In-Service BER Based Estimation of OFDM PAPR and CFO-induced Peak Phase Deviation

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Abstract—High Peak-to-Average Power Ratio (PAPR) of the transmitted OFDM signal and the Carrier Frequency Offset (CFO) are well-known major drawbacks of the Orthogonal Frequency-Division Multiplexing (OFDM). Therefore, to avoid the High-Power Amplifier (HPA) inefficient operation with large back-off between the operating input power and its saturation region, specifically with the Long-Term Evolution (LTE) systems downlink, PAPR and CFO must be minimized. Considering that in many practical situations, this task demands (often unavailable) complex test equipment, such as e.g. Vector Signal Analyzer (VSA), in this paper, we develop a simple model for estimating the (residual) PAPR and CFO-induced peak phase deviation, by applying link abstraction, i.e. considering the easy-measurable BER degradation due to HPA non-linearity, as if it were the consequence of the according additive white Gaussian noise (AWGN). Moreover, the classic out-of-service BER testing, which requires network operator to interrupt its revenue-generating traffic, is substituted by in-service BER estimation from in-phase and quadrature-phase eye diagram closures, measured on live traffic, by means of a simple oscilloscope. The analytical model is verified by the appropriate Monte-Carlo simulations.

Index Terms— PAPR, OFDM, LTE, BER.

I. INTRODUCTION

Orthogonal Frequency-Division Multiplexing (OFDM) offers high-speed and excellent-performance data transmission over time-dispersive channels [1]. Therefore, it has been extensively used throughout the second consecutive decade, ever since it started to be deployed with the wireless local-area networks, through digital video broadcasting, and finally to the fourth-generation (4G) mobile communication systems – the Long-Term Evolution (LTE), in particular [2-5].

However, two main imperfections of OFDM systems still remain in focus when the OFDM transmission performance is in question, namely the Carrier Frequency Offset (CFO) that disrupts the orthogonality and so causes the Inter-Carrier Interference (ICI), as well as the excessive Peak-to-Average Power Ratio (PAPR) that is imminent even to the ideal OFDM signal, as its particular subcarriers combine one with each other either constructively, or destructively [6-7].

On the other hand, at the transmitter, large OFDM signal peaks pass through the High-Power Amplifier (HPA), which therefore is necessary to operate in its linear region, i.e. with large back-off between the operating input power and its value at which the HPA starts entering its saturation region.

Specifically with LTE downlink, some kind of PAPR reduction must be utilized, while in LTE uplink, the Single-Carrier-Frequency-Division Multiple Access (SC-FDMA), characterized by lower PAPR, and the mobile station – User Equipment (UE) operating at much lower power level, both imply that no PAPR reduction needs to be done.

However, as LTE-Advanced enables carrier aggregation in the downlink, even higher PAPR values can be expected, demanding wider linear HPA dynamic range, while the clustered SC-FDMA implies the same in the uplink [3].

With this regard, specifically with LTE-Advanced downlink, quick and simple OFDM PAPR estimation can be of interest. Therefore, in this paper the link abstraction principle is applied on the BER considered to be determined just by the HPA nonlinear distortion, i.e. provided that the channel is noiseless and time-dispersion-free (so that errors occur just due to HPA nonlinearity and the actual PAPR).

Finally, we extend the model to include the BER based estimation of CFO-induced peak phase deviation.

In Section 2, concepts of DPD and PAPR reduction are presented as well as relevant measurement examples. Then, in Section 3, we present the BER based PAPR, as well as CFO-induced peak phase deviation estimation, for the high-Signal-to-Noise Ratio (SNR) and multipath-free common AWGN channel that would have produced the same BER value. In Section 4, we review the in-service BER estimation from eye diagram closures measured on live traffic, using an oscilloscope. The analytical model is verified by means of Monte Carlo (MC) simulations, while conclusions are summarized in Section 5.

II. HPA NONLINEARITY COMPENSATION

Linear HPA operation implies low power efficiency:

\[ \eta = \frac{P_{\text{out}} - P_{\text{in}}}{P_{\text{DC}}} \]  

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where $P_{\text{out}}$, $P_{\text{in}}$ and $P_{\text{DC}}$ denote output and input powers, and the power taken from a battery, respectively, Fig. 1.

With low $\eta$ in linear regime, typically less than 10%, i.e. with more than 90% of the DC power dissipated into heat, the HPA operating costs become unacceptable, especially for mobile devices.

However, when an HPA operates in its saturation region, it introduces spectral regrowth that interferes with adjacent channels, Fig. 2.

Therefore the signal dynamic range must be accommodated to (mostly) fit the HPAs linear region either by digital predistortion (DPD) or by crest factor reduction (CFR).

Whatever the means to reduce OFDM signal peaks, the benefit is the possibility to increase the allowed average signal power, which implies the higher DC power usage efficiency, i.e. shifting the operating point to the right, and enabling longer distance coverage by a mobile station [1].

The effect on the Error Vector Magnitude (EVM) is presented in Fig. 3.

Various means to reduce PAPR in OFDM systems have been proposed so far [8-9], ranging from amplitude clipping [9-10], through coding methods [12], as well as the ones that introduce no distortion, with or without side information sent to the receiver [13-15].

III. TESTING PAPR REDUCTION

Testing PAPR without and with reduction can discover effectiveness of the applied PAPR reduction algorithm. It can be accomplished in a number of ways.

Specifically, the LTE lab with the state-of-the-art vector signal analysis (VSA) and Error Vector Magnitude (EVM) measurements [16] is the best solution for the analysis with this regard [17-18].

The exemplar PAPR complementary cumulative distribution function (CCDF) curve is presented in Fig. 4, where we can see that the probability that PAPR without CFR gets larger than 8.9 dB equals 0.1%, which holds already for the PAPR of 7.2 dB, with CFR applied.
However, in many practical situations no complex test equipment is available or affordable, so a simple and effective (residual) PAPR estimation can be of interest.

Although state-of-the-art wireless systems transmission performance specifications for network operators, such as the LTE ones, are block-error oriented, rather than bit-error based [4], still BER is inevitable performance indicator in research, product design and manufacturing, and is frequently and widely used. So, the idea here is to estimate the post-reduction PAPR by means of BER, which ultimately reflects the transmission performance [1].

Thereby, as common BER testing is much easier to accomplish than the VSA, it could be useful to relate PAPR directly to BER targets, rather than the EVM ones, as it follows.

A. BER based PAPR Estimation

Essentially, the wideband signals’ PAPR can be quite large, with the values of up to 13dB and more, so CFR can supplement DPD, significantly reducing the necessary back-off from saturated power, and thus increasing the HPA efficiency at no cost of signal degradation.

On this way, the distortion metric due to PAPR reduction is usually expressed in terms of EVM. Accordingly, for the CFR of the LTE-Advanced 20MHz downlink, the EVM target values for the QPSK, 16QAM and 64QAM are below 10%, 8% and 6%, respectively [3].

Let us visualize the relationship among the relevant factors of the PAPR reduction, which enables the HPA operating point (conforming to the average signal power) to be less backed off, where the saturation is simplified as beginning at once, rather than smoothly, Fig. 5. (We consider “smooth” transition (bending) not crucial for the PAPR model, but just the width of the linear dynamic range.)

As it can be seen in Fig. 5, ideally, if the whole dynamic range in between the symbol average and peak power, i.e. the actual PAPR value (expressed in decibels), is lesser or equal to the HPA operating point back-off, then there should be no non-linear product at the HPA output.

However, when the maximal-amplitude symbols enter into the negative-back-off zone, then, as already mentioned above, the HPA non-linear distortion occurs, usually expressed as the peak normalized EVM for the largest symbol amplitudes, relative to the ideal (reference) k-th symbol power [17]:

\[ EVM_{k_{max},M-QAM} = \frac{s_{k_{max},M-QAM} - s_{k_{max},M-QAM}'}{s_{k_{max},M-QAM}} \quad 1 \leq k \leq M \]  

(2)

where \( s_{k_{max},M-QAM}' \) is the HPA-compressed symbol \( s_{k_{max},M-QAM} \).

So, for known HPA operating point back-off, PAPR (in dBs) can be estimated from EVM (in dBs) as it follows:

\[ PAPR_k = 10 \cdot \log \left( \frac{BER \cdot OFF_{BACK} + EVM_{k}}{10} \right) \quad 1 \leq k \leq M \]  

(3)

Here we consider BER as being determined just by HPA distortion, as strong signal implies large SNR, while also no time dispersion (and so no multipath fading and the consequent inter-symbol interference) is present due to long enough cyclic prefix.

Now we recall the relationship between the EVM and the HPA-distortion determined BER [17]:

\[ EVM(BER) = \sqrt{\frac{3}{M-1}} \cdot \frac{1}{Q^{-1}(BER \cdot \log_2 M \cdot 4)} \]  

(4)

where \( Q^{-1} \) is the inverse of the Gaussian tail function, and M-QAM transmission over the AWGN channel is observed [7].

Finally, by substituting (4) into (3), we obtain the PAPR (in dBs):

\[ PAPR(BER) = 10 \cdot \log \left( \frac{BER \cdot OFF_{BACK} + EVM_{k}}{10} \right) \]  

(5)

B. BER based Estimation of CFO-induced Peak Phase Deviation

Furthermore, the CFO can create cumulative modulation symbols phase errors, which need to be measured. However, without complex equipment needed to perform constellation analysis, such as Vector Signal Analyzers (VSA) [3] this is not an easy task to do, and simple checking of the achieved CFO compensation, i.e. estimating whether the residual CFO-caused phase error is appropriate, might be very useful in many practical situations ranging from research and product development and manufacturing, to the exploitation phase with LTE network operators.

With this regard, specifically with LTE-Advanced downlink, quick and simple OFDM phase deviation estimation can be of interest. Therefore, in this paper the link abstraction principle is applied on the residual Bit-Error-Rate (BER) that is considered to be determined just by the CFO, i.e. provided that the channel is noiseless and time-dispersion-free (so that errors occur just due to actual CFO).

Fig. 5. Relationship between PAPR and EVM, for given back-off [1]
Moreover, recently, it has been shown that the maximal CFO-caused squared phase deviation is linear with the instantaneous (per-OFDM-symbol) PAPR. This implies that any PAPR reduction technique, such as e.g. simple clipping or coding, indirectly suppresses the CFO-induced phase deviation, too [6]. On the other hand this enables determining the CFO-induced phase deviation simply from BER degradation by applying the link abstraction method.

Thereby, it could be useful to relate the peak CFO-induced phase error to BER, by using the CFO’s relationship with PAPR [6] that is expressed by BER applying link abstraction principle.

As it can be seen in Fig. 6, CFO produces parasitic amplitude and phase modulation, since the CFO-induced error vector $E_{V_k}$ modulates the amplitude and the phase of the received nominal (ideal) symbol $\delta_k = s_k e^{j\phi_k}$ into $\hat{\delta}_{k,n,\Delta CFO}$, where the maximal error $\Delta \Phi_k$ of the latter occurs with $E_{V_k}$ perpendicular to the nominal $k$-th symbol vector $\delta_k$.

As it is obvious in Fig. 6 that $\frac{E_{VM_k}}{s_k} << 1$, the phase deviation can be approximated as:

$$\Delta \Phi_k \approx \frac{E_{VM_k}}{s_k} = \Delta \Phi_{CFO} \cdot \frac{E_{VM_k}}{s_k} = \Delta \Phi_{CFO} \cdot \frac{1}{s_k} ; \quad 1 \leq k \leq M$$

so that the peak CFO-induced (squared) phase deviation is:

$$\Delta \Phi_k^2 \approx \left( \Delta \Phi_{CFO} \cdot \frac{1}{s_k} \right)^2 \sum_{m=k}^{M} \frac{s_m e^{j\phi_m}}{m-k} ; \quad 1 \leq k \leq M$$

(7)

whose maximal value $\Delta \Phi_{k,\text{max}}^2$ is shown to be linear with the instantaneous (per-OFDM-symbol) PAPR [1]:

$$\Delta \Phi_{k,\text{max}}^2 \approx 3 \cdot \text{PAPR}_k \cdot \left( \frac{E_{VM_k}}{s_k} \right)^2 ; \quad 1 \leq k \leq M$$

(8)

Now, let us relate $\Delta \Phi_{k,\text{max}}^2$ to BER.

Finally, we substitute (5) into (8) to obtain the relationship between $\Delta \Phi_{k,\text{max}}^2$ and BER as it follows:

$$\Delta \Phi_{k,\text{max}}^2(\text{BER}) = 30 \cdot \left( \frac{\Delta \phi_{CFO,MT} M}{s_k^2} \right)^2 \cdot \log_{10} \left( \frac{E_b}{N_0} \cdot \log_2 M \right)$$

(9)

IV. PRACTICAL IN-SERVICE BER ESTIMATION

The well-known BER expression for the M-QAM signal transmission over AWGN channel is [18]:

$$\text{BER} = \frac{4}{\log_2 M} \cdot Q\left( \sqrt{\frac{2 E_b}{N_0} \cdot \log_2 M} \right)$$

(10)

where $Q$ denotes the Gaussian tail function, Fig. 7.

However, in practical measurements with digital transmission systems, the eye diagram in-phase (I) and quadrature-phase (Q) closures - $I_{CLO}$ and $Q_{CLO}$, respectively, are used to express a general overview of a condition of a radio system, primarily the margin against noise and other impairments, represented by the percentage of modulation symbols separation in I and Q directions, respectively [18].

I and Q eye diagrams can be obtained by means of a simple dual-beam oscilloscope, by synchronizing its time base with data clock, Fig. 8, and setting up infinite screen persistence with analog scopes, or cumulative data acquisition mode, with digital ones, Fig. 9.
So, to estimate BER ($E_b/N_0$) in (10), we need to calculate $E_b/N_0$ from closure. It has been shown that the closure is related to $C/N$ as it follows [18]:

$$\frac{C}{N} = 20 \log \left( \frac{100 \cdot C_1}{\sqrt{I_{CLD}^2 + Q_{CLD}^2}} \right)$$

(11)

where the constant $C_1$ represents the rms power of the carrier, and depends on the actual modulation scheme (e.g. $\sqrt{10}$ for 16 QAM).

Furthermore, as:

$$\frac{E_b}{N_0} = \frac{C}{N} \cdot B_{N} \cdot T$$

(12)

then, taking into account (11), (12) can be rewritten as:

$$\frac{E_b}{N_0} = 20 \log \left( \frac{100 \cdot C_1}{\sqrt{I_{CLD}^2 + Q_{CLD}^2}} \right) \cdot B_{N} \cdot T$$

(13)

For the LTE receiver noise bandwidth of $B_N=1.25$ MHz, (13) can be presented in tabular form, which, for 16 QAM modulation, is Table 1.

This way, the closure values in (13) or Table 1 can be substituted into (9). So, e.g. for 16 QAM, where $M=16$, substituting (13) into (10), provides BER as it follows:

$$BER = 4 \cdot \sqrt{\frac{100 \cdot \sqrt{10}}{\sqrt{I_{CLD}^2 + Q_{CLD}^2}}} \cdot B_N \cdot T$$

(14)

Finally, for complete in-service PAPR estimation, (14) has to be substituted into (5).

V. VERIFICATION

In order to verify (5), it was necessary to conduct Monte Carlo (MC) simulations of various symbols belonging to the three LTE-relevant modulation types: 4-QAM, 16-QAM and 64-QAM, and perform uncoded BER tests without time dispersion (equivalent to the perfect CP-protected channel), and with high SNR [1].

A. Verification of PAPR Estimation

With back-off values ranging from 0 to 12 dB, the HPA non-linearity was introduced just to the maximal-amplitude outer symbols in the constellation, so reducing the PAPR and enabling bit error occurrences to be counted for the short-term BER calculation and checking (5).

In the simplified PAPR model, we considered that without normalization, the I and Q component of the maximal 4, 16 and 64-QAM modulation symbols’ amplitudes are equal to 1, 3 and 7, respectively. The normalization is performed by dividing the above values by $\sqrt{2}$, $\sqrt{10}$ and $\sqrt{42}$, respectively, so enabling the unit mean symbol energy [4], i.e.:

$$\mathbb{E}[s_k^2] = 1$$

(15)

This implies:

$$s_{k\text{, max }4\text{-QAM}}^2 = \frac{1}{2} (1^2 + 1^2) = 1$$

$$s_{k\text{, max }16\text{-QAM}}^2 = \frac{1}{10} (3^2 + 3^2) = 1.8; \quad 1 \leq k \leq M$$

(16)

$$s_{k\text{, max }64\text{-QAM}}^2 = \frac{1}{42} (7^2 + 7^2) \approx 2.33$$
The instantaneous (per-OFDM-symbol) PAPR, defined as [4]:

\[
PAPR_{\text{avg}} = \frac{P_{\text{peak}}}{P_{\text{avg}}} = \frac{\max \left\{ \frac{1}{2} \sum_{m=1}^{M} s_{m}^2 \right\}}{\sum_{m=1}^{M} s_{m}^2} = \frac{\max \left\{ \sum_{m=1}^{M} s_{m}^2 \right\}}{\sum_{m=1}^{M} s_{m}^2} \tag{17}
\]

can be regarded as determined by the maximal amplitude of the 64-QAM symbols only, providing the highest PAPR values:

\[
PAPR_{\text{absmax}} = \frac{M \cdot s_{\text{max}64-QAM}^2}{\sum_{m=1}^{M} s_{m}^2} \leq 1 \leq M \tag{18}
\]

On the other hand, more realistic PAPR values are obtained by accordingly modifying (17) to take into account not only the (rarely appearing) 64-QAM highest-amplitude symbols, but all maximal amplitudes’ values from (16), weighted by their probabilities of occurrence in the OFDM symbol, which can take various values, such as e.g. [19]:

\[
P(s_{k} \in 4-QAM) = 0.4
\]
\[
P(s_{k} \in 16-QAM) = 0.5; \quad 1 \leq k \leq M
\]
\[
P(s_{k} \in 64-QAM) = 0.1
\]

Accordingly, the corresponding PAPR plots for the three LTE-relevant modulations, and back-off values of 5, 10 and 15dB, are presented in Figures 10, 11 and 12, respectively.

Observed intra-figures, the below plots reflect dispersion among the modulation types for the same PAPR value, where the higher modulation order implies the larger BER. Moreover, observed inter-figures, it is obvious that for any modulation algorithm, the larger back-off values increase the corresponding tolerable PAPR values for the same BER value.

Finally, in Table II the exemplar PAPR vs BER relationship with 5 dB back-off for 4-QAM, 16-QAM and 64-QAM modulation schemes, are presented, as coming out of the estimation (5) and MC simulations.

As it can be seen from the table, excellent matching between the estimated and the MC-simulated PAPR values is found (with the tolerance to PAPR decreasing with the order of QAM), which verifies the model under the presumed conditions.

B. Verification of Estimation of CFO-induced Phase Deviation

Furthermore, the corresponding phase deviation plots for the three LTE-relevant modulations, and the CFO values of 5, 70 and 300 Hz, are presented in Figures 13, 14 and 15, respectively.
If compared intra-figures, the above plots reflect dispersion among the CFO values for the same modulation type, where the higher CFO implies the larger BER. Moreover, compared inter-figures, the higher modulation order is less tolerable to CFO for the same BER value.

Finally, in Table III, the exemplar $\Delta \Phi_{k_{\text{max}}}^2$ vs BER relationship with 5 dB back-off and 300 Hz CFO, for 4-QAM, 16-QAM and 64-QAM modulation schemes, are presented, as coming out of the estimation (5) and MC simulations. The table exhibits excellent matching found between the estimated and the MC-simulated BER values, with the tolerance to CFO phase deviation decreasing with the order of QAM. This verifies the model under the presumed conditions.
VI. CONCLUSION

In many practical situations, specifically with OFDM transmission over LTE-Advanced downlink, applying carrier aggregation, PAPR values can be excessive, so quick and simple PAPR estimation can be of interest. Therefore, in this paper the link abstraction principle is applied on BER that is considered to be determined just by the HPA nonlinear distortion, i.e. that errors occur just due to HPA nonlinearity and the inadequate PAPR vs back-off relationship.

Moreover, by capitalizing on its relationship with PAPR, the other dominant OFDM-inherent impairment - the CFO-caused phase deviation in particular, can also be estimated by means of common BER testing.

However, as network operators consider out-of-service BER testing not preferable, under such circumstances, we propose simple in-service BER estimation from I and Q eye diagrams, and then using the obtained value for the so achieved complete in-service PAPR estimation.

The proposed algorithm for estimating both PAPR and CFO-induced phase deviation from BER, is verified by its excellent matching with the corresponding MC-simulated scenario.

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