

Digital Pre-distortion in GNU Radio

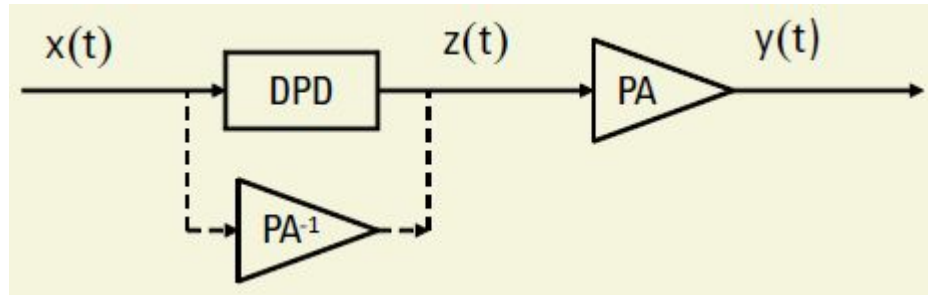
gr-dpd

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What is Digital Pre-distortion (DPD) ?

- **Digital Pre-Distortion (DPD)** is a technique to increase linearity or compensate for non-linearity in power amplifiers.
- Applies **inverse distortion**, using a pre-distorter, at the input signal of the PA to cancel the distortion generated by the power amplifier.



<https://rfmw.em.keysight.com/>

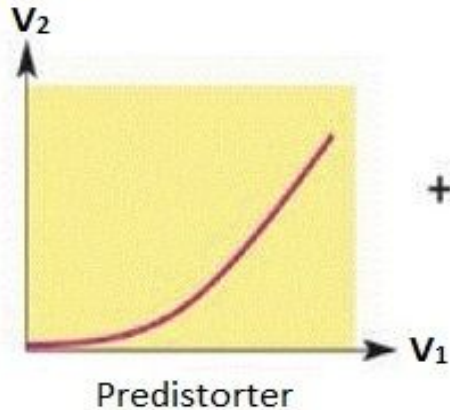
General Flow



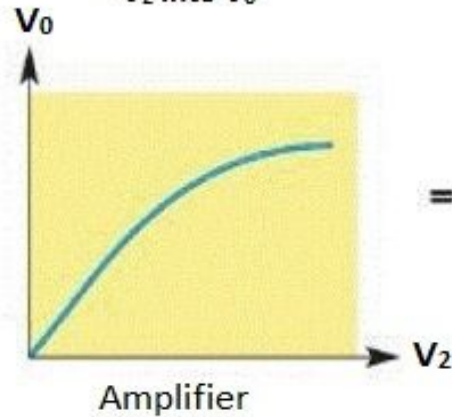
- Input V_1 is pre-distorted into V_2

- PA non-linearities distort V_2 into V_0

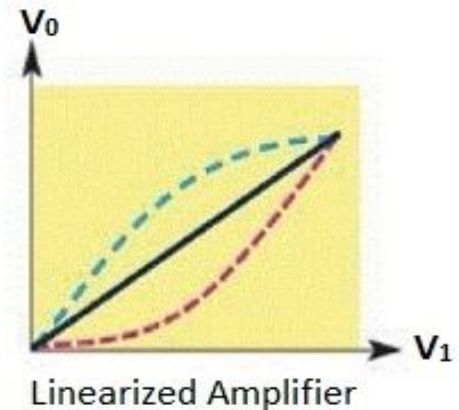
- PA output V_0 is an amplifier version of V_1



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Need for Digital Pre-distortion

- **Power amplifiers** are essential components of communication systems, but they are **inherently nonlinear**, which can cause interference, degrade bit error rate (BER) and data throughput.
- Ideally, all amplifiers should be perfectly linear but not possible due to nature of amplifier devices such as **transistors** or **vacuum tubes**.
- **DPD** is a **cost-effective** linearization technique.
- Can show improvements of up to **40%** in PA efficiency.

Power Amplifier (PA) Models in **gr-dpd**

Volterra-based models

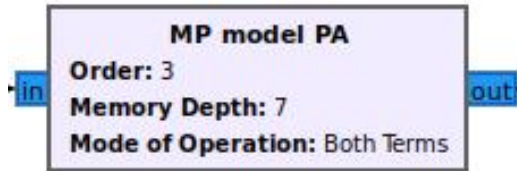
Memory Polynomial (MP) Model

- MP model based Power Amplifier produces an output based on the *Memory Polynomial* formed with current block parameters.
- An extension of the polynomial model to include memory or as a reduction of the Volterra series in which only products with the same time-shifts are included.
- Here is the function to describe relation between Input and Output of MP based model:

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

Where x is the input, y is the output, K is the maximum power order, M is the maximum memory depth, and a_{kq} is the kernels (coefficients) of the system.

MP Model PA Block



Properties: MP model PA

General | Advanced | Documentation

Order: 3

Memory Depth: 7

Mode of Operation: Both Terms

Coeff. Vector ($K_a \times L_a$): (5)

OK Cancel Apply

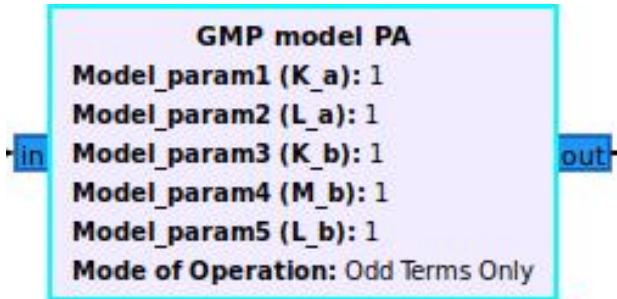
Generalised Memory Polynomial (GMP) Model

- GMP model based Power Amplifier produces an output based on the Generalised Memory Polynomial formed with current block parameters.
- Extends the MP model by also introducing products with different time-shifts, generally referred to as cross terms.
- Here is the function to describe relation between Input and Output of GMP based model:

$$z[n] = \sum_{k=0}^{K_a-1} \sum_{l=0}^{L_a-1} a_{kl} y[n-l] |y[n-l]|^k + \sum_{k=1}^{K_b} \sum_{m=1}^{M_b} \sum_{l=0}^{L_b-1} b_{kml} y[n-l] |y[n-l-m]|^k.$$

The representation on the right consists of two components, namely, the signal-and-aligned-envelope component and the signal-and-lagging-envelope component. Based on (1), the number of coefficients to be estimated are $M := K_a L_a + K_b M_b L_b$ across $\bar{M} := K_a + K_b M_b$ channels.

GMP Model PA Block



Properties: GMP model PA

General Advanced Documentation

Model_param1 (K_a) 1

Model_param2 (L_a) 1

Model_param3 (K_b) 1

Model_param4 (M_b) 1

Model_param5 (L_b) 1

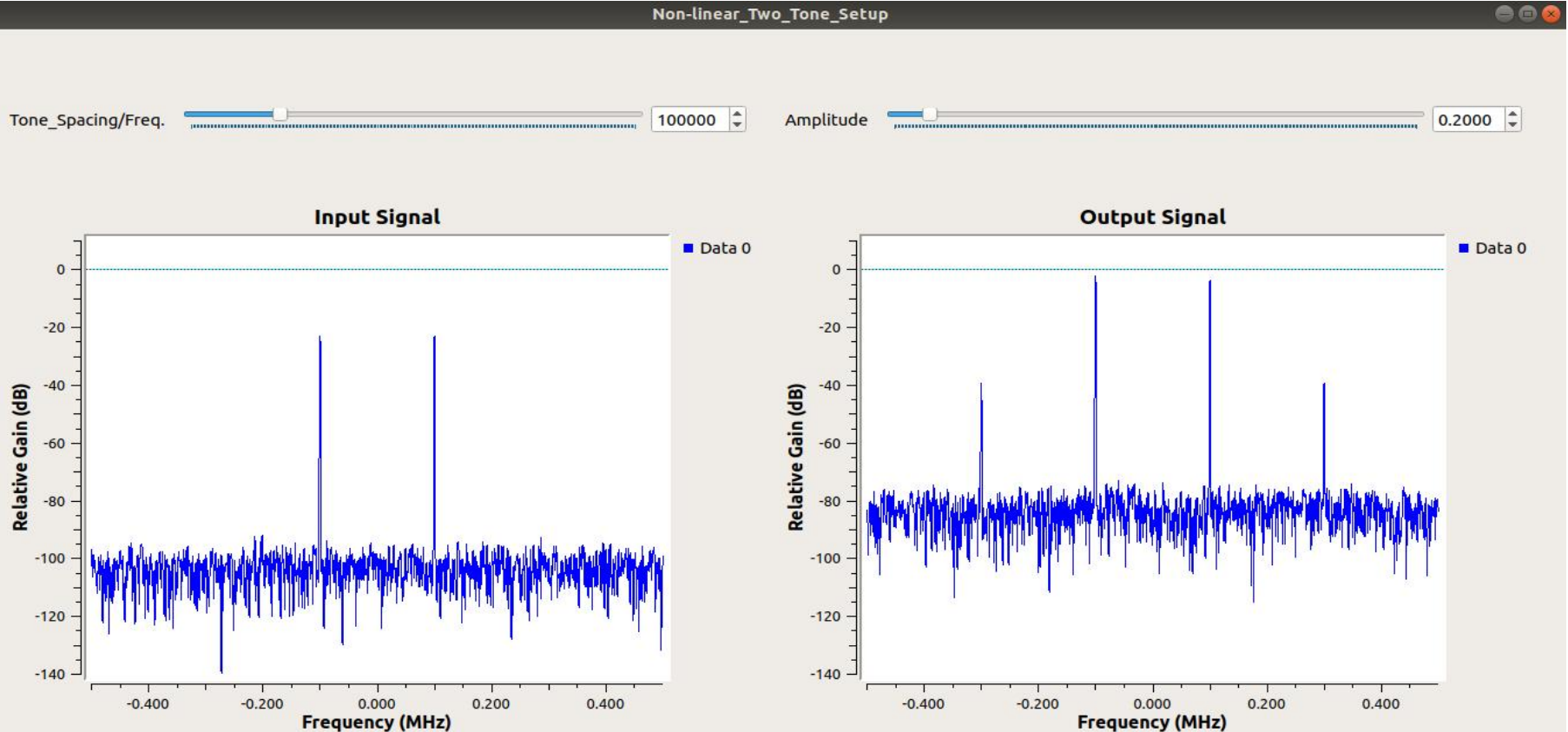
Mode of Operation Odd Terms Only

Coeff. Vector I (K_a x L_a) (10)

Coeff. Vector II (K_b x M_b x L_b) (1)

OK Cancel Apply

Example Output of a PA model in GRC



DPD Algorithms in **gr-dpd**

Recursive Least Squares (RLS) Algorithm

- *RLS postdistorter* block uses the fast RLS (Recursive Least Squares) Algorithm based on **Hyperbolics** and **Givens Rotations**.
- Major mathematical utilities to be used in the RLS Algorithm are:
 - A shift-structured **GMP vector** to store the input signal values as a **regression vector**.
 - Use of **Givens** and **Hyperbolic Givens** rotation to obtain the time updates for various parameters such as **gain vector**, **conversion factor** and **a priori output error**. [S. Pagadarai, R. Grover, S. J. Macmullan and A. M. Wyglinski]

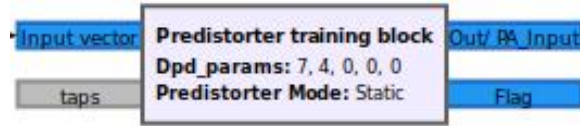
$$w_i = w_{i-1} + \left[g_i \gamma^{-1/2}(i) \right] \left[\gamma^{-1/2}(i) \right]^{-1} [z[i] - y_i w_{i-1}]$$

Weight vector or taps are updated as shown above.

Least Mean Squares (LMS) Algorithm

- *LMS Postdistorter* block uses the LMS (Least Mean Squares) Algorithm with two methods options, namely, **Newton based** and **EMA based** methods.
- Brief of the two methods used for updation of weight vector:
 - **Newton Based**: It involves calculation of error value for least squares based estimation.
 - This is weight-vector updation formula:
$$w_{iMinus1} = w_{iMinus1} + (ls_result * learning_rate)$$
 - **EMA**: This directly uses the corresponding PA input and output for least squares estimation.
 - This is weight-vector updation formula:
$$w_{iMinus1} = (w_{iMinus1} * (1 - learning_rate)) + (ls_result * learning_rate)$$

Predistorter training block



Properties: Predistorter training block

General | Advanced | Documentation

Dpd_params: (7,4,0,0,0)

Predistorter Mode: Static

Predistorter Taps: (0.749853 + 0.0721735j, 0.402292-0.0341349j, -0.220823-0.068930j)

OK Cancel Apply

Properties: Predistorter training block

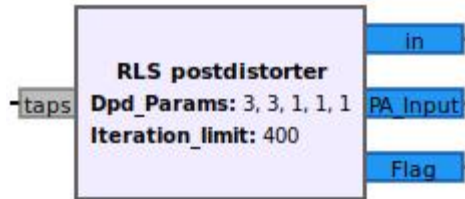
General | Advanced | Documentation

Dpd_params: (3,3,1,1,1)

Predistorter Mode: Training

OK Cancel Apply

RLS Postdistorter Block



Properties: RLS postdistorter

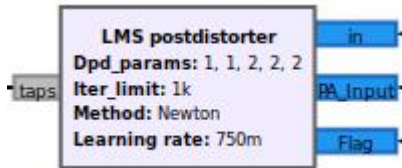
General Advanced Documentation

Dpd_Params (3, 3, 1, 1, 1)

Iteration_limit 400

OK Cancel Apply

LMS Postdistorter Block



Properties: LMS postdistorter

General Advanced Documentation

Dpd_params (1,1,2,2,2)

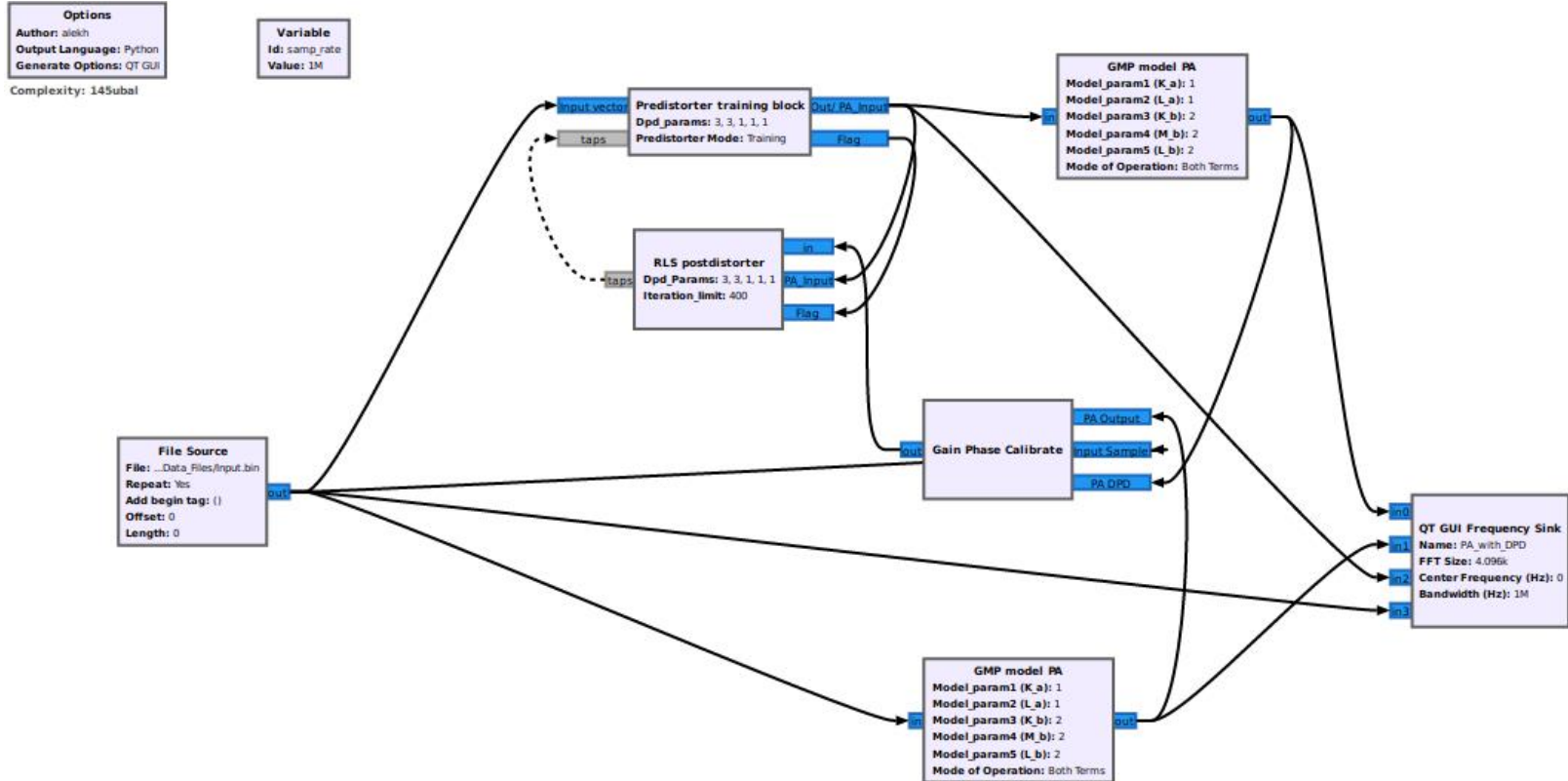
Iter_limit 1000

Method Newton

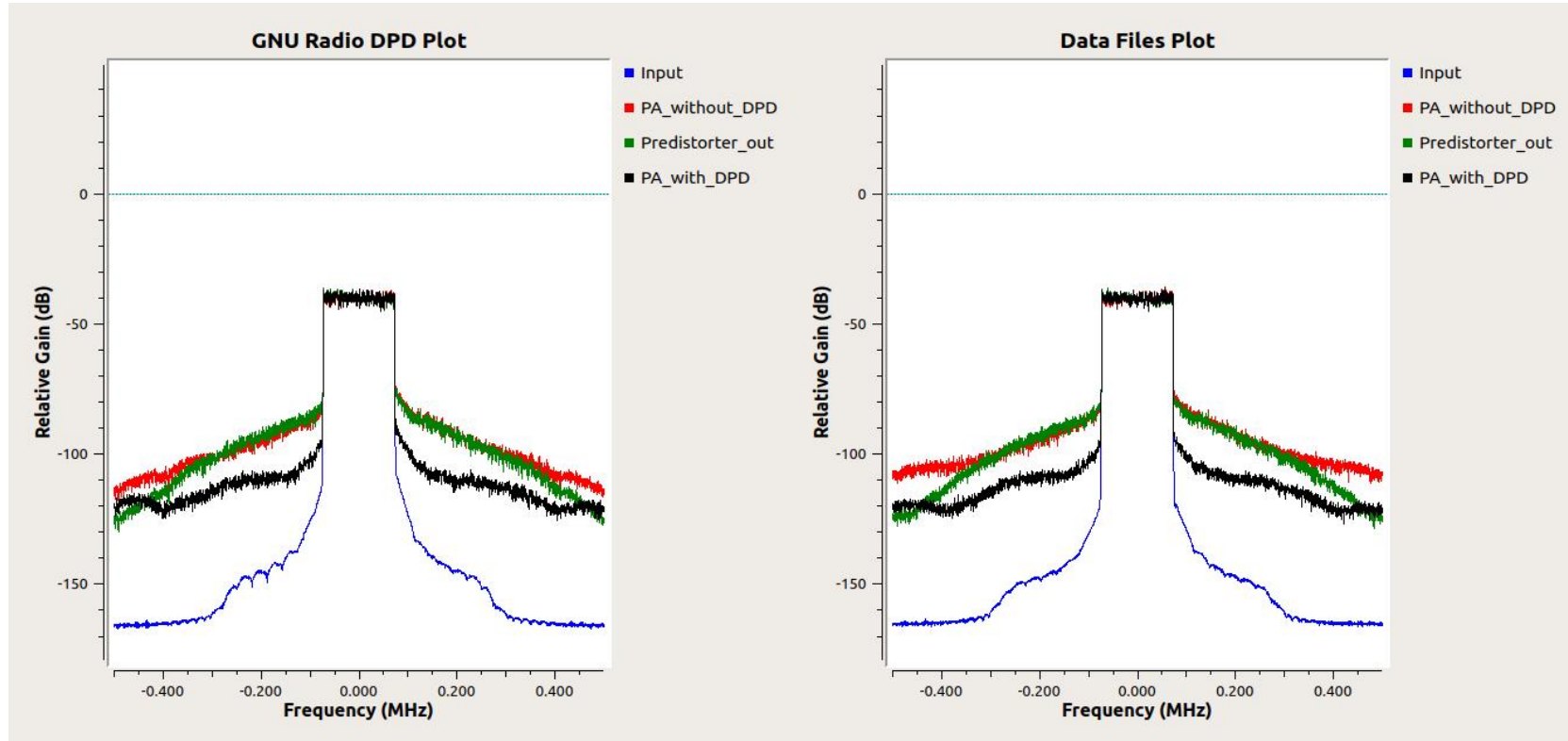
Learning rate 0.75

OK Cancel Apply

Example Flowgraph with RLS Postdistorter



DPD Output (Static Predistorter) performance comparison



Live Demo!!