

An Efficient Machine Learning-based Channel Prediction Technique for OFDM Sub-Bands

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- Channel state information (CSI)
- Channel estimation or **channel prediction**

To propose an efficient machine learning-based technique for channel prediction in OFDM sub-bands

- To deal separately with relevant aspects of **wave propagation** and antenna characteristics
- To evaluate the channel on a **two-dimensional plane** through which the mobile radio channel simulator consists of reflection points of the signal transmitted in the 2-D plane
- To address the birth and death problem of multipath clusters by allowing some **reflecting points to move randomly** during the channel simulation
- The simulated environment is such that there are **multipaths**;
- The **free space attenuation** is assumed for each multipath;

- The **relative position** of the transmitter \mathbf{P}_{tx} and receiver \mathbf{P}_{rx} , and initial position of the reflection points \mathbf{P}_r (randomly generated) in Cartesian coordinates;
- The **speed** vector of the receiver \mathbf{S}_{rx} and reflection points \mathbf{S} ;
- The total number of **reflection points** N_r and **mobile reflection points** N_m ;
- The carrier frequency f_c , the sampling frequency f_s , the bandwidth of the transmitted signal B , and the sampling time window p ;

- 1 \mathbf{P}_r , \mathbf{S}_{rx} , and \mathbf{S} are generated according to a Gaussian distribution
- 2 The length of a multipath $l_i = |\mathbf{P}_i - \mathbf{P}_{rx}| + |\mathbf{P}_i - \mathbf{P}_{tx}|$
- 3 The delay and phase of the multipath i are calculated as $\tau_i = l_i/c$ and $\phi_i = -l_i f_c/c \bmod 2\pi$, respectively,
- 4 The **channel impulse response**

$$I = \sum_{i=1}^N \frac{c}{4\pi f_c l_i} \exp(\phi_i + 2\pi D_i l_i) \delta(\tau_i - t) \quad (1)$$

- 5 Obtain the **frequency response** of the channel by the DFT
- 6 The positions of the receiver and other mobile points are updated

$$\mathbf{P}_{rx} \leftarrow \mathbf{P}_{rx} + \mathbf{S}_{rx} \Delta t \quad (2)$$

$$\mathbf{P}_r \leftarrow \mathbf{P}_r + \mathbf{S} \Delta t, \quad (3)$$

Table 1: Summary of Channel Simulation Settings.

Symbol	Value	Description
P_{tx}	$(-200, 0)$	Transmitter's relative starting position
P_{rx}	$(200, 0)$	Receiver's relative starting position
N_{r}	256	Number of reflection points
N_{m}	63	Number of mobile reflection points
f_{c}	900 MHz	Carrier frequency
f_{s}	51.2 MHz	Sampling frequency
B	12.8 MHz	Bandwidth of the transmitted signal
p	$10 \mu\text{s}$	Sampling time window
SNR	12 dB	Signal-to-Noise ratio at receiver
Δt	$500 \mu\text{s}$	Time between two consecutive simulations of the channel

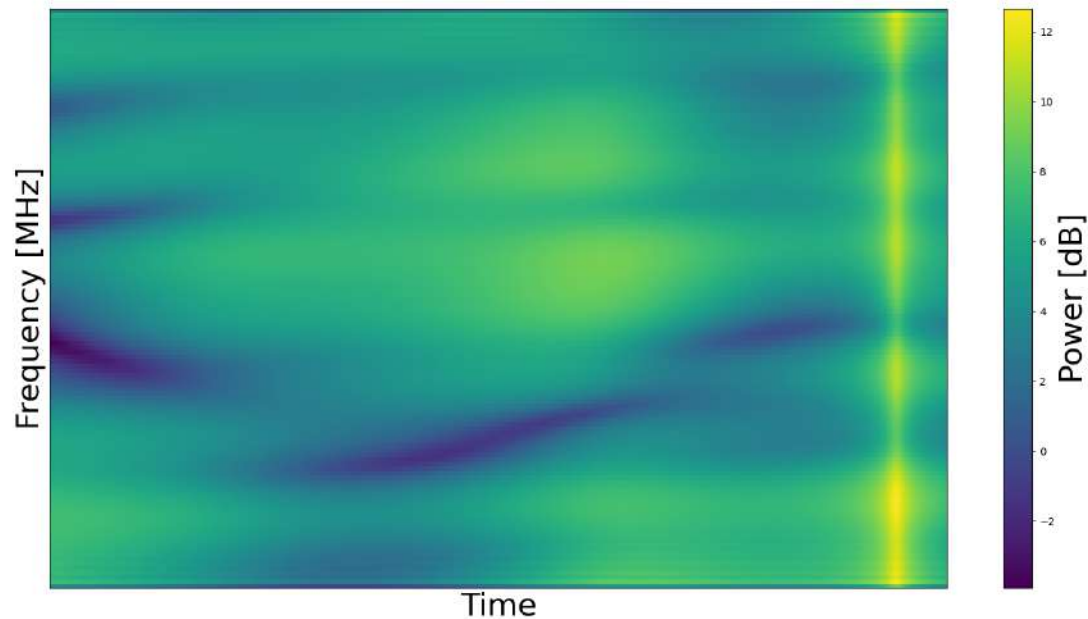


Figure 1: Frequency response of the channel over time.

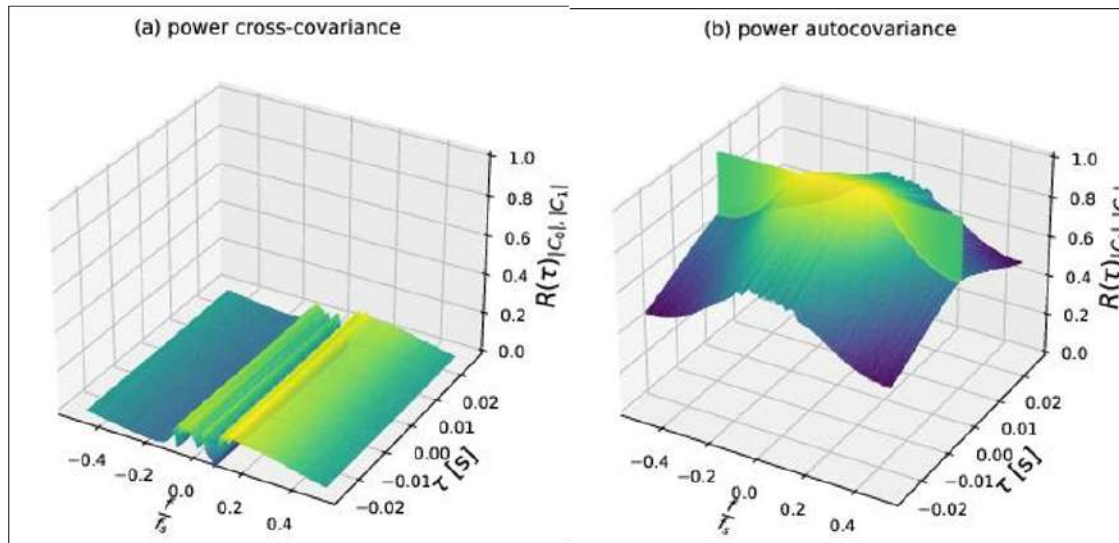


Figure 2: (a) Cross-covariance and (b) autocovariance over τ for the frequency band of interest.

- Predict D steps ahead of the channel frequency response (CNN Predictor)
- Predict whether the channel intensity will be below a threshold D steps ahead (CNN Classifier).

Table 2: Summary of the Characteristics of the Predictor Layers.

#	Channels	Kernel size	Dilation rate	Activation function
1	2	(3×10)	(1×1)	$\tanh(\cdot)$
2	3	(10×10)	(1×16)	$\tanh(\cdot)$
3	3	(10×10)	(10×1)	$\tanh(\cdot)$
4	2	(10×3)	(1×64)	$\tanh(\cdot)$
5	D	(1×64)	(1×1)	exponential

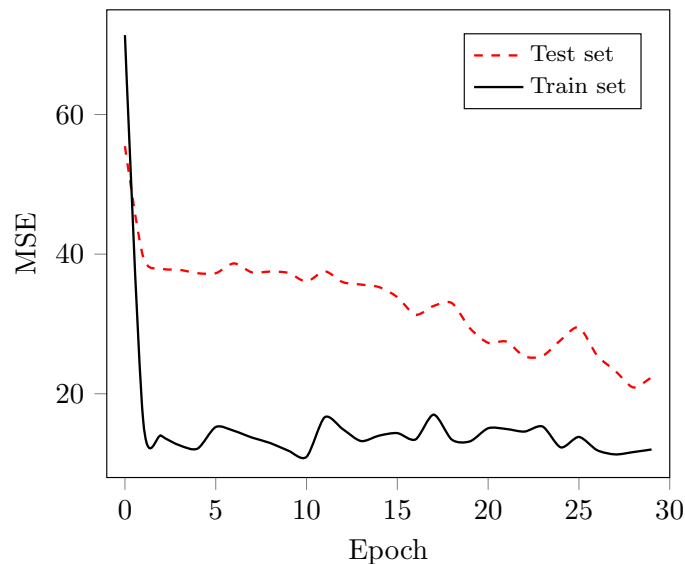


Figure 3: Mean Squared Error during training.

Predictor has smoother training and converges (≈ 8 training epochs)

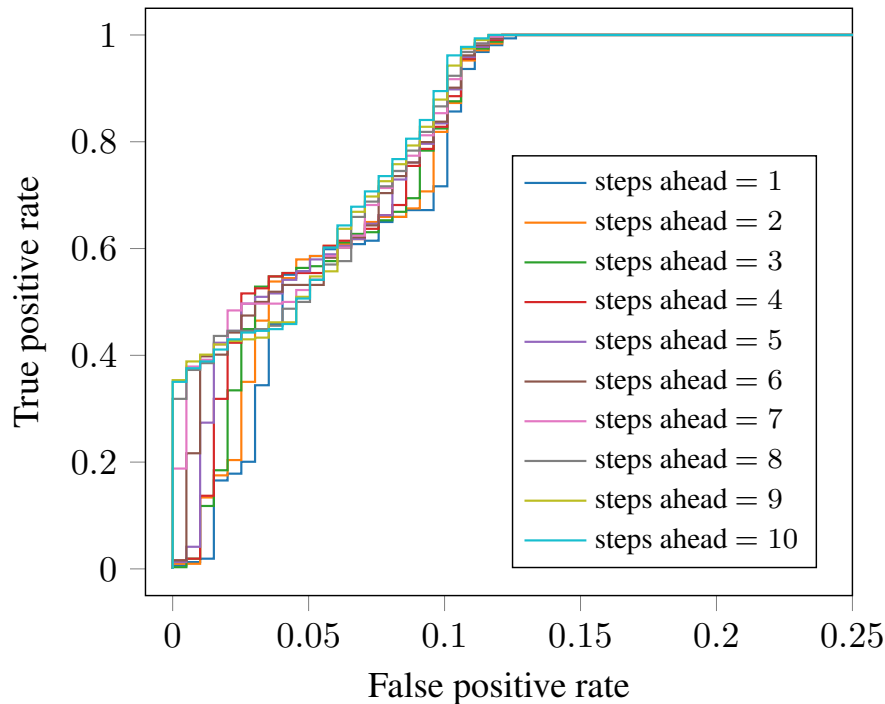


Figure 4: ROC curves for a single sub-band and for the different steps ahead.

Table 3: Mean Squared Error over the 4096 samples from a **new and uncorrelated channel**.

Step ahead	MSE
1	24.455336
2	22.610251
3	26.049115
4	24.488873
5	23.515360
6	23.826602
7	25.350836
8	25.708386
9	25.830940
10	26.221350

- A simplified way to simulate a mobile radio channel with fading has been proposed
- A CNN predictor and classifier have been successfully employed to predict the multi-step fading channel in wireless broadband systems
- Both the predictor and the classifier performed adequately when training and testing were performed on uncorrelated physical scenarios

- Include the antenna parameters
- Different channels, like Rician
- Find a better CNN

Thank you!

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