

Digital Predistortion for Wideband 5G Transmitters

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Digital Predistortion for Wideband 5G Transmitters

21st March 2019

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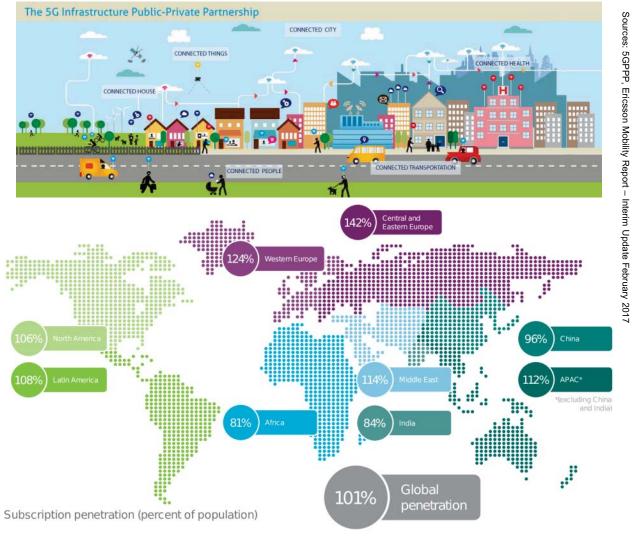
Outline

- Introduction
- Power amplifier characterization
- Power amplifier modeling for digital predistortion
- The digital predistortion technique
- DPD challenges & Solutions for 5G
- Conclusion



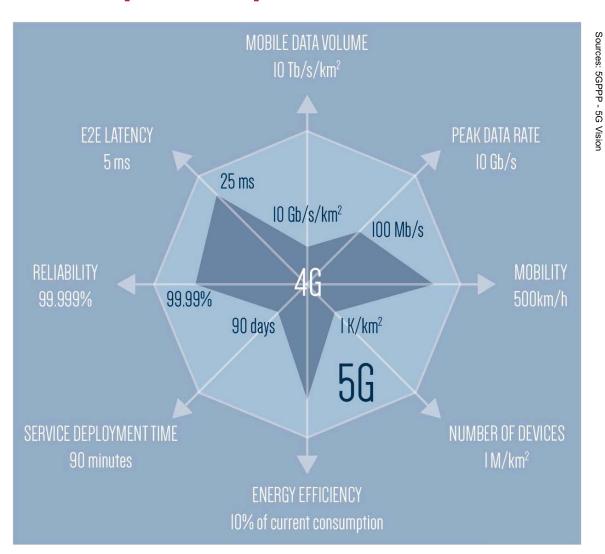


Wireless civilization





5G disruptive capabilities







5G Key enabling technologies

- Wide and contiguous spectrum bandwidth
- New flexible resource management and sharing schemes
- Flexible air interfaces
- New waveforms
- Advanced multi-antenna beam-forming and beamtracking and MIMO techniques
- Millimeter-waves
- Denser cells leading to ultra-dense networks





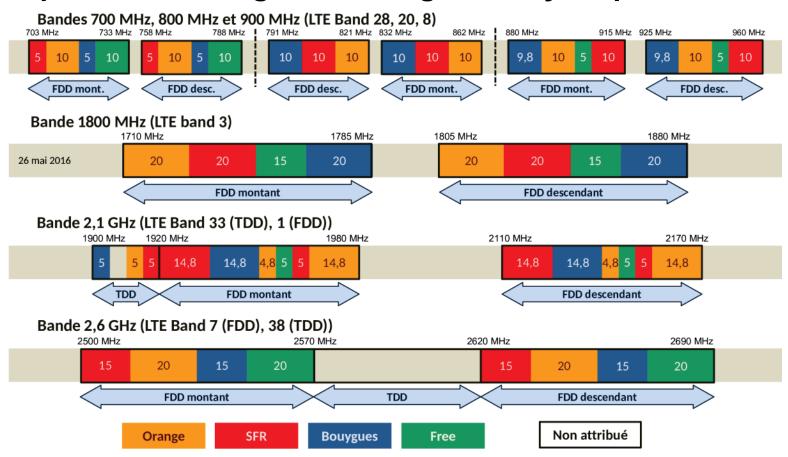
5G issue #1 : Spectrum scarcity

Spectrum allocation in France FIXE MOBILE SCIENTIFIQUE SATELLITE RADIOLOCALISATION MARITIME **AERONAUTIQUE** RADIOAMATEUR METEOROLOGIE RADIODIFFUSION Services primaires secondaires Appareils de faible portée



5G Issue #1 : Spectrum scarcity – « Zoom »

Spectrum sharing must be rigourously respected



Note: les éventuelles bandes de garde de 200 kHz, en haut ou bas de bande, ou entre les attributions de deux opérateurs adjacents ne sont pas mentionnées

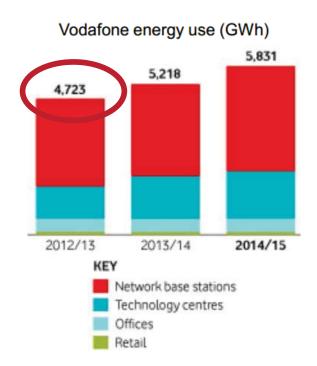




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5G issue #2: Energy consumption

Vodafone energy use (GWh)



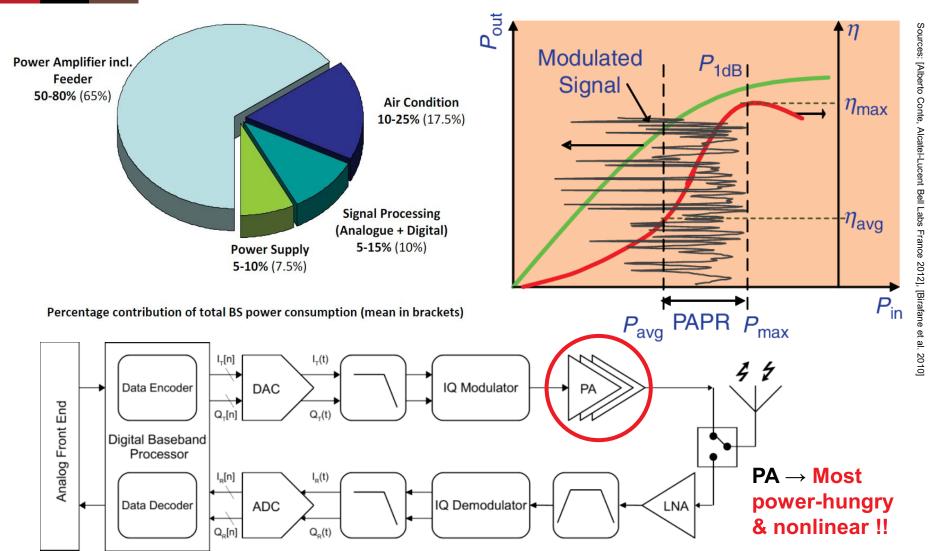


65% in BSs

6,043 2016 6,197 2017 2018 1.000 2,000 3.000 4.000 5,000 6,000 7,000 0 Offices Network base Technology Retail station sites centres 29% 65% 5% 1% base station sites retail stores technology centres offices

Sources: Vodafone Sustainability Reports 2015 & 2018

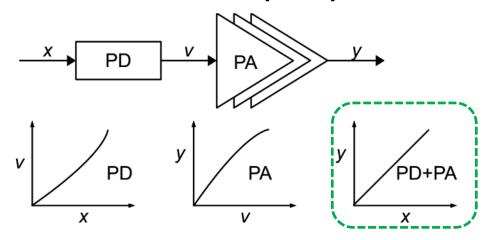
The power amplifier linearity/efficiency trade-off





Predistortion and Challenges in 5G

Predistortion principle



Predistortion challenges in 5G BSs

- High signal bandwidths (>100 MHz) → Memory Effects ↑
- Spectral efficient modulation formats → PAPR ↑
- Energy efficient → Nonlinearity ↑
- Low-cost → Cheaper PA with Nonlinearity ↑



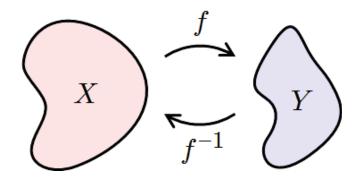


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Predistortion with « inverse » functions

Definition

- g(x) is inverse of f(x) when g(f(x)) = x
 - -g(x) is usually denoted $f^{-1}(x)$ by mathematicians



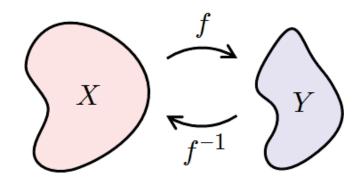
- -g(x) does not always exist! Particularly true for nonlinear functions.
 - g(f(x)) = x is usually possible only for a limited range of x

○ Example :
$$f(x) = x^2$$
 ; $g(x) = \sqrt{x}$ only for $x \ge 0$ (Note the different « nature » of $f(\cdot)$ and $g(\cdot)$)





Consequence on predistortion process



- For DPD systems we only search for approximate inverse
- We need to « know » f(x) to find its inverse g(x)
 - First, find an adequate approximation of f(x)
 - Second, find an approximate inverse of f(x)





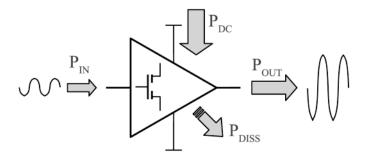
Power amplifier characterization



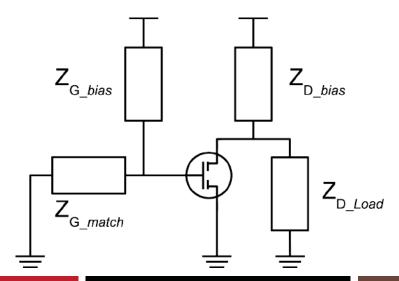


Considered power amplifier

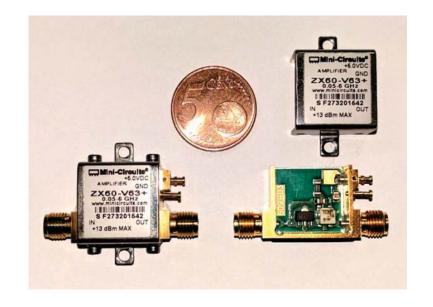
Abstract view



General circuit model



Device example





PA characteristics – more details

Actual PA example TYPICAL CHARACTERISTICS — 2110-2200 MHz

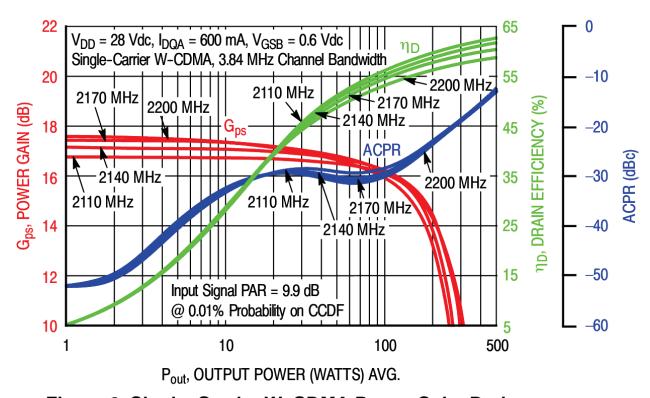


Figure 6. Single-Carrier W-CDMA Power Gain, Drain **Efficiency and ACPR versus Output Power**

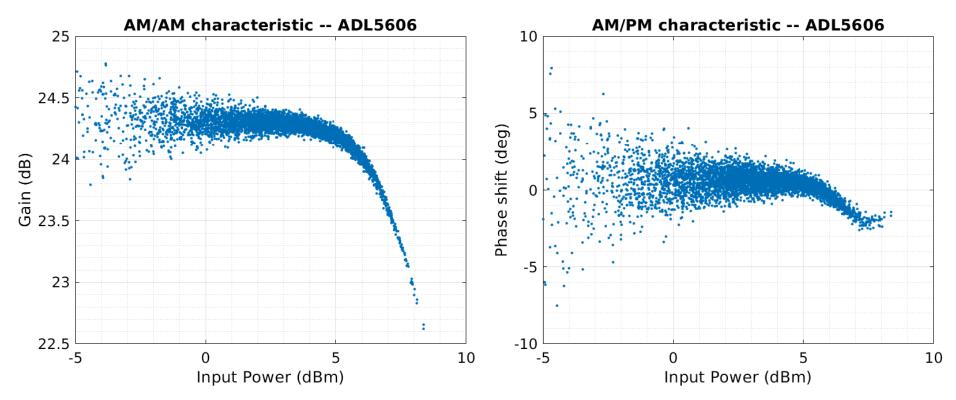
Source: NXP Semiconductors -- RF Power LDMOS Transistor - A3T21H360W23S





Dynamic characterization with Modulated Signals

Amplitude and Phase transfer function



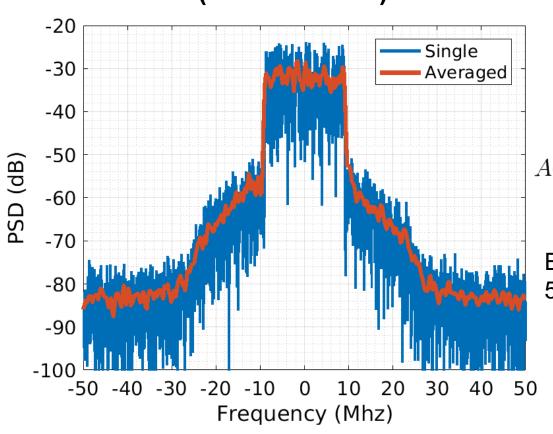
Measurement results with LTE signal BW= 20 MHz, PAPR=12,1 dB





Dynamic characterization with Modulated **Signals**

Adjacent Channel Leakage (or Power) Ratio (ACLR/ACPR)



$$ACPR_{dB} = 10 \log_{10} \left(\frac{\int_{BW \text{main}} P(f) df}{\int_{BW \text{adj}} P(f) df} \right)$$

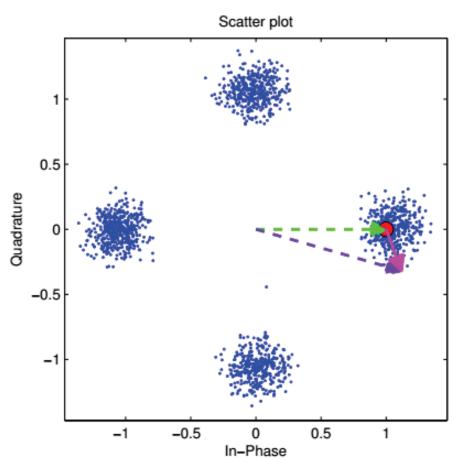
Example: 5G specifications ACLR > 45 dBc





Dynamic characterization with Modulated Signals

Error Vector Magnitude (EVM)



$$EVM(\%) = \sqrt{\frac{\frac{1}{N} \sum_{i=1}^{N} |S_{actual,i} - S_{ideal,i}|^2}{\frac{1}{N} \sum_{i=1}^{N} |S_{ideal,i}|^2}}$$

Example:

EVM<12.5% pour LTE-A pour 16-QAM





Linearity requirements for 3G/4G/5G base stations

| TABLE 1. Linearit | y rec | juirements in t | ypica | 13G/ | /4G base stations. |
|--------------------------|-------|-----------------|-------|------|--------------------|
|--------------------------|-------|-----------------|-------|------|--------------------|

| Standard Multiplexing Type | UMTS [37] WCDMA | WiMAX [38] OFDMA | LTE [39] OFDMA | LTE-A [40] OFDMA |
|---|--------------------|---------------------|--------------------------|---------------------|
| Single-channel bandwidth (MHz) | 5 | 1.25, 5, 10, 20 | 1.4, 3, 5, 10, 15, 20 | 20 |
| Maximum aggregated bandwidth (MHz) | 60 (12-band) | 20 | 20 | 100 (5-band) |
| In-band requirement EVM ^a (%) | < 12.5 | <6 | < 12.5 | < 12.5 |
| Out-of-band requirement ACLR1 ^b (dBc) ACLR2 ^c (dBc) | <-45 <-50 | <-45 <-50 | <-45 <-45 | <-45 <-45 |

^{a.} Based on the 16-QAM modulation scheme.

Source: 2014 - Guan, Zhu - Green Communications: Digital Predistortion for Wideband RF Power Amplifiers





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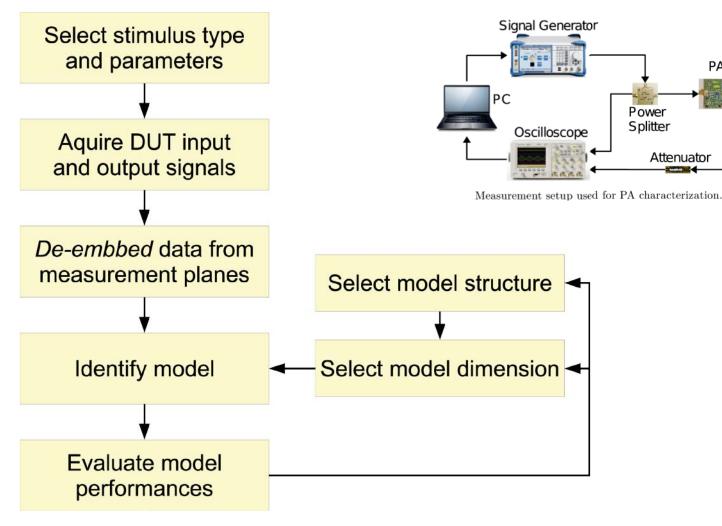
b. Refers to the first adjacent channel leakage power ratio.
c. Refers to the second adjacent channel leakage power ratio.

PA modeling for Digital Predistortion





Characterization and modeling method



Source: 2015 - Ghannouchi, Hammi, Helaoui - Behavioral Modeling and Predistortion of Wideband Wireless Transmitters



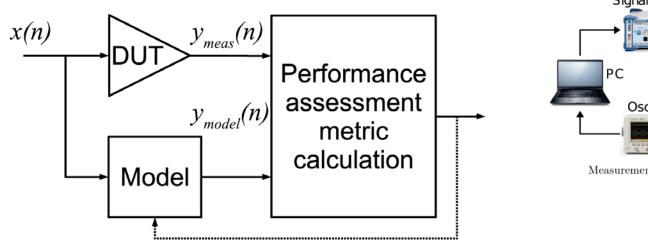


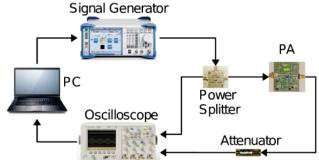
PA

Power Splitter

Attenuator

Modeling accuracy assessment





Measurement setup used for PA characterization.

Time domain metric

Normalized Mean Square Error

$$-NMSE = 10 \log_{10} \left(\frac{\sum_{\ell=1}^{L} |y_{model}(l) - y_{meas}(l)|^2}{\sum_{\ell=1}^{L} |y_{meas}(l)|^2} \right)$$





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Nonlinear models – the most popular

Baseband equivalent signal

$$x(t) = A(t) e^{j\theta(t)} (A(t), \theta(t) \in \mathbb{R})$$

Memoryless systems

$$y(t) = x(t) \cdot G\{A(t)\} = A(t) G_A\{A(t)\} e^{j(\Phi_G\{A(t)\} + \theta(t))}$$

Polar Saleh Model

$$G_A\{A(t)\} = \frac{\alpha_a}{1 + \beta_a A^2}$$

$$\alpha_{\Phi}$$

$$\Phi_G\{A(t)\} = \frac{\alpha_\Phi}{1 + \beta_\Phi A^2}$$

Polynomial

$$y(t) = \sum_{k=1}^{N} a_k |x(t)|^{k-1} x(t)$$

Odd order only Polynomial

$$y(t) = \sum_{k=0}^{N} a_k |x(t)|^{2k} x(t)$$





Nonlinear models – the most popular

Memory polynomial based models

Memory polynomial

$$y_{MP}(n) = \sum_{m=0}^{M} \sum_{k=1}^{K} a_{mk} x(n-m) |x(n-m)|^{k-1}$$

Odd orders only memory polynomial

$$y_{MP}(n) = \sum_{m=0}^{M} \sum_{k=0}^{K} a_{mk} x(n-m) |x(n-m)|^{2k}$$





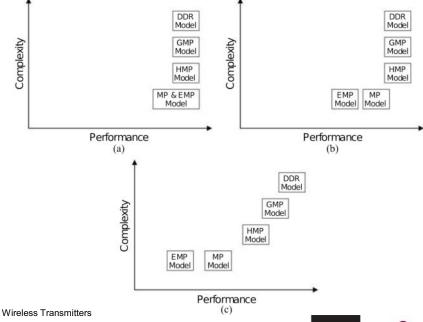
Nonlinear models – the most popular

Volterra series models

$$y_{\text{volterra}}(n) = \sum_{k=1}^{K} \sum_{m_1=0}^{M} \cdots \sum_{m_k=0}^{M} h_k(m_1, \cdots, m_k) \prod_{j=1}^{k} x(n-m_j)$$

Many variations

 Comparison between memory polynomial based models. (a) Weakly nonlinear memory effects, (b) mildly nonlinear memory effects, and (c) strongly nonlinear memory effects



Source: 2015 - Ghannouchi, Hammi, Helaoui - Behavioral Modeling and Predistortion of Wideband Wireless Transmitters





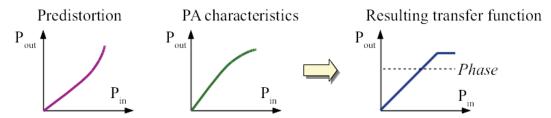
The digital predistortion technique



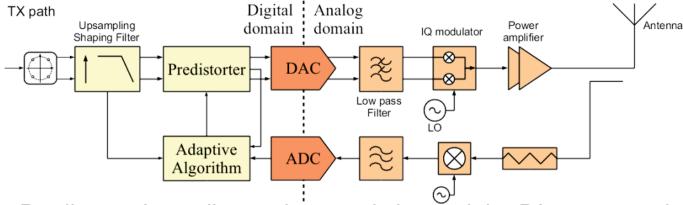


Fundamental elements

Principle (reminder)



Transmitter architecture



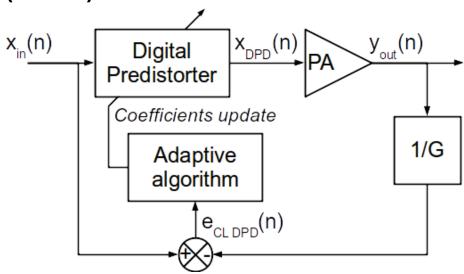
- Predistorter's nonlinear characteristics and the PA must match
- Nonlinearity of the PA varies with time due to changes in the drive signal, aging, or drifts
 - Track variations and update predistortion function





DPD architectures – Closed loop

Closed loop (direct) architecture



Advantages

- No modeling of the PA
- Compensates for time-varying effects

Drawbacks

- Suitable for memoryless systems
- No direct relation between error signal and predistorter's coefficients

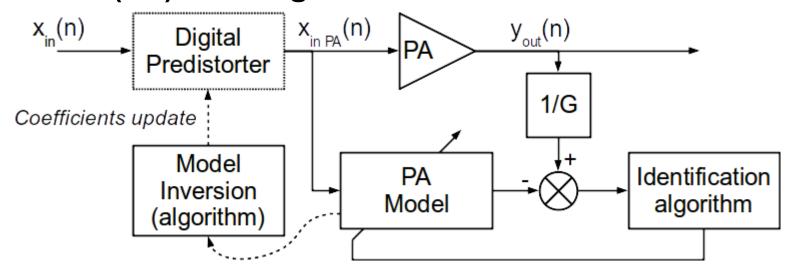
References:

- C. D. Presti, D. F. Kimball and P. M. Asbeck, "Closed-Loop Digital Predistortion System With Fast Real-Time Adaptation Applied to a Handset WCDMA PA Module," in IEEE TMTT; March 2012.
- Fadhel M. Ghannouchi, Oualid Hammi, Mohamed Helaoui, "Behavioral Modeling and Predistortion of Wideband Wireless Transmitters", John Wiley & Sons; Jul 2015



DPD architectures – Open loop

Direct (PA) learning architecture



« 2 steps » learning

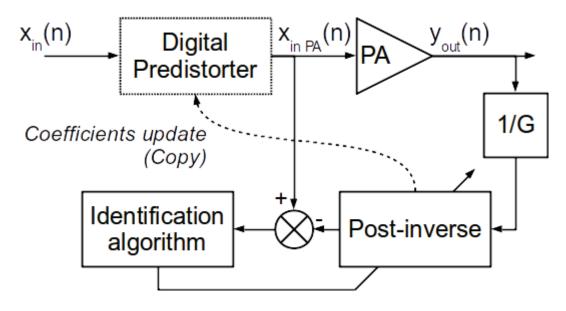
- PA model is identified first; then the inverse is derived from the PA
- A theoretical inverse model can be computed as the p-th order inverse
 - Suitable for memoryless systems





DPD architectures – Open loop

Indirect learning architecture



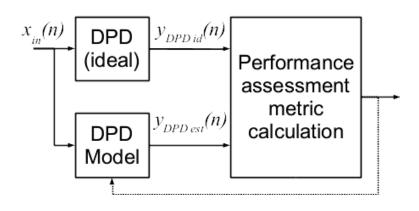
« 1 step » learning

- the PD function is directly derived by calculating the post-inverse of the PA
- Suitable for memory systems



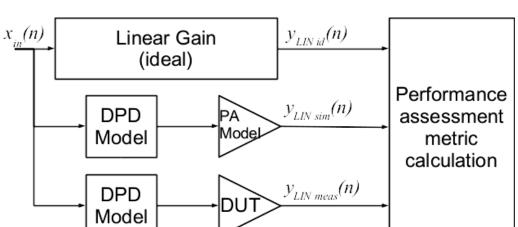


Computation methodology of the inverse



- Match $y_{DPD,id}$ and $y_{DPD,est}$
- Example of model oriented metrics
 - Normalized Mean Square Error

$$-NMSE =$$

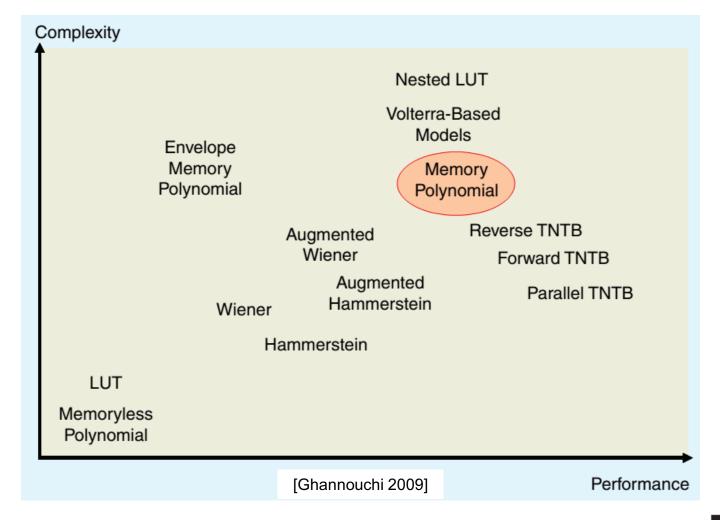


$$10\log_{10}\left(\frac{\sum_{\ell=1}^{L}|y_{est}(l)-y_{id}(l)|^{2}}{\sum_{\ell=1}^{L}|y_{id}(l)|^{2}}\right)$$





Conventional DPD models







Objective functions and computational aspects

- Least-squares (LS): $\min_{x \in \mathcal{X}} \|\overrightarrow{y}(n) \Gamma_{x}(n) \cdot A\|^{2}$
 - Common approaches: Moore–Penrose pseudo-inverse $A^{+} = (A^{H}A)^{-1}A^{H}$, QR decomposition, SVD
 - Significant computational complexity ($\mathcal{O}((M \times K)^3)$)
- Least-mean-squares (LMS): $\min_{i} E[|y(n) \overrightarrow{A}^{H} \cdot \overrightarrow{\gamma}_{x}(n)|^{2}]$
 - Iterative approach: $e(n) = y(n) \overrightarrow{A}^{H}(n) \cdot \overrightarrow{\gamma}_{x}(n)$ $A(n+1) = A(n) + \mu e^{*}(n) \overrightarrow{\gamma}_{x}(n)$
 - Reduced computational complexity ($\mathcal{O}(M \times K)$)
 - Convergence issues (µ)





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Objective functions and computational aspects

Recursive (weighted) least-squares (RLS):

$$\min_{\overrightarrow{A}} \sum_{i=0}^{k} \lambda^{k-i} |y(i) - \overrightarrow{A}^{H}(i) \cdot \overrightarrow{\gamma}_{x}(k)|^{2}$$
• Iterative approach: $e(k) = y(k) - \overrightarrow{A}^{H}(k-1) \cdot \overrightarrow{\gamma}_{x}(n)$

$$\overrightarrow{s}(k) = \mathbf{S}(k-1) \cdot \overrightarrow{\gamma}_{x}(n)$$

$$\overrightarrow{\kappa}(k) = \frac{\overrightarrow{s}(k)}{1 + \overrightarrow{\gamma}_{x}^{H}(n) \cdot \overrightarrow{s}(k)}$$

$$\mathbf{S}(k) = \frac{1}{\lambda} [\mathbf{S}(k-1) - \frac{\overrightarrow{\kappa}(k) \cdot \overrightarrow{\kappa}^{H}(k)}{\lambda + \overrightarrow{\gamma}_{x}^{H}(n) \cdot \overrightarrow{s}(k)}]$$

$$\overrightarrow{A}(k) = \overrightarrow{A}(k-1) + e^{*}(k) \cdot \overrightarrow{\kappa}(k)$$

- Increased computational complexity $(\mathcal{O}((M \times K)^2))$
- Robust convergence





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Conclusion

- Main learning architectures
 - DLA
 - ILA
- Choosing an appropriate minimization method requires a reasonable amount of knowledge of the specific identification problem
 - Stability, speed of convergence and implementation complexity can largely vary between the different methods.





DPD challenges & Solutions for 5G

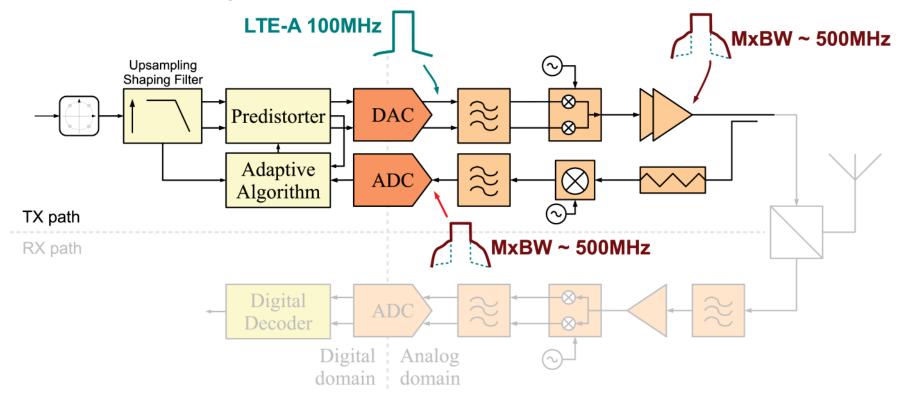




TX feedback path: the bottleneck

ADC bandwidth feedback limitation

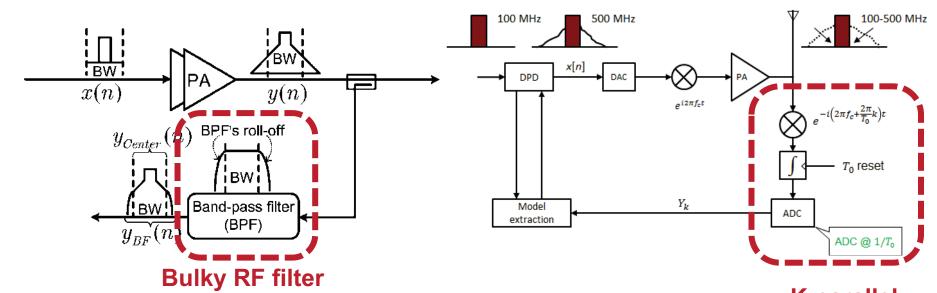
Learning phase





TX feedback path: Band limited and/or low rate DPD techniques

Band-limited feedback [Ma, 2014], [Zhang, 2015] Low rate identification [Hammler, 2014]



Lower ADC sampling rate





- Compute PA model first (
- Sampling frequency of PA model

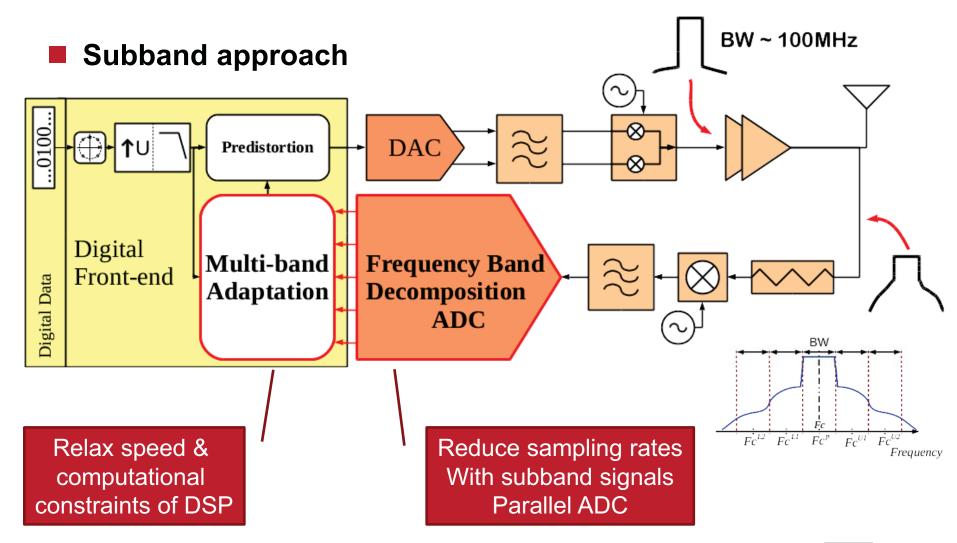








TX feedback path: New low rate DPD architecture

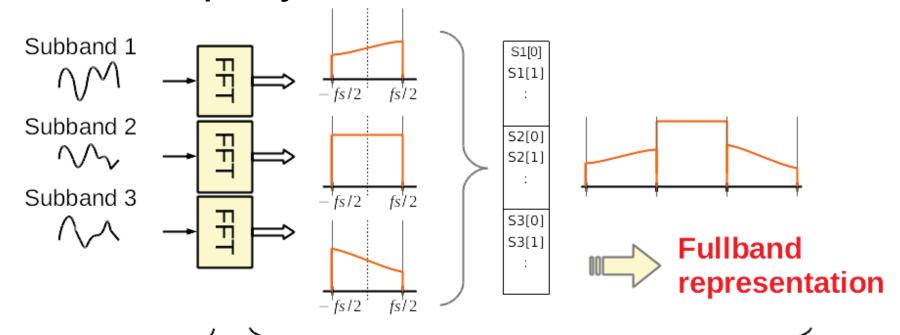






TX feedback path : Solution to the signal reconstruction

Frequency domain



Time-domain

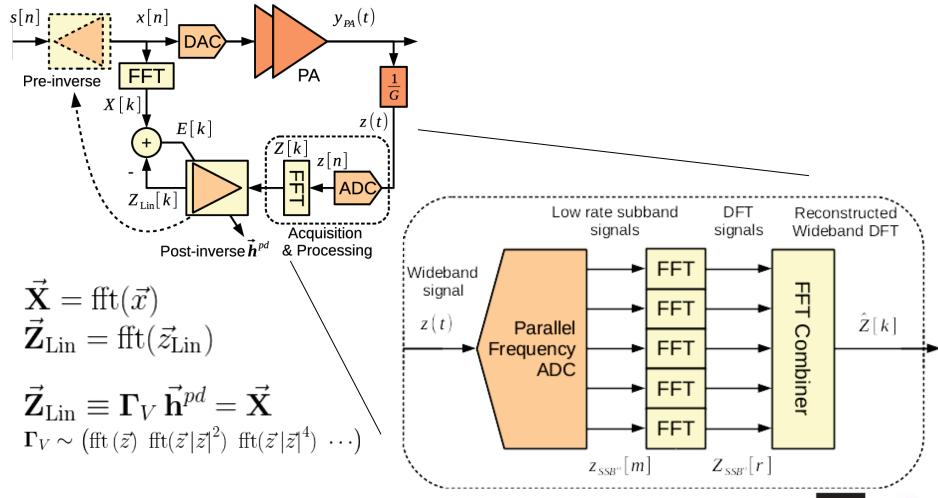
Low sampling frequency Frequency domain





TX feedback path: New subband DPD architecture

FFT-based subband digital predistortion

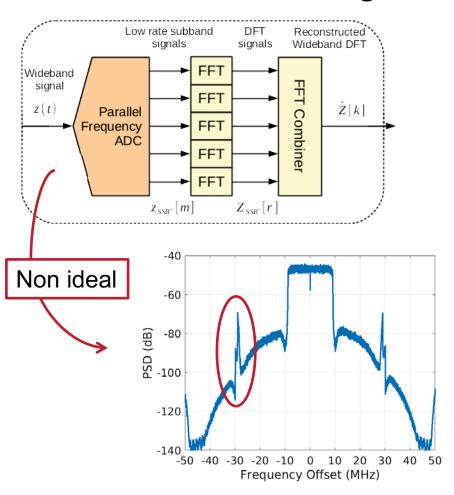


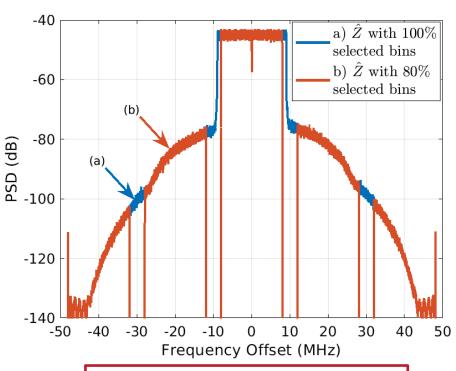


TX feedback path: FFT-based subband DPD

- Mitigating subband edge effects

Limited subband signal reconstruction





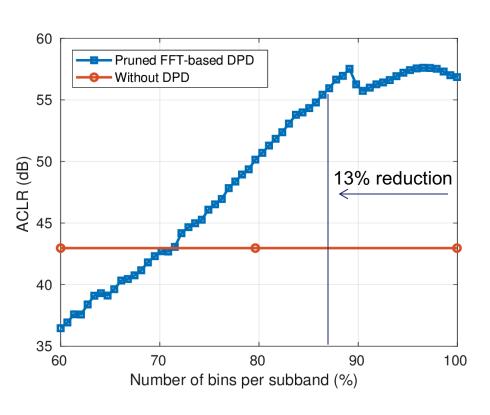
Exclusion of subband edges

$$ec{\mathbf{Z}}_{\mathrm{Lin}} \equiv \widetilde{\mathbf{\Gamma}_{V}} \overset{\overrightarrow{\mathbf{h}}^{pd}}{\mathbf{h}^{pd}} = \overset{\overrightarrow{\mathbf{X}}}{\mathbf{X}}$$
 $\widetilde{\Gamma_{V}} \sim \left(\widetilde{\mathrm{fft}(ec{z})} \ \widetilde{\mathrm{fft}(ec{z}|ec{z}|^{2})} \ \widetilde{\mathrm{fft}(ec{z}|ec{z}|^{4})} \ \ldots \right)$





TX feedback path: FFT-based subband DPD – Linearization performance



20 principal coeffcients (dB) 0 -10 -30 -40 of PD -50 MSE -60 -70 70 80 90 100 60 Number of bins per subband (%)

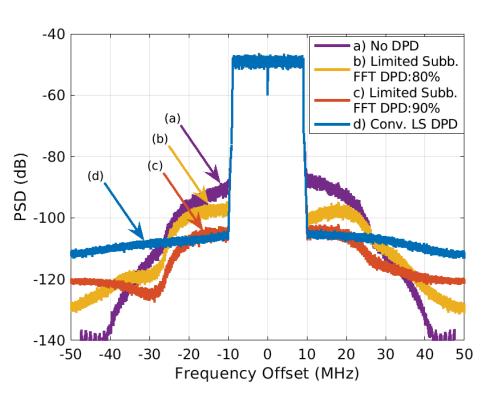
Correction performance of FFT-based Limited Subband DPD

Error between ideal PD and FFT-based Limited Subband PD





TX feedback path: FFT-based subband DPD – Linearization performance



0.6 No DPD Limited Subb. 0.5 FFT DPD:80% Limited Subb. FFT DPD:90% **Jutput magnitude** Conv. LS DPD 0.3 0.55 0.2 0.5 0.45 0.1 1.4 1.6 1.2 0.5 0 1.5 2 Input magnitude

Output spectra comparison

AM/AM comparison

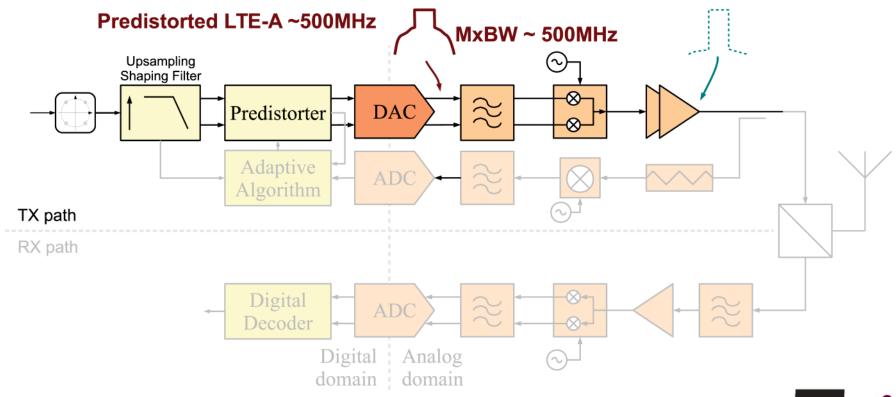




RX forward path: the bottleneck

DAC bandwidth limitation

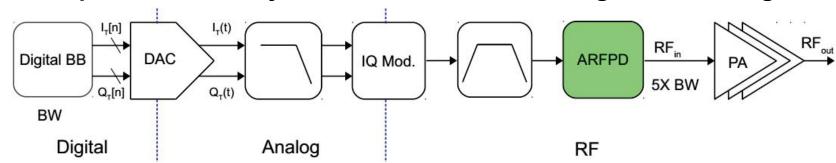
Correction phase



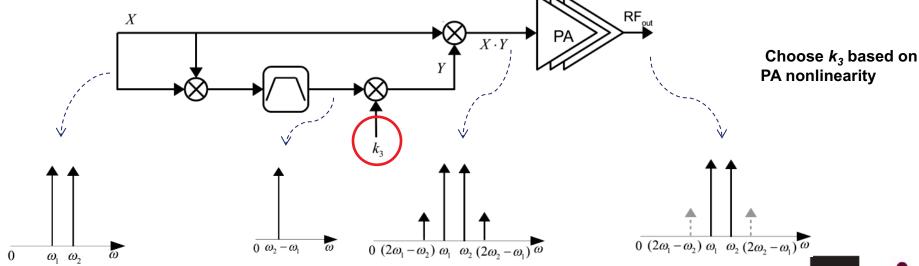


Basic Principle of Analog RF Predistortion

ARFPD performed mostly in RF, with baseband analog correction signal



- Analog multipliers are used to generate correction signal
- For a two-tone signal with memoryless PA

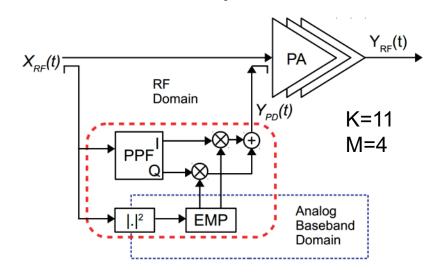




Memory-Aware ARFPD

- Roger ISSCC 2013
- CMOS IC implementation in 180nm
 - Power consumption = 200 mW
 - Max. sig. BW = 20 MHz
- EMP based predistorter

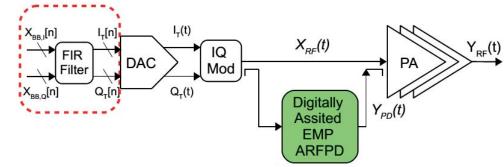
$$z_{PD,EMP}(t) = x(t) \sum_{k=0}^{K-1} \sum_{q=0}^{Q} a_{kq} \left| x(t - t_p) \right|^k$$



- Huang et al. TMTT 2015
- Measurement-instruments-based ARFPD platform
 - Max. sig. BW = 80 MHz
- FIR filter in digital BB added to EMP
 - Improves linear memory distortion correction performance

$$z_{PD,FIR-EMP}[n] = \sum_{l=0}^{L} h_l x[n-l] \times$$

$$\sum_{k=0}^{K-1} \sum_{q=0}^{Q} a_{kq} \left| \sum_{l=0}^{L} h_{l} x [n-l-q] \right|^{k}$$







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Advantages and Disadvantages of ARFPD

Advantages:

- Digital baseband clocked at normal clock rates
- Relaxed specifications of the entire Tx



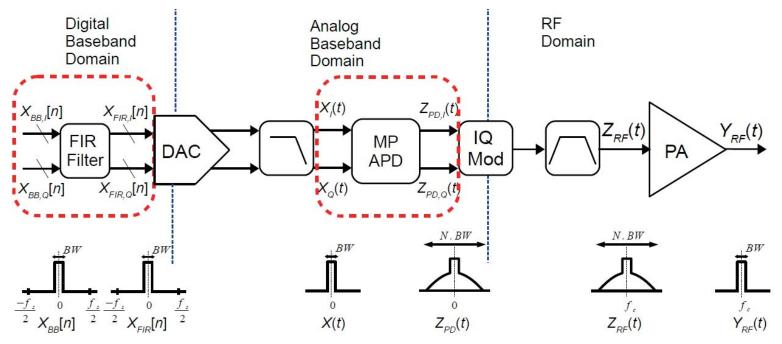
Disadvantages:

- In practice only EMP can be used → limited performance
 - MP needs Q number of RF delays and RF vector multipliers
- Analog implementation → noise, mismatch, offsets, PVT variations
- Inherent nonlinearity is caused by
 - Signal amplitude expansions and compressions, ex: $(0.1)^2 = 0.01$ and $10^2 = 100$, require high dynamic range
 - Internal bandwidth expansion, ex: correction up to IMD5 requires 1X, 3X,
 5X internal bandwidths





TX forward path: improving DPD architecture – The hybrid Mixed-Signal Predistorter (MSPD)



Advantages:

- FIR-MP MSPD provides good linearization performance
- Digital baseband needs to support just the BW
- Relaxed specifications for DACs and reconstruction filters

Disadvantages:

- Modulator and Bandpass filter still need 5X BW
- Analog implementation challenges because of non-idealities





Conclusion





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Conclusion

- Digital predistortion is a hot research topic
- Fundametal design considerations of DPD systems have been introduced
- It requires trans-disciplinary skills
 - Analog/RF
 - Data converters
 - Digital
- Many design elements are interacting
 - Multi-level approach is required
- New approaches are required for integration with disruptive technologies for 5G
 - **Massive MIMO**
 - mmWave



