

On the Design of Stable, High Performance Sigma Delta Modulators

M.A.Sc. Thesis Defence

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2. Stability and Performance
3. Similar Work
4. Optimization
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Introduction



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Ideally, the goals of the method are to:

- Model the nonlinear system accurately in a way that allows analysis of existing designs.
- Reduce dependence on simulation.
- Provide a way to design guaranteed stable sigma delta modulators in a way that minimizes conservatism.

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Noise Shaping

The use of a filter to push quantization noise out of the signal band by wrapping the quantizer in a feedback loop.



Figure 1: An example EEG signal [1] digitized to 5 bits with naïve quantization (left), 10 times oversampled quantization (middle), and first-order sigma delta modulation (right).

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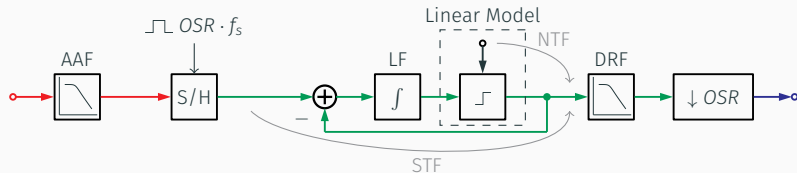


Figure 2: A simplified block diagram of a discrete-time sigma delta A/D converter.

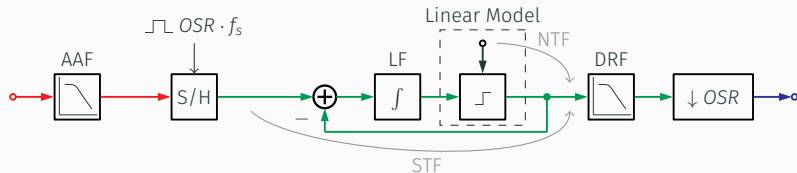


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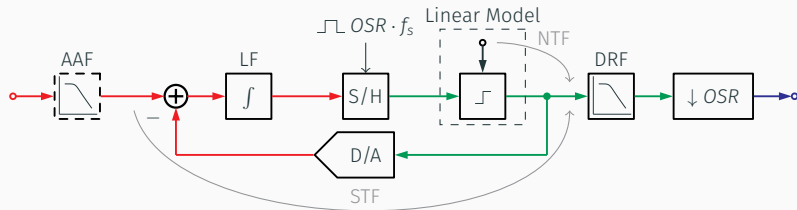


Figure 3: A simplified block diagram of a continuous-time sigma delta A/D converter.

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The design process often relies on extensive simulation to confirm that stability is likely during normal operation and the circuit may include complicated instability detection and recovery mechanisms [4, 5, 6, 7].

Stability and Performance

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- Generally conservative for 2nd order loops, approximately correct for 3rd order, and inadequate for higher order [9].
- Common in existing design tools [3, Appx. B].
- Easy to apply as a control optimization problem.
- No straightforward relationship between maximum stable input and the Lee criterion value.

Root Locus Stability Criterion

The describing function of the nonlinear quantizer is a variable gain dependent on quantizer input amplitude. A modulator loop is stable if the root locus remains in the stable region of the complex plane when sweeping through valid quantizer gains.

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- Want to maximize the quantization noise rejection in the signal band.
- Two main options:
 - Addition of weighting filters to the NTF channel, then use \mathcal{H}_{∞} control techniques.
 - Apply the Generalized KYP lemma which provides a link between a finite-frequency inequality and a set of LMIs.

Similar Work

Table 1: A comparison of some recent work on sigma delta modulator design as a control optimization problem.

Ref.	Optimized norms	NTF Type	Performance goal	Stability criteria
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This	\mathcal{H}_2, ℓ_1	IIR	GKYP	\mathcal{H}_∞ , root locus, \mathcal{H}_2, ℓ_1

¹ Only the zeros of the IIR filter are optimized.

Optimization

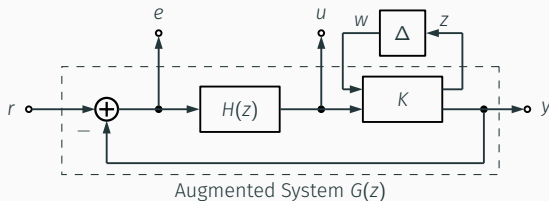


Figure 4: The augmented system model with channels of interest extracted.

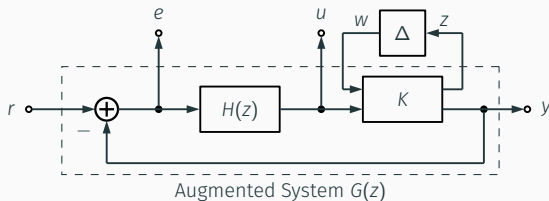


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$$G : \begin{bmatrix} \dot{x} \\ z \\ e \\ u \\ y \end{bmatrix} = \left[\begin{array}{ccc|ccc} A_H - k_{22}B_H C_H & -k_{21}B_H & B_H & & & \\ \hline & k_{12}C_H & & k_{11} & 0 & \\ & -k_{22}C_H & & -k_{21} & 1 & \\ & C_H & & 0 & 0 & \\ & k_{22}C_H & & k_{21} & 0 & \end{array} \right] \begin{bmatrix} x \\ w \\ r \end{bmatrix}$$

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- LMIs are non-convex for IIR filters because there is a product term of the pole coefficients.
- The \star -norm LMI has a non-convex scalar term.

The authors in [18] showed how to manipulate the GKYP LMI by assuming the augmented system is in CCF so that there is only one occurrence of the non-convex term in a form like:

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- Applying Shor's relaxation to linearize the problem.
- Performing an iterative method [24].

Examples

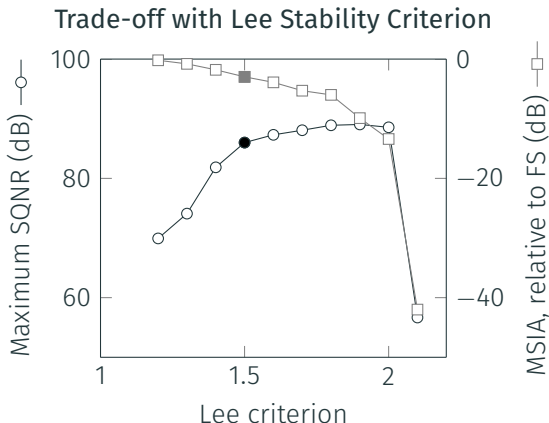


Figure 5: The performance (maximum simulated SQNR) and stability (simulated MSIA) achieved with the \mathcal{H}_∞ modulator design for different Lee criterion goals.

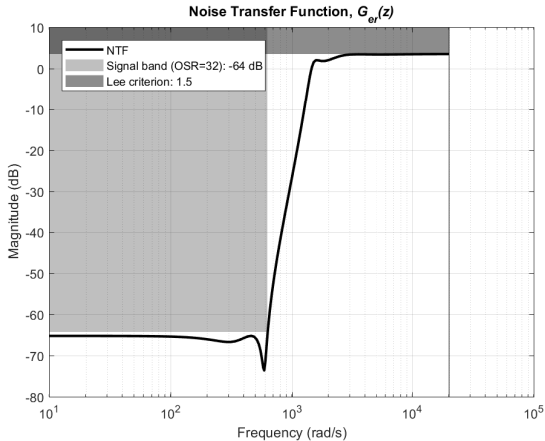


Figure 6: The noise transfer function generated with the \mathcal{H}_∞ stability criterion and associated optimization targets.

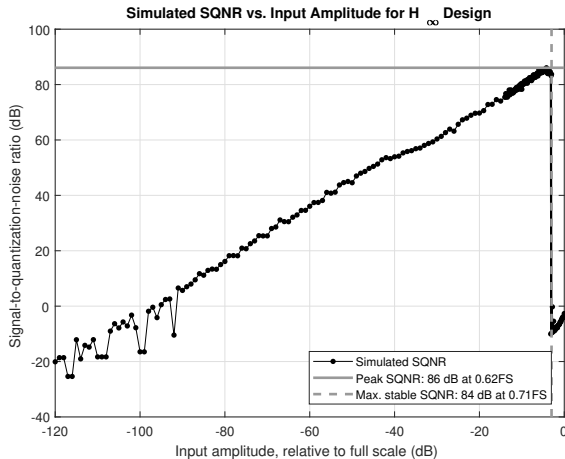


Figure 7: Simulation data for the modulator generated with the \mathcal{H}_∞ stability criterion.

Trade-off with Root Locus Stability Criterion

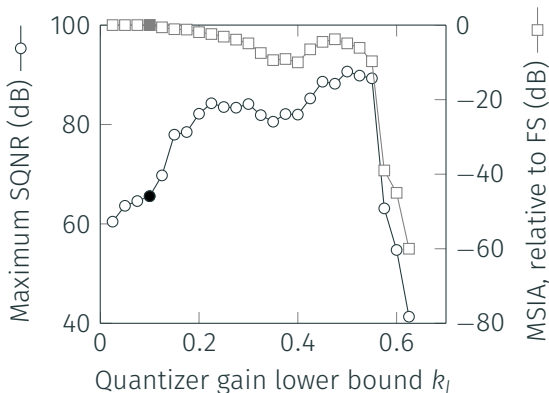


Figure 8: The performance (maximum simulated SQNR) and stability (simulated MSIA) achieved with the root locus modulator design for different quantizer gain robustness goals.

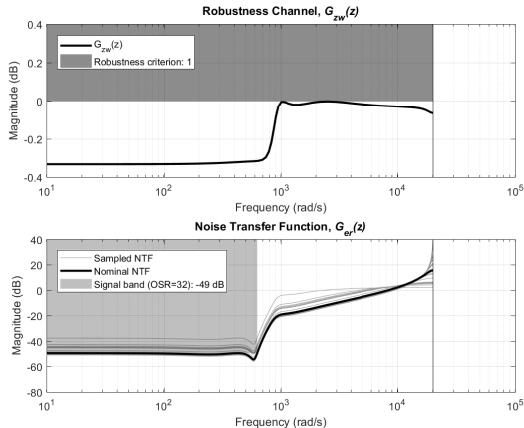


Figure 9: The noise transfer function generated with the root locus stability criterion and associated optimization targets.

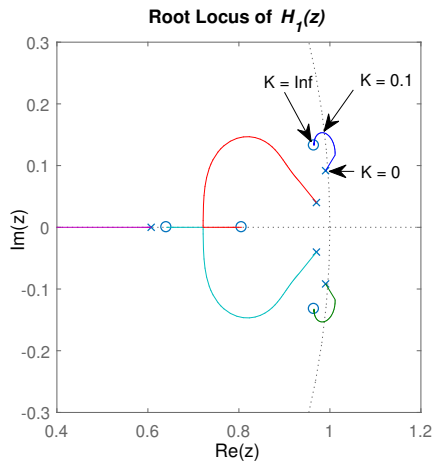


Figure 10: The root locus for the design produced when $[k_l, k_h] = [0.1, \infty)$.

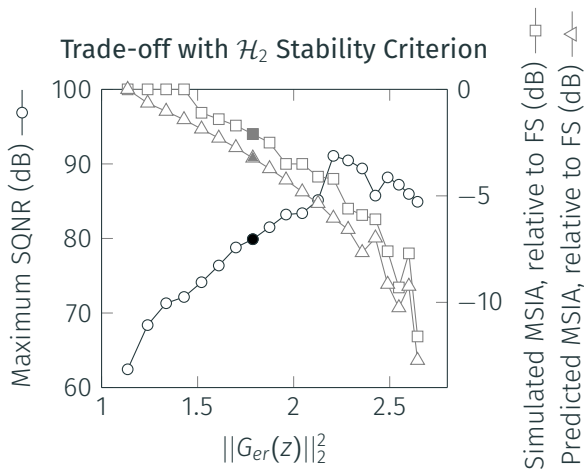


Figure 11: The performance (maximum simulated SQNR) and stability achieved with the modulator design for \mathcal{H}_2 norm goals.

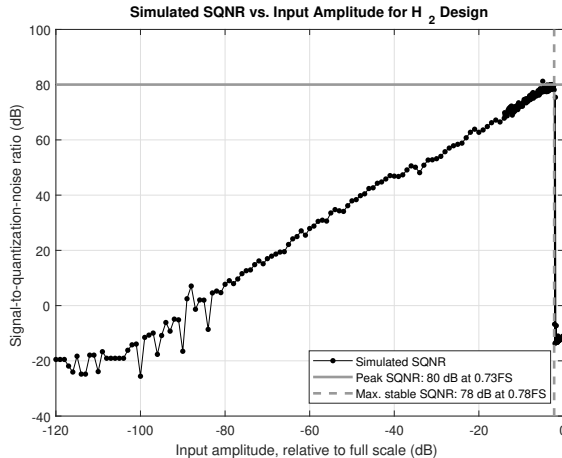


Figure 12: Simulation data for the modulator generated with the \mathcal{H}_2 stability criterion.

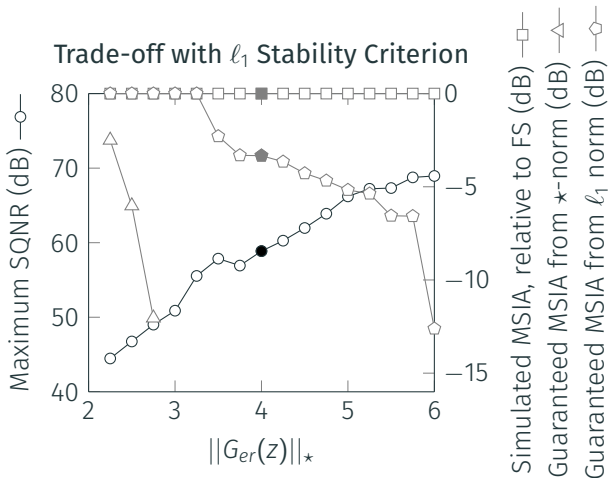


Figure 13: The performance (maximum simulated SQNR) and stability achieved with the modulator design for \star -norm goals.

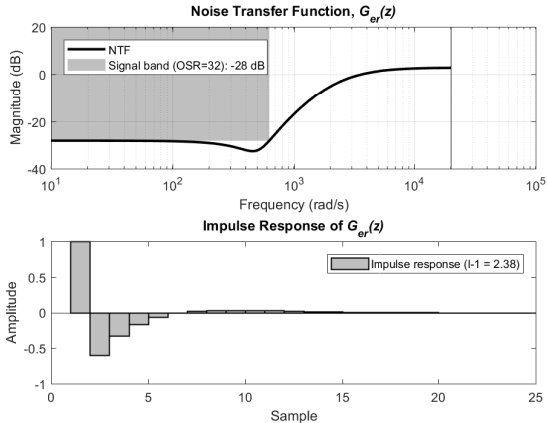


Figure 14: The noise transfer function generated with the ℓ_1 stability criterion and associated optimization targets.

Table 2: The performance and stability observations from simulations done on modulators designed using each stability criterion.

Type	Peak SQNR	MSIA (pred.)	MSIA (sim.)	LRIA
DSToolbox	87 dB	N/A	0.76 FS	-96 dB
\mathcal{H}_∞	86 dB at 0.62 FS	N/A	0.71 FS	-91 dB
Root locus	66 dB	N/A	1 FS	-52 dB
\mathcal{H}_2	78 dB at 0.73 FS	0.69 FS	0.78 FS	-84 dB
ℓ_1	59 dB	0.68 FS	1 FS	-41 dB

Conclusion

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Questions?





-  B. Blankertz, G. Dornhege, M. Krauledat, K. R. Müller, and G. Curio, “The non-invasive Berlin Brain-Computer Interface: Fast acquisition of effective performance in untrained subjects,” *NeuroImage*, vol. 37, no. 2, pp. 539–550, 2007.
-  S. Hein and A. Zakhor, “On the Stability of Sigma Delta Modulators,” *IEEE Transactions on Signal Processing*, vol. 41, no. 7, pp. 2322–2348, 1993.
-  R. Schreier and G. C. Temes, *Understanding Delta-Sigma Data Converters*, vol. 53.
Wiley, 1997.




-  N. Wong and T.-s. Ng, “Fast detection of instability in sigma-delta modulators based on unstable embedded limit cycles,” *IEEE Transactions on Circuits and Systems II*, vol. 51, no. 8, pp. 442–449, 2004.
-  N. S. Sooch, “Gain Scaling of Oversampled Analog-to-Digital Converters,” 1989.
-  S. M. Moussavi and B. H. Leung, “High-Order Single-Stage Single-Bit Oversampling A/D Converter Stabilized with Local Feedback Loops,” *IEEE Transactions on Circuits and Systems*, vol. 41, no. 1, pp. 19–25, 1994.
-  F. O. Eynde, G. M. Yin, and W. Sansen, “A CMOS Fourth-order 14b 500k-sample/s Sigma-delta ADC Converter,” 1991.

-  K. C. H. Chao, S. Nadeem, W. L. Lee, and C. G. Sodini, “A Higher Order Topology for Interpolative Modulators for Oversampling A/D Converters,” *IEEE Transactions on Circuits and Systems*, vol. 37, no. 3, pp. 309–318, 1990.
-  R. Schreier, “An Empirical Study of High-Order Single-Bit Delta-Sigma Modulators,” *IEEE Transactions on Circuits and Systems II: Analog and Digital Signal Processing*, vol. 40, no. 8, pp. 461–466, 1993.
-  C.-c. Yang, K.-d. Chen, W.-C. Wang, and T.-h. Kuo, “Transfer function design of stable high-order sigma-delta modulators with root locus inside unit circle,” in *Proceedings. IEEE Asia-Pacific Conference on ASIC*, pp. 5–8, 2002.

-  T.-H. Kuo, C.-C. Yang, K.-D. Chen, and W. C. Wang, “Design Method for High-Order Sigma Delta Modulator Stabilized by Departure Angles Designed to Keep Root Loci in Unit Circle,” *IEEE Transactions on Circuits and Systems II: Express Briefs*, vol. 53, no. 10, pp. 1083–1087, 2006.
-  K. Kang, *Simulation, and Overload and Stability Analysis of Continuous Time Sigma Delta Modulator*.
PhD thesis, University of Nevada, 2014.
-  L. Risbo, *Sigma Delta Modulators - Stability Analysis and Optimization*.
Doctor of philosophy, Technical University of Denmark, 1994.

-  D. Anastassiou, “Error Diffusion Coding for A/D Conversion,” *IEEE Transactions on Circuits and Systems*, vol. 36, no. 9, pp. 1175–1186, 1989.
-  A. Oberoi, *A Convex Optimization Approach to the Design of Multiobjective Discrete Time Systems*.
Master of science, Rochester Institute of Technology, 2004.
-  M. M. Osqui and A. Megretski, “Semidefinite Programming in Analysis and Optimization of Performance of Sigma-Delta Modulators for Low Frequencies,” in *Proceedings of the 2007 American Control Conference*, no. 6, pp. 3582–3587, 2007.
-  M. Nagahara and Y. Yamamoto, “Frequency Domain Min-Max Optimization of Noise-Shaping Delta-Sigma Modulators,” *IEEE Transactions on Signal Processing*, vol. 60, no. 6, pp. 1–12, 2012.

-  X. Li, C. Yu, and H. Gao, “Design of delta-sigma modulators via generalized Kalman-Yakubovich-Popov lemma,” *Automatica*, vol. 50, no. 10, pp. 2700–2708, 2014.
-  M. R. Tariq and S. Ohno, “Unified LMI-based design of $\Delta\Sigma$ modulators,” *EURASIP Journal on Advances in Signal Processing*, vol. 2016, no. 1, p. 29, 2016.
-  T. Iwasaki and S. Hara, “Generalized KYP Lemma: Unified Frequency Domain Inequalities with Design Applications,” *IEEE Trans. Autom. Control*, vol. 50, no. 1, pp. 41–59, 2005.
-  I. Masubuchi, A. Ohara, and N. Suda, “LMI-based controller synthesis: a unified formulation and solution,” *Robust and Nonlinear Control*, vol. 8, no. 9, pp. 669–686, 1998.

-  J. Bu and M. Sznaier, “Linear matrix inequality approach to synthesizing low-order suboptimal mixed l_1/H_∞ controllers,” *Automatica*, vol. 36, no. 7, pp. 957–963, 2000.
-  A. Oberoi and J. C. Cockburn, “A simplified LMI approach to l_1 Controller Design,” in *Proceedings of the 2005 American Control Conference*, (Portland), pp. 1788–1792, 2005.
-  S. L. Shishkin, “Optimization under non-convex Quadratic Matrix Inequality constraints with application to design of optimal sparse controller,” *IFAC-PapersOnLine*, vol. 50, no. 1, pp. 10754–10759, 2017.

Modelling Uncertain Quantizer Gain

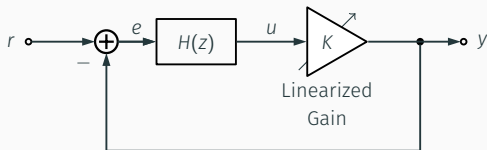


Figure 15: The sigma delta loop with the quantizer represented as an uncertain gain.

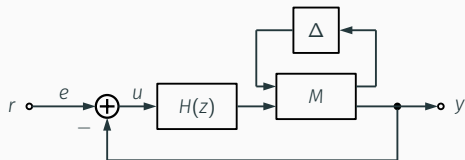


Figure 16: The linearized block diagram with the quantizer replaced by a multiplicative uncertainty extracted via LFT.

Iterative Convexification Procedure

- Find feasible initial condition a_0 .
- Separate out non-convex term:

$$\begin{bmatrix} aa^T & a \\ a^T & 0 \end{bmatrix} = \begin{bmatrix} (a_0 + a_1)(a_0 + a_1)^T - a_1 a_1^T & a_0 + a_1 \\ (a_0 + a_1)^T & 0 \end{bmatrix} = \begin{bmatrix} (a_0 a_0^T + a_0 a_1^T + a_1 a_0^T & a_0 + a_1 \\ (a_0 + a_1)^T & 0 \end{bmatrix}$$

- Using known a_0 , solve the non-convex problem in a_1 .
- Repeat process until termination criteria met.

$$a = a_0 + a_1$$

Convergence

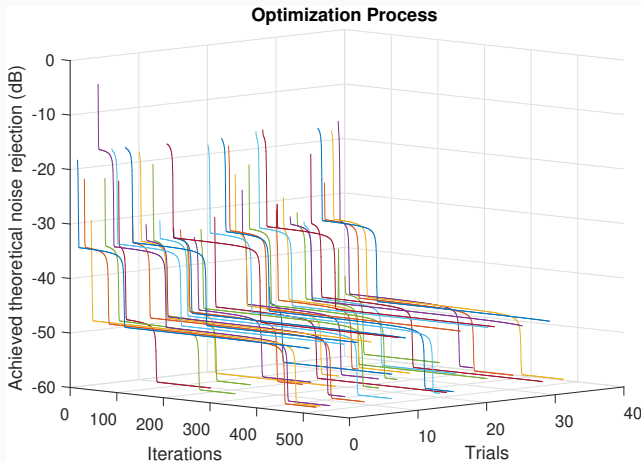


Figure 17: An example of the dependence of the iterative optimization scheme on initial conditions.

Limitations

- The influence of integrator state saturation is not captured by the model.
- LMI expressions are not feasible for loop filters with poles or zeros exactly on the unit circle.
- Convergence is dependent on choice of optimization parameters and initial conditions.
- It can be difficult to avoid pole-zero cancellations with high order designs.

Future Work

- Incorporate constraints on transfer function coefficients for ease of implementation.
- Explore implementation of a sigma delta DAC designed using this method on an FPGA.
- Find better optimization targets for continuous-time modulator design.