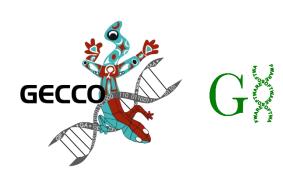
Evolutionary Approximation of Software for Embedded Systems: Median Function

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Outline



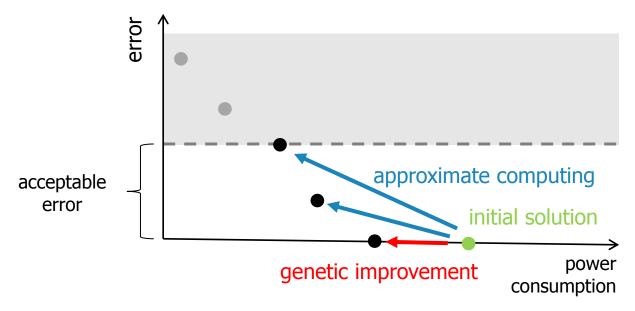
- Approximate computing
- Median function
 - properties, implementation, application in image processing
- Evolutionary approximation of median function
 - the proposed method
 - analysis of the results for real microcontrollers

Approximate computing



- Motivation: many real-world applications are error-resilient
- Principle: relaxation in accuracy can be used to simplify the complexity of computations and reduce the power consumption
- Applicability: 83% of runtime spent in computations can be approximated

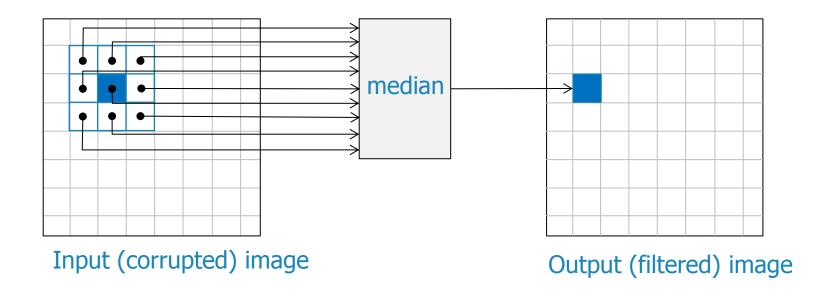
V. K. Chippa, S. T. Chakradhar, K. Roy and A. Raghunathan, *Analysis and characterization of inherent application resilience for approximate computing*, DAC 2013.



Median function



- Median: a value separating a finite sequence of data samples to two halves
- Typical application: smoothing of acquired (measured) data
- Example: noise removal in an image using a concept of sliding window



Median in image processing



corrupted image (10% pixels, impulse noise)

filtered image (9-input median filter)



Implementation of median filter



- To determine the median, we can employ:
 - a sorting algorithm
 - a selection algorithm
 - a median network

Median network

- a structure consisting of compare & swap operations
- an optimal network is known for some sizes

```
if ((a)>(b))
    PIX_SWAP((a),(b));

}

{
    PIX_SORT(p[1], p[2]); PIX_SORT(p[4], p[5]); PIX_SORT(p[7], p[8]);
    PIX_SORT(p[0], p[1]); PIX_SORT(p[3], p[4]); PIX_SORT(p[6], p[7]);
    PIX_SORT(p[1], p[2]); PIX_SORT(p[4], p[5]); PIX_SORT(p[7], p[8]);
    PIX_SORT(p[0], p[3]); PIX_SORT(p[5], p[8]); PIX_SORT(p[4], p[7]);
    PIX_SORT(p[3], p[6]); PIX_SORT(p[1], p[4]); PIX_SORT(p[2], p[5]);
    PIX_SORT(p[4], p[7]); PIX_SORT(p[4], p[2]); PIX_SORT(p[6], p[4]);
    PIX_SORT(p[4], p[2]); return(p[4]);
}

Source: http://ndevilla.free.fr/median/median.pdf
```

#define PIX_SORT(a,b) {

Implementation of median filter



- Alternatively, max and min operations can be used
 - the sequence of operations is invariant w.r.t. the input data
 - suitable for HW architectures equipped with MIN/MAX instruction
 - easier evaluation of the correctess (zero-one theorem, AND/OR)

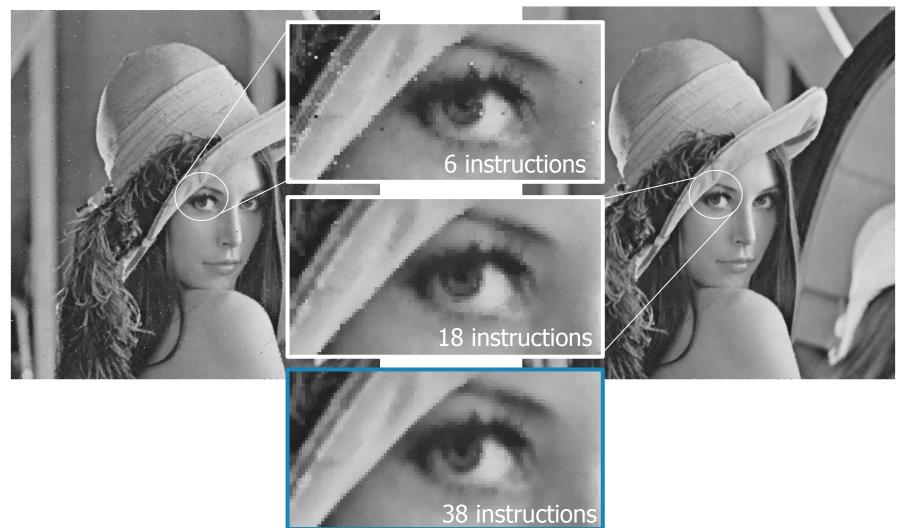
```
pixelvalue
            approx_med9 (pixelvalue * p)
{
                      s00=MIN(p[2],p[3]), s01=MAX(p[5],p[4]), s02=MAX(p[2],p[3]);
          pixelvalue
          pixelvalue
                      s03=MIN(p[4],p[5]), s04=MIN(p[0],p[1]), s05=MAX(p[7],p[6]);
          pixelvalue
                      s06=MIN(p[8],s05), s07=MAX(p[0],p[1]), s08=MAX(s04,s00)
          pixelvalue
                      s09=MAX(s08,s03)
                                         , s10=MIN(p[6],p[7]), s12=MIN(s01,s07)
          pixelvalue
                      s13=MIN(s12,s02)
                                        , s14=MAX(s06,s09) , s15=MIN(s06,s09)
          pixelvalue
                      s16=MAX(s13,s15)
                                         , s17=MAX(s10,s16)
                                                             , s18=MIN(s14,s17)
          return s18;
}
                                                         Approximate median – 18 operations
```

Approximate median filter



filtered image (9-input median filter – 6 instructions)

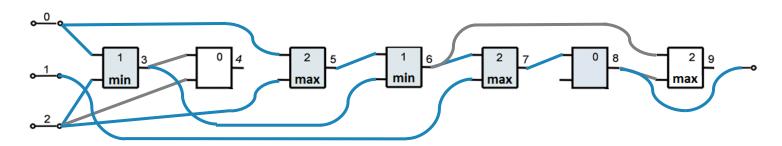
filtered image (9-input median filter – 18 instructions)



Approximate circuit design by means of CGP



- Median network (consisting of up to N operations) is represented by means of an one-dimensional array of N nodes.
- Each node can act as: identity (0), minimum (1), maximum (2)
- Each node can be connected to a node situated in the previous columns (no feedbacks are allowed).
- The configuration of nodes (the function and connection) is encoded using 3N + 1 integers.



Chromosome: $0, 2, \underline{3}; 3, 2, \underline{0}; 0, 2, \underline{2}; 5, 3, \underline{1}; 6, 1, \underline{2}; 7, 0, \underline{0}; 6, 8, \underline{2}; 8$

9

The fitness function



 The quality of approximation is measured as the sum of absolute differences between the output value of a candidate solution and reference

$$error = \sum_{i \in S} |O_{candidate}(i) - O_{reference}(i)|$$

- Scalability issue
 - |S| could be reduced from 2^{8} to 2^{n} using the zero-one principle.
 - However, it would be impossible to reasonably quantify the error (It is not important, how many invalid responses are produced).
- Solution
 - Use a randomly generated subset of S of a "reasonable" size

Z. Vasicek and L. Sekanina. *Evolutionary approach to approximate digital circuits design*. IEEE trans. on Evolutionary Computation, Vol. 19, No. 3, 2015

Evolutionary design of approximate medians



- Resource-oriented design approach is employed.
 - The evolutionary approximation exploits the idea that CGP is capable of minimizing the error even if the number of available functional nodes is not sufficient for obtaining a fully functional solution.
- Experimental setup:
 - (1+4)-ES, no crossover, 5 % of the chromosome mutated

	Median-9	Median-25	
Inputs	9	25	
Outputs	1	1	
Generations	3×10^6 (3 hours)	3×10^5 (3 hours)	
Training vectors	1×10^4	1×10^5	
Reference solution	38 operations	220 operations	
Number of nodes	6 – 34 operations	10 – 200 operations	

Quality of the evolved approximations



 The principle of construction of a median network guarantees that the output value is always one of the input values.

Consequence:

• If a sequence of 2n + 1 successive numbers R = [-n, ..., n] is used as the input, the absolute value of the highest obtained number equals to the worst-case distance from (n + 1)th lowest element

$$median_{(2n+1)}(\{-n, -n+1, ..., 0, ..., n-1, n\}) = 0$$

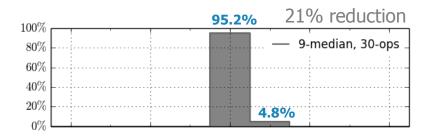
- Permutations of R can be used instead of all possible input combinations
 - 9-median: 3.62×10^5 permutations (vs. 6.27×10^{21} combinations)
 - 25-median: 1.25×10^{25} permutations (vs. 4.20×10^{60} combinations)

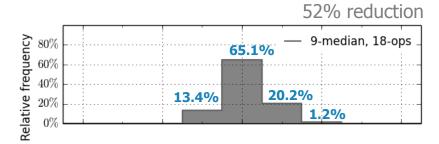
Quality of some evolved approximations

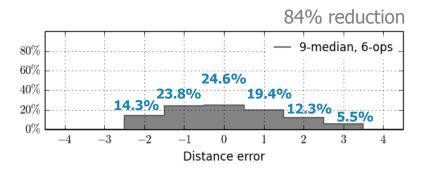


9-input median

fully-working: 38 operations

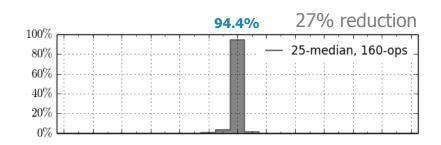


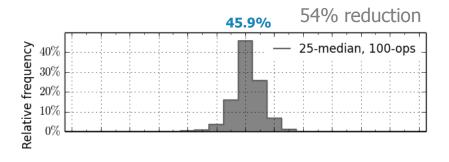


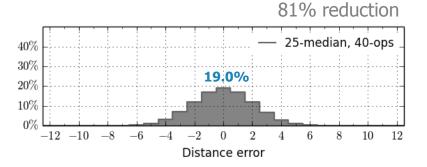


25-input median

fully-working: 220 operations







Evaluation of power consumption



Target platforms:

- Microchip PIC16F628
 - 8 bit microprocessor
 - accumulator architecture
- Microchip PIC24F08
 - 16 bit microprocessor
 - register architecture
- ST STM32F100RB
 - 32 bit microprocessor
 - ARM Cortex M3 core







Power consumption measured on real chips.

Execution time and power consumption 9-median



Impl.	Time $[\mu s]$			Ene	Energy [nWs]		
	STM32	PIC24	PIC16	STM32	PIC24	PIC16	
6-ops	2.8	54.5	170.5	86	377	342	
10-ops	3.3	70.8	251.5	102	490	504	
14-ops	3.9	86.8	336.5	118	600	674	
18-ops	4.5	104.5	424.1	138	723	850	
22-ops	5.0	116.7	487.8	151	808	978	
26-ops	5.9	130.0	558.0	179	900	1118	
30-ops	6.0	142.0	627.4	181	983	1257	
34-ops	6.4	154.0	819.7	196	1066	1643	
38-ops	6.9	165.5	885.0	210	1145	1774	
qsort	28.5	1106.2		869	7655		

34.9% error prob., max. error dist. 2 52% power reduction

4.8% error prob., max. error dist. 1 21% power reduction

fully-working median

- Quick-sort based implementation is slower and consumes significantly more energy compared to the median network.
- Due to the limited resources, quick-sort can't be even implemented on PIC16.
- 21% reduction in power consumption was achieved in the case of 30-ops median providing a negligible error

Execution time and power consumption 25-median



Impl	Tim	$e [\mu s]$	Energy	Energy [nWs]		
	STM32	PIC24	STM32	PIC24		
10-ops	3.4	71.5	104	495		
40-ops	8.1	188.5	246	1304		
70-ops	13.3	303.0	406	2097		
100-ops	17.3	401.6	528	2779		
130-ops	22.1	491.2	673	3399		
160-ops	27.4	581.4	836	4023		
170-ops	29.1	609.8	888	4220		
200-ops	34.8	698.3	1063	4832		
220-ops	39.3	755.3	1200	5227		
qsort	101.6	3067.5	3099	21227		

- 25-input median consisting of up to 220 operations offers a higher potential for power savings.
- There is nearly linear dependency between the number of operations and consumed energy (approx. 5 nW per operation for STM32).
- PIC24 requires five times more energy to accomplish the same operation.

Conclusions



- A new approach to the approximation of software routines for MCUs was presented.
- We confirmed that CGP is able to find a good trade off between error and code size even if the code size is intentionally constrained.
- A significant improvement in power consumption, code size and time of execution was achieved.
- A new method for analysis of quality of the proposed approximations was proposed.