

# Music demixing with the sliCQ transform

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Music demixing is the task of decomposing a song into its constituent sources, which are typically isolated instruments (e.g., drums, bass, and vocals). Open-Unmix (UMX) [1] and CrossNet-Open-Unmix (X-UMX) [2] are models for music demixing that use the Short-Time Fourier Transform (STFT) to represent musical signals, but the time-frequency uncertainty principle states that the STFT of a signal cannot have maximal resolution in both time and frequency [3]. The tradeoff in time-frequency resolution can significantly affect music demixing results [4]. The STFT is computed by applying the Discrete Fourier Transform on fixed-size windows of the input signal, but for auditory and musical considerations, variable-sized windows are preferred to vary the time-frequency resolution by frequency region [5]. Our proposed adaptation of UMX and X-UMX, called xumx-sliCQ,<sup>1</sup> replaces the STFT with the sliCQT [6], an invertible transform with varying time-frequency resolution. It uses a convolutional network architecture [7] trained on the MUSDB18-HQ [8] dataset. On the test set, xumx-sliCQ achieved a median SDR of 3.6 dB versus the 4.64 dB of UMX and 5.54 dB of X-UMX, unfortunately performing worse than the original STFT-based models.

## References

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<sup>1</sup><https://github.com/sevagh/xumx-sliCQ>