MHz. The signal spreading in delay of the LOS path, visible in Figure 4 and Figure 6, results from the rectangular transmitted signal in frequency domain, i.e. a sinc function in time domain. In Figure 4 all MPCs follow clear trends and can be seen easily.

In comparison, for the short range measurements the MPCs can not be clearly visually extracted in Figure 6.

The four MPCs displayed in Figure 5 will be described in more detail in the following. The LOS path is continuously tracked starting with a delay of 4.27 µs which is related to the link distance of 1.28 km. Fading of around 15 dB can be observed on the LOS path amplitude. MPCs close to the LOS path with additional delay of 5 to 15 ns might be caused by obstacles of the train structure. Two further MPCs were tracked with similar delay change as the LOS path with an additional offset of 0.26 µs and 0.55 µs. Those MPCs might be caused by the noise barriers located parallel to the track.
The second evaluation presented in Figure 7 shows four dominant paths. The LOS path is tracked well with a delay of 0.22 µs, i.e. a link distance of 66.1 m at the start. The power varies over a range of 25 dB. Two MPCs are tracked with a similar delay slope being parallel to the LOS starting at 0.37 µs and 0.4 µs. One MPC is starting at 0.37 µs and ends at 0.43 µs while the delay increases over the moved distance.

VI. CONCLUSION

In this contribution we presented first results of the post-processing of T2T channel sounding measurements in HSR environments. The KEST algorithm was used to estimate and track the channel parameters of MPCs over time. The results for two link distances of around 60 m and 1.2 km in a rural environment were shown. We could prove that for both measurements the KEST algorithm is estimating the measured CIR and is tracking the MPCs over several meters.

Due to the SISO measurement setup and the movement of Tx and Rx, we could only estimate the complex amplitude and the delay with KEST. Therefore, the geometrical representation of MPCs using point scatterers and possibilities to identify scattering positions without angular information were presented.

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