

Highly-Efficient Number-Crunching-Performance SoC Macros for Singular Value Decomposition

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Opinion

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Depending from the specifications a huge design space, featuring up to thousands of possible implementations is available from the QRD-architecture template [1]. In order to support design-space exploration the parameterized cost model as well as routines for pruning (e.g., according to maximum latency), Pareto optimization etc. are implemented in a MATLAB-based optimization environment [2]. The execution time for a whole design space exploration (one set of specification parameters) is in the order of a few minutes only. Cost breakdown tables and figures can be generated automatically to get detailed insights to the cost contributions of the individual building

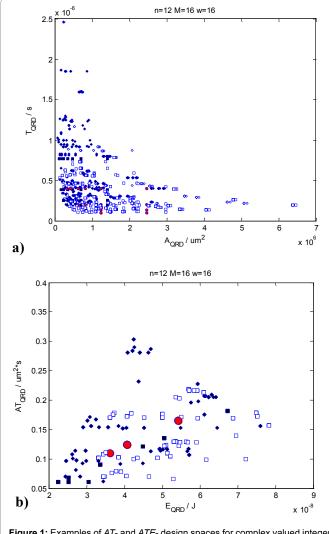


Figure 1: Examples of *AT*- and *ATE*- design spaces for complex valued integer full ([*R*] and [*Q*]) QRD of matrix size *N*=12.

blocks in order to identify bottlenecks and to guide optimization [3-5]. Constraints can be set on the maximum clock frequency e.g. to avoid unreasonably high clock frequencies due to *o*-fold Coordinate Rotate Digital Computer (CORDIC) multiplexing or on selected clock frequencies being available on a SoC [6]. Arbitrarily high throughput rates can be achieved at almost unchanged *ATE* complexity and latency by multiplexing parallel QRD blocks [1]. Therefore, especially the areaand energy-optimization for less challenging throughput specifications is a valuable capability of this optimization environment [1]. Finally, it can be applied in early cost estimation to support system conception and design.

Just to give an idea of the capabilities of the optimization environment exemplary results are in Figure 1a which shows an example of *AT*- and *ATE*-design spaces for complex valued integer full ([*R*] and [*Q*]) QRD of matrix size *N*=12, iteration count *M*=16 and word length *w*=16 [7]. The whole design space with 1,440 possible implementations is pruned for $ATE \le 10$. ATE_{min} in Figure 1b. The execution time for this example is 1 minute 42 seconds, only [8].

Figure 2 shows the *AT* as well as the *ATE* (insets) design spaces for different QRD specifications (applying extra delay stages in the PEs). The technology used is 40 nm CMOS technology. Here, for delay and output slopes SS-technology and slow-application (temperature) corner features were used while for energy and power features derived in FF-technology and fast-application corner were applied [9,10]. Even though no fabricated silicon die would feature this combination of cost figures, it is still the adequate worst case approach to ensure meeting specified figures. The back-biasing experiments were conducted assuming a reverse back-biasing voltage of V_{BS} =-0.5*V* in order to reduce leakage power in the FF-corner and assuming a forward back-biasing voltage of V_{BS} =+0.5*V*. Volts in order to reduce critical path delay in the SS-corner. Supply voltage is 0.8*V*; word length and CORDIC iteration count are specified to be *w*=*M*=16. Energy figures are given for the case that no clock or power gating is applied.

In Figure 2a-2c the variation with matrix sizes from N=14 to N=18 for a real-valued, integer-data format QRD is shown. The corresponding total numbers of possible implementations are 960, 1,200, and 1,440. The *AT*-design spaces feature hyperbola-like Pareto-optimal fronts, offering trade-offs between throughput and silicon area [11]. The *ATE*-design spaces are pruned for implementations featuring an *ATE*

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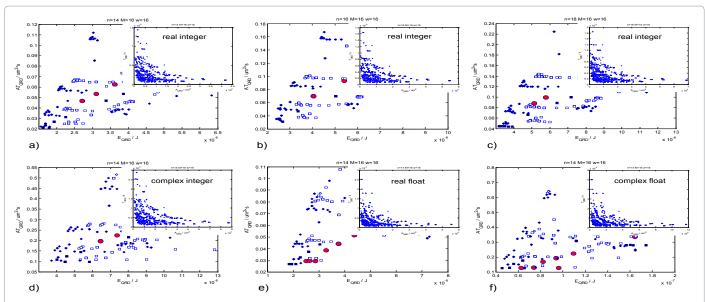


Figure 2: (a-c) AT- and ATE- design space for (a-c) real-valued integer QRD with matrix size a) N=14, b) N=16, c) N=18, and d) complex-valued integer N=14, e) real-valued float N=14, f) complex-valued float N=14; all figures for ([R] and [Q]), 40-nm CMOS worst case, integer word lengths w=16 bit, floating-point word lengths $w_{mantisse}=16$ bit, $w_{exc}=6$ bit, CORDIC iteration count M=16.

not being smaller than one tenth of the optimum *ATE*. Filled markers depict carry-ripple and unfilled markers depict carry-select adderbased implementations. Latency optimal implementations are depicted by red-filled circles. Squares mark latencies up to two, and diamonds up to five times the minimum latency. For this word length, *ATE*-optimal implementations in the lower left corner are solely carry-ripple adder based [1].

Figure 2d-2f show the results for matrix size of N=16 for complexvalued/integer - $w_{mantissa}=16$ and the exponent word length is $w_{exp}=6$. As can be seen from comparing the *ATE*-optimal implementations, the overhead for floating-point data format extensions is rather small (in the order of 26 % for *AT* and 14 % for *E*). In contrast to that, the overhead for the extensions for complex-valued matrix processing is quite high: Both, area and period costs are increased by a factor of more than two, resulting in a 4.9 times larger *AT*-complexity. Energy also is increased by more than a factor of 2.3.

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