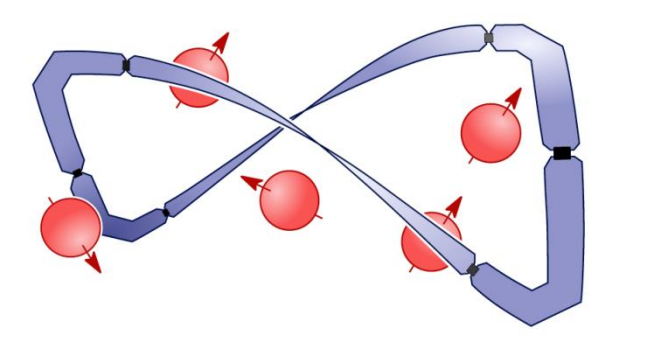


An FPGA based real-time data processing platform: Application to real-time coil data compression



Wearable MRI

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Introduction

Real-time and large-scale computing capabilities are essential prerequisites for the development of new methods in MRI. Especially, fast imaging techniques (e.g., SENSE [1], GRAPPA [2]) rely on data acquisition from a larger number of receive channels simultaneously scaling the demand for computing and data handling. Fast MRI data processing and image reconstruction has already been reported on clusters [3], cloud computers [4] and GPUs [5]. However, many of the numerically costly steps ranging from simple data compression, channel combination or phase extraction up to echo-alignment and frequency correction, could be performed efficiently and with low latency directly on the acquisition system by means of field programmable gate arrays (FPGAs).

Unfortunately, the clinical systems are typically not accessible for generic programming, neither equipped with the necessary processing power. Therefore, we developed a generic and flexibly programmable real-time processing architecture implemented on a custom spectrometer platform. The capability of it is demonstrated on the example of noise pre-whitening, channel combination and compression in order to reduce the data flow and storage requirements without the requirement of additional CPU load.

Discussion

The proposed FPGA architecture allows implementing high-throughput processing units which can perform highly parallelized and low latency computing tasks in custom configured pipelines. Frequently encountered bottlenecks such as back planes, ethernet links and mass storage drivers can be successfully avoided. Further reconstruction steps can be performed on CPUs focusing on the more complex tasks such as gridding without significantly reduced numerical burden.

Methods

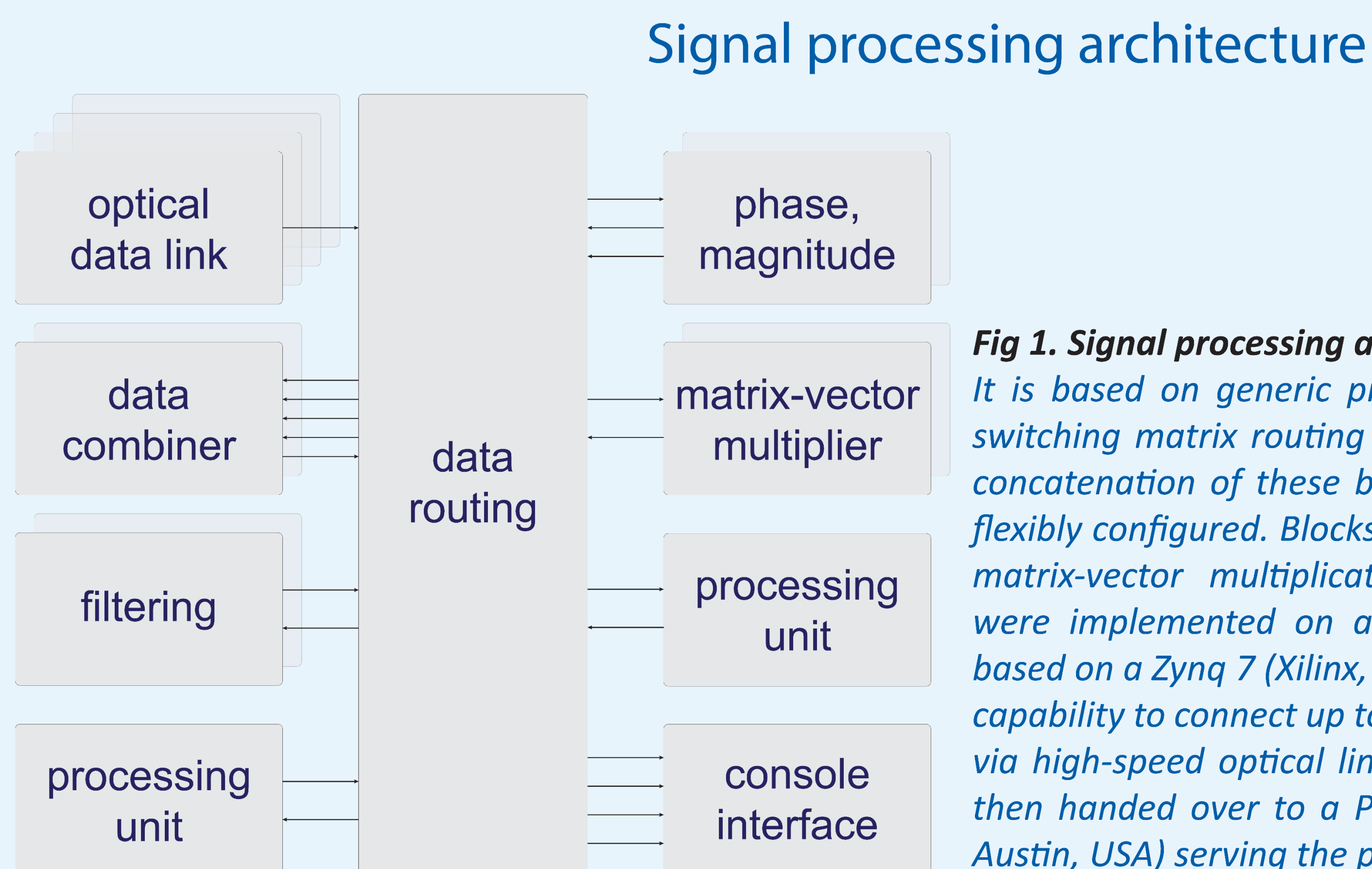


Fig 1. Signal processing architecture:

It is based on generic processing blocks connected to a switching matrix routing the respective data streams. By concatenation of these blocks the numerical task can be flexibly configured. Blocks for phase/amplitude extraction, matrix-vector multiplication and variable-rate filtering were implemented on a custom spectrometer platform based on a Zynq 7 (Xilinx, San Jose, USA) SoC [6]. It has the capability to connect up to four in-field receive modules [7] via high-speed optical lines. The resulting data stream is then handed over to a PXIe system (Nation Instruments, Austin, USA) serving the purpose of a console.

Coil compression

The data has been acquired using a 3T, 8-channel head coil (Achieva, Philips, Best, The Netherlands). A noise scan and coil sensitivity data were previously acquired and the coil combination and compression coefficients were calculated by the PCA approach [8] prior to scanning while the channel compression step was performed entirely on the FPGA while scanning.

Results

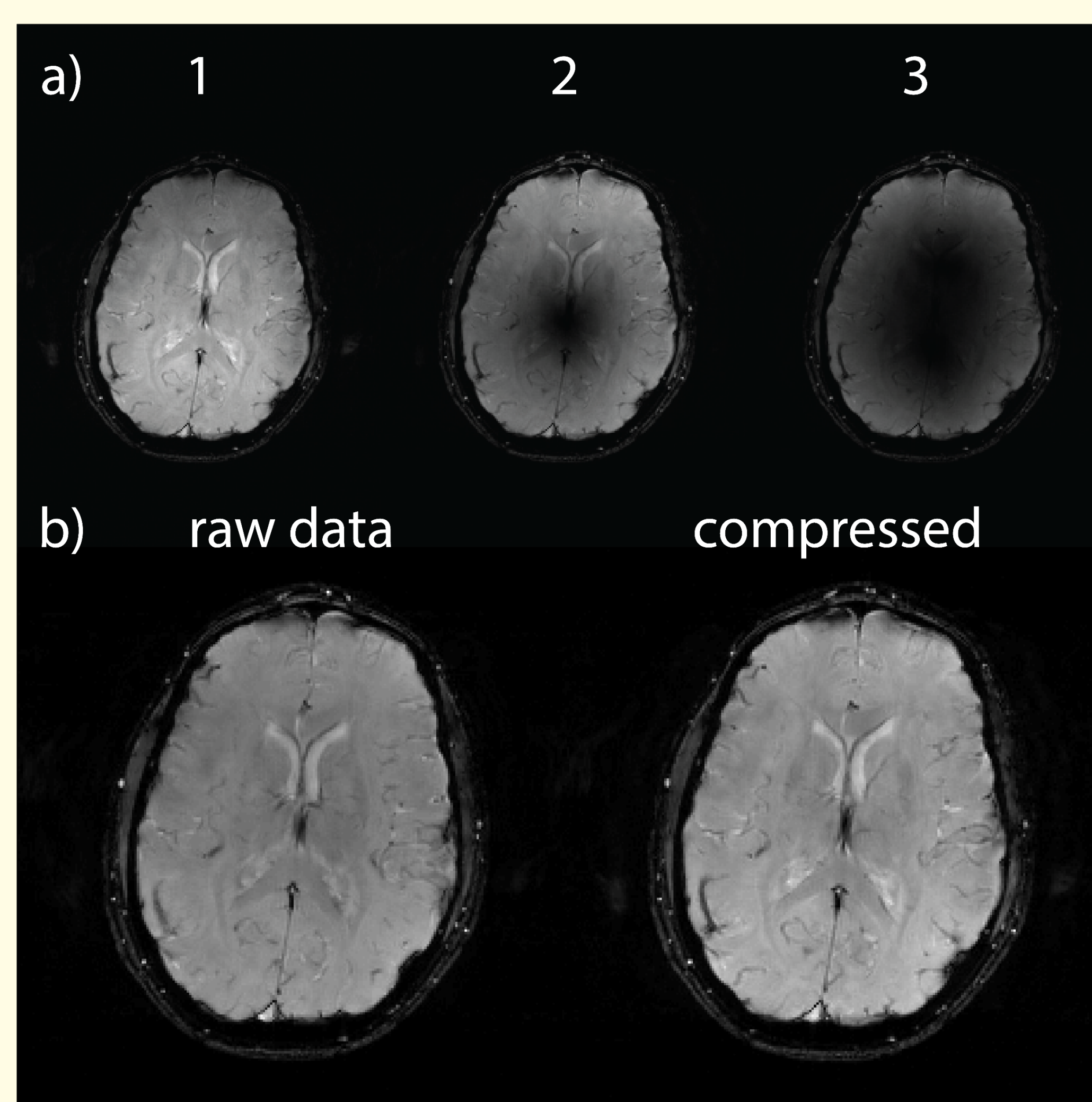


Fig 3. Coil combination experiment - in-vivo imaging: a) The first three virtual coils - principal components of a human head image acquired using a 3T 8-channel head coil. As expected the first principal virtual coil carries most of the information, while the other two carry information only on the border where the coils correlate less emphasizing the need for further principal components to faithfully reconstruct the image. b) A comparison of an image reconstructed using all 8 raw data channels and the three most significant virtual coils. As the last carry more than 97% of the information differences can't be hardly observed by naked eye.

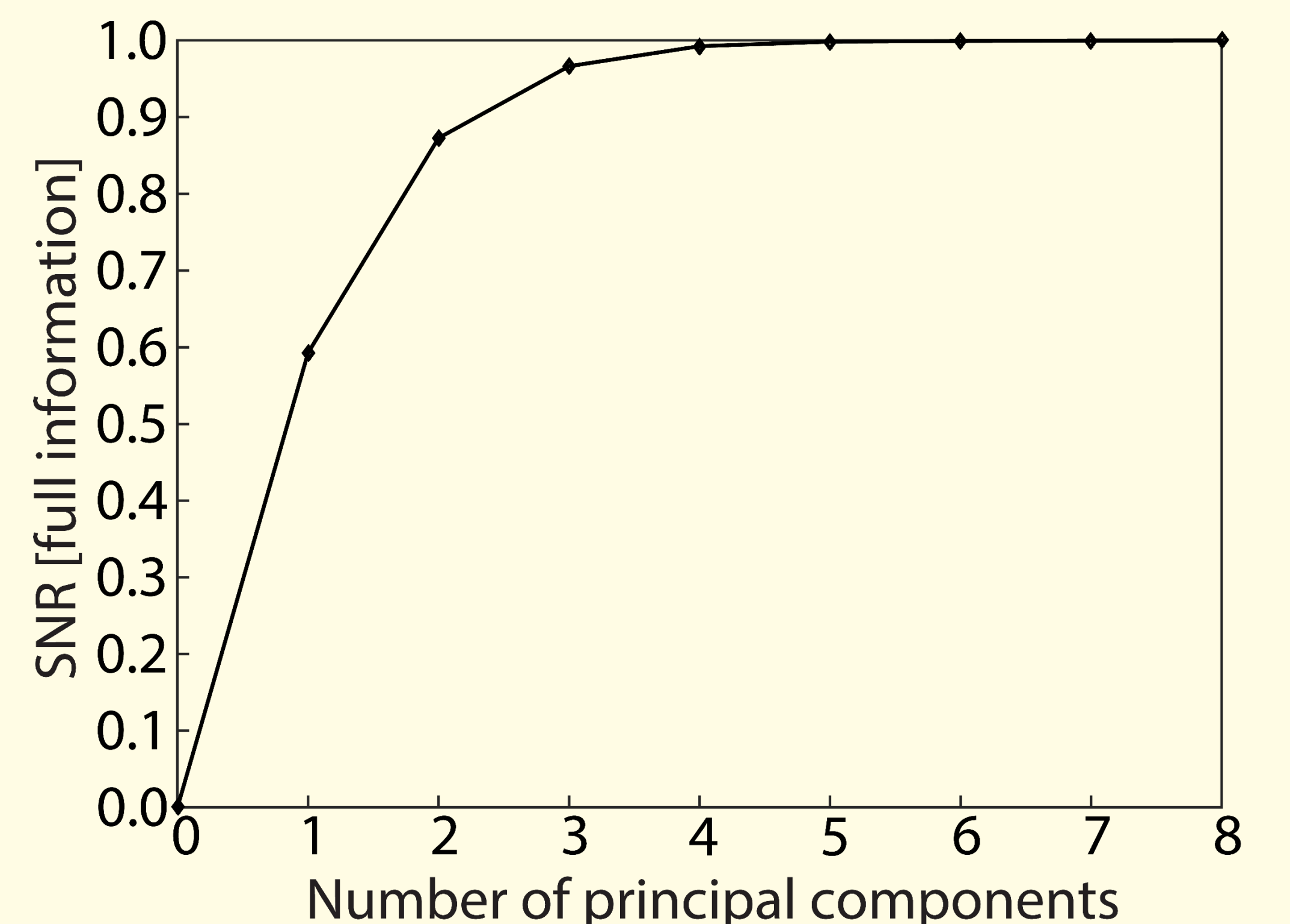


Fig 2. Coil combination experiment - Principal components: A cumulative SNR plot relative to the raw data case in dependence on the number of used eigenmodes. Due to the high correlation of the MRI coils, the first principal component of the 3T 8-channel head coil carries around 60% of the information, the second additional 30%, and the third around 7%.

References

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