

Fast Black-Box Optimizers for Low Delay Audio Source Separation

Gerald Schuller

*Ilmenau University of Technology, Institute for Media
Technology, Ilmenau, Germany*

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Introduction

- ▶ Black-box optimizers have made significant progress in recent years.
- ▶ But are these optimizers applicable for real-time signal processing tasks, with a fixed time budget?
- ▶ This presentation explores the use of black-box optimizers in audio source separation.

Digital Signal Processing and Optimization

- ▶ Digital signal processing has long been linked to optimization.
- ▶ Early methods focused on filter design and adaptive filters (e.g., LMS, RLS).
- ▶ These methods are based on gradient descent, which works best with convex loss functions.

Filter Banks and Non-Convex Functions

- ▶ Later developments: Filter banks for perfect reconstruction.
- ▶ Filter banks often lead to non-convex objective functions.
- ▶ Non-convexity makes optimization more challenging.

Multichannel Blind Source Separation

- ▶ Multichannel blind source separation is another example of non-convex optimization.
- ▶ This problem is often solved in the STFT domain.
- ▶ Time-domain processing avoids delay but has higher rates of local minima.

Recurrent Neural Networks and Vanishing Gradients

- ▶ Recurrent Neural Networks (RNNs) suffer from the vanishing gradient problem.
- ▶ Black-box optimizers can help with optimizing RNNs.
- ▶ LSTMs were introduced to mitigate this problem but have higher complexity.

Applications of Black-Box Optimizers

- ▶ Speech recognition on embedded devices, adapted to the user.
- ▶ Non-differentiable loss functions in adversarial machine learning.
- ▶ Embedded reinforcement learning for small robots.

Black-Box Optimization Methods

- ▶ Black-box optimization treats the objective function as a "black box."
- ▶ Gradient information is not required.
- ▶ These optimizers search over a probability distribution of potential solutions.

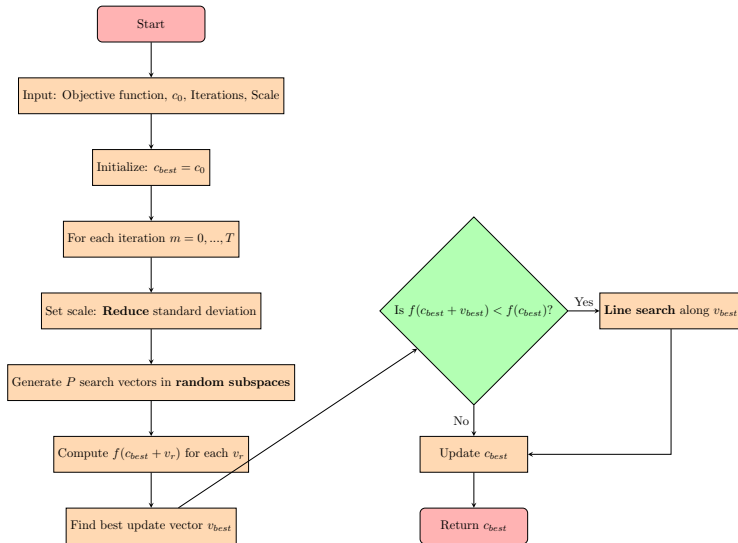
The Method of Random Directions

- ▶ The proposed method was developed for optimizing low delay filter banks and blind audio source separation.
- ▶ Uses normally distributed search vectors and shrinking standard deviation.
- ▶ Introduces random subspaces for high-dimensional problems.

Algorithm of Random Directions

- ▶ Pseudo-code for the Random Directions method:
- ▶ The algorithm guarantees that optimization does not lead to worse results.
- ▶ Random sub-space updates speed up high dimensional problems.
- ▶ Line search increases speed in smoother areas.

Algorithm of Random Directions



Random Directions Algorithm - Video Demonstration

Compared Black-Box Optimizers

- ▶ Mixture Model-based Evolution Strategy (MMES).
- ▶ Rank-One Evolution Strategy (R1ES).
- ▶ Limited-Memory Matrix Adaptation Evolution Strategy (LMMAES).
- ▶ Gradientless Descent (GLD).
- ▶ Separable Natural Evolution Strategy (SNES)
- ▶ BErnoulli Smoothing (BES)

Application 1: Blind Audio Source Separation

- ▶ Time-domain source separation using an unmixing matrix.
- ▶ Optimized using the Kullback-Leibler Divergence as the objective function.
- ▶ Parallel optimization during audio processing.
- ▶ More Information and examples in our repository
[https://github.com/TUilmenauAMS/
LowDelayMultichannelSourceSeparation_
Random-Directions_Demo](https://github.com/TUilmenauAMS/LowDelayMultichannelSourceSeparation_Random-Directions_Demo)

Results: Source Separation

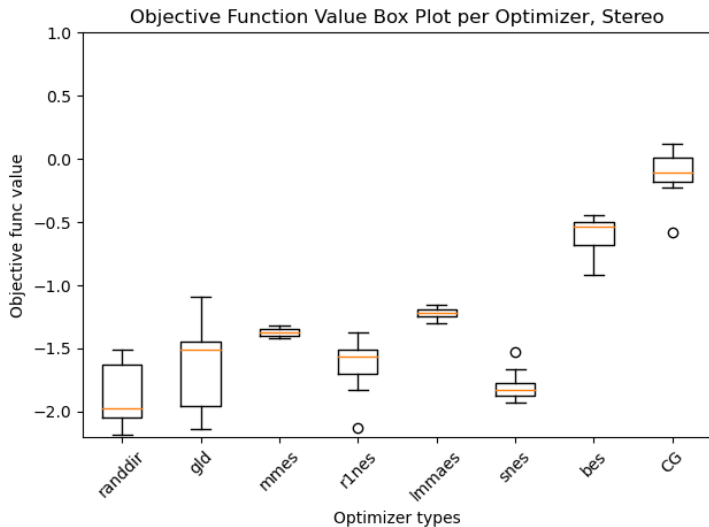
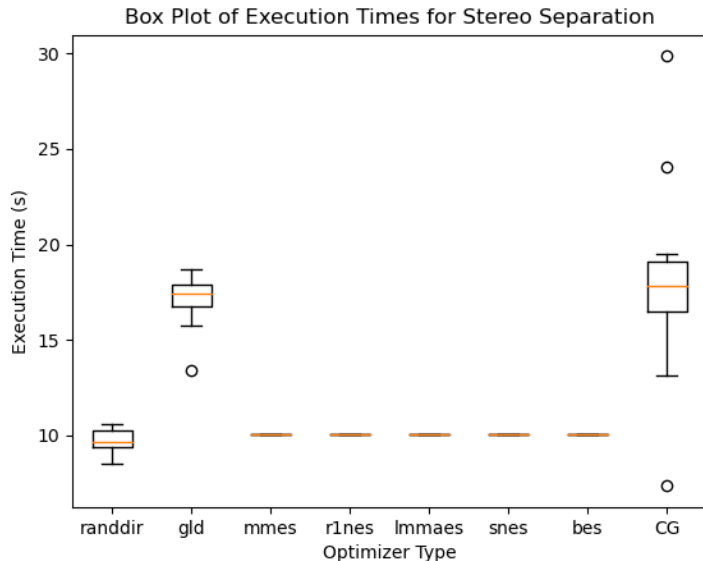


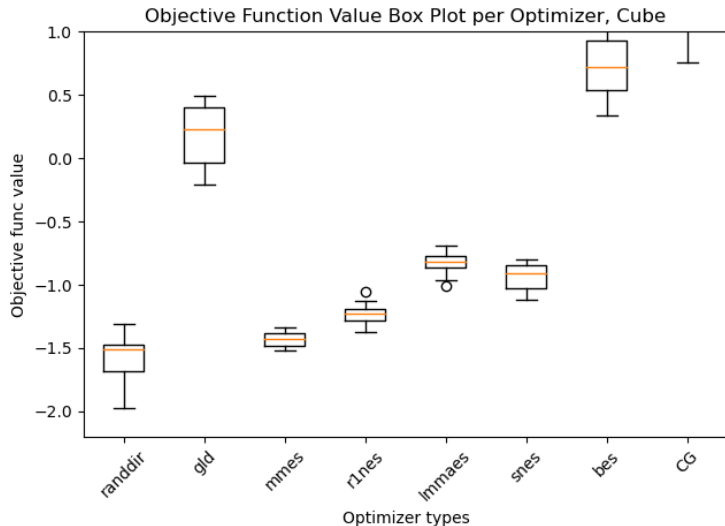
Figure: Stereo setup: Achieved minimum objective value for different optimizers.

Results: Processing Time



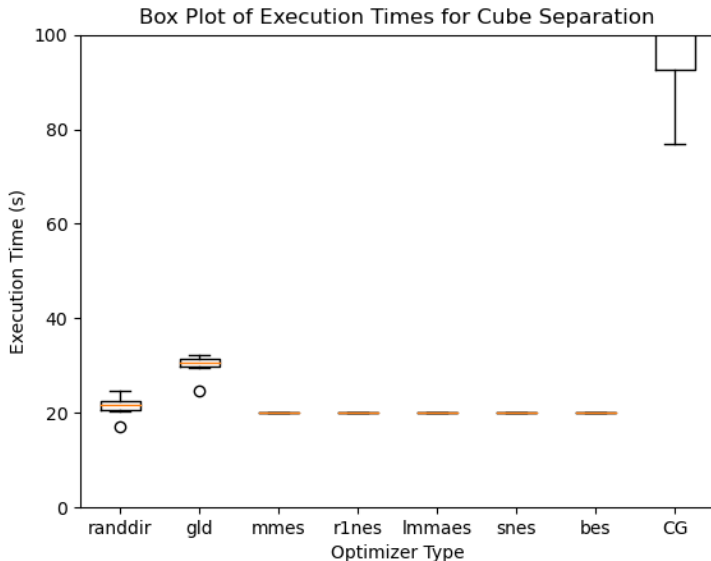
Processing times for different optimizers in stereo setup

Results: Cube Microphone Setup



Achieved minimum objective value for different optimizers.

Results: Processing Time

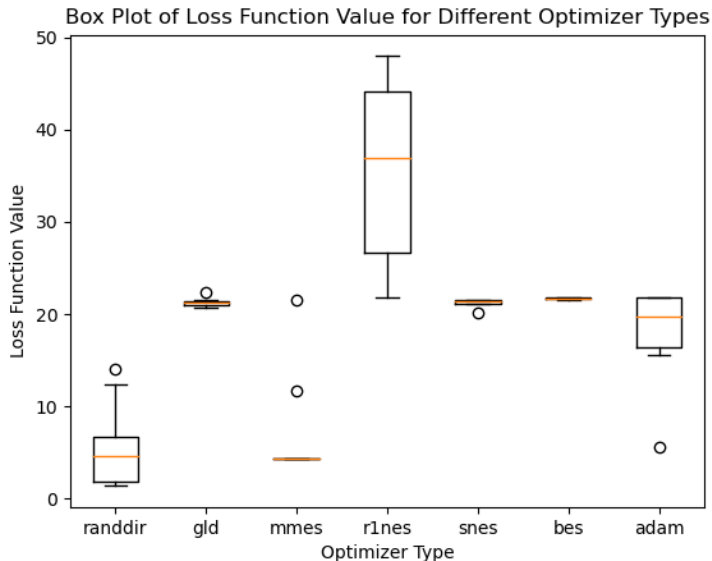


Processing times for different optimizers in cube setup.

Application 2: Recurrent Neural Network

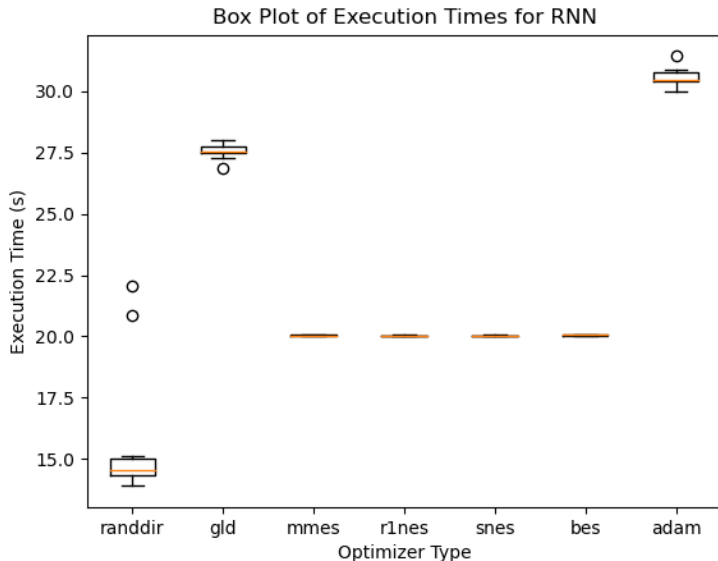
- ▶ Target signal: decaying sinusoid modeled by a 2nd order IIR filter.
- ▶ Optimized using mean squared error as the loss function.
- ▶ The optimizer must handle long-term dependencies.

Results: Recurrent Neural Network



Achieved minimum objective value for RNN optimization.

Results: Processing Time



Processing times for different optimizers in cube setup.

Conclusion

- ▶ Black-box optimization can be applied to real-time audio signal processing.
- ▶ The Random Directions optimizer consistently performed among the best.
- ▶ Future work: Apply these methods to more complex neural networks and signal processing tasks.

Thank You

Software and description available at:
[https://github.com/TUIlmenauAMS/
BlackBoxOptimizerSPcomparison](https://github.com/TUIlmenauAMS/BlackBoxOptimizerSPcomparison)



Thank you for your attention!