



ESIEE Engineering
IF4-ARCH - PROJECT

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8.1.2011

Contents

1	Introduction	7
1.1	Document organisation	8
1.2	Measurement methodology	8
2	Hardware & software resources	10
2.1	Development kit & environment	10
2.2	TMS320C64x+ processor architecture	10
2.3	Memories description	11
2.4	Profiling techniques	13
3	Algorithms description & analysis	14
3.1	Deriche smoother	14
3.2	Deriche derivator	15
3.3	Thresholding & binarisation	16
3.4	Hough transform	16
3.5	Printing detected lines	17
4	Basic implementation (version I)	19
4.1	Features	19
4.2	Basic implementation on PC	19
4.3	Basic implementation on DSP	20
4.4	Implementation notes	20
4.4.1	Basic version of Deriche smoother	20
4.4.2	Roberts operator	21
4.4.3	Hough transform	21
4.5	Performance analysis	21
4.6	Conclusions	22
5	Algorithm-architecture matched version (version II)	23
5.1	Features	23
5.2	Hardware notions	23
5.3	Implementation notes	23
5.3.1	Lowering memory usage of the Deriche smoother	23
5.3.2	Smart transpositions	24
5.3.3	Optimizing memory accesses of the Deriche smoother	24
5.3.4	Optimizing convolutions of the Deriche smoother	24
5.3.5	Loop unrolling and software pipelining	25
5.4	Performance analysis	26
5.5	Conclusions	27

6	Algorithm-optimized version (version III)	28
6.1	Implementation notes	28
6.1.1	Image matrix represented in fixed point	28
6.1.2	Deriche filter towards fixed point	28
6.1.3	O’Gorman and Clowes version of Hough transform	29
6.1.4	Lookup tables instead of trigonometric functions	30
6.1.5	Using LUT tables in fixed point	31
6.1.6	Utilizing DMA	32
6.2	Performance analysis	32
6.3	Conclusions	34
7	Conclusion	35
8	Appendix A - Basic implementation on a PC	38
9	Appendix B - Basic implementation on a DSP	45
10	Appendix C - Algorithm-architecture matched version	51
11	Appendix D - Optimized version	57

List of Tables

1	PC version profiling results	19
2	Basic version profiling results	21
3	Basic version summary	22
4	Profiling results of algorithm-architecture matched version	26
5	Algorithm-architecture matched version summary	27
6	Profiling results of algorithm optimized version (without compiler help)	32
7	Profiling results of algorithm optimized version (with compiler help)	33
8	Summary of optimized version	34
9	Optimization gain	35

List of Figures

1	Image processing chain	7
2	TMX320DM6437 EVM DaVinci board	11
3	TMS320C64x+ DSP Block Diagram (taken from [4])	11
4	C64x+ Cache memory architecture	12
5	Cache data flow (see [7])	12
6	Example: (a) original image, (b) smoothed image, (c) binarized edges, (d) result	20
7	Basic version - CPU load by different operators	22
8	Algorithm-architecture matched version - CPU load by different operators	27
9	Algorithm-optimized version - CPU load by different operators	33
10	Compared FPS for different versions (on logarithmic scale)	35

List of listings

1	Time measurement approach	13
2	Deriche smoother	14
3	Deriche derivator - Roberts	16
4	Image binarisation	16
5	Standard Hough transform algorithm	17
6	Deriche smoother register rotation	24
7	Deriche derivator basic version	25
8	Deriche derivator optimised version	25
9	Using of MUST_ITERATE pragma	25
10	Converting floating point to a fixed point	29
11	Optimized Hough transform implementation	30
12	Used implementation of LUT tables in floating point	31

13	Used implementation of LUT tables in fixed point	32
14	Used implementation of LUT tables in fixed point	38
15	project.c - image processing chain implementation	39
16	testme.sh - script used to speedup development process	41
17	profile-prj.sh - script used to profile program performance	41
18	result-prof.txt - profiling results	41
19	Basic implementation on a DSP	45
20	Algorithm-architecture matched version	51
21	Optimized version	57

List of terms and abbreviations

BIOS	Basic Input-Output System
B&W	Black and White
CSL	Chip-Support Library
DSP	Digital Signal Processor
DDR2	Dual Data Rate 2
EDMA	Enhanced Direct Memory Access
EMIF	External Memory Interface
FP	Floating Point
FPS	Frames per second
GPP	General Purpose Processor
GPR	General Purpose Registers
HAL	Hardware Access Layer
JTAG	Joint Test Action Group (IEEE 1149.1)
LUT	Lookup table
MIPS	Milion instructions Per Second
MSB	Most significant bit
PP	Parallel Processing
TI	Texas Instruments

1 Introduction

Our goal is to implement and optimise a series of image processing algorithms on a DSP development kit TMS320DM6437. The result should produce real-time line detection mechanism on the image stream.

Firstly, we capture an image from the camera via the DSP and then we make it treatable by converting it into a black and white image. Second, we use on the black and white image, the Deriche Filter (optimised by Garcia-Lorca). The Deriche derivator uses the Robertson operator to get the gradients of the edges¹. Then we binarize the image by tresholding and apply a Hough transform. The Hough transform will give a result on the hough plane, where the points with the maximum intensities will represent the lines that are occurring the most. For detection of the lines we use thresholding which is an effective technique put in place to select the lines with highest votes. Once, we have extracted the relevant peaks on the hough plane then we can map them back onto the image and get the lines.

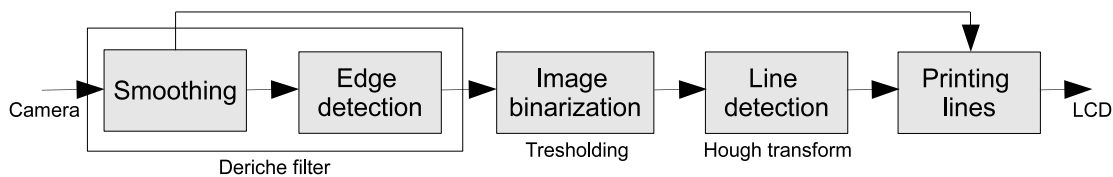


Figure 1: Image processing chain

Deriche operators have emerged for edge detection in many application areas in image processing and in computer vision. Their main drawback is their slow performance due to the large volume of calculations they require. However, their good results led to study different solutions to accelerate their performance. One of the optimised version is the Garcia Lorca approximative Deriche algorithm. The Garcia Lorca filter reduces the complex equations of the Deriche filter into two cascaded filters in causal and anti-causal senses. The causal and anticausal equations are represented and explained in the proceeding chapters. As a by-product of this implementation we get the gradients of the image in question. These gradients are kept for later use.

Binarisation is simply an application of thresholding. We select an intensity value for the pixels. Whatever is above this particular value is given the highest intensity, and whatever is below is given the lowest intensity. This renders the image comparable to a zero or one format since there are only two pixel values in question now, and hence the term binarisation. The binarisation is a necessary step because the Hough transform acts on the binary images.

Hough's transform is a method to represent all possible points on lines, as points. This is an application of the fact that a line can be uniquely identified by two parameters, which make the x and y axes of the hough space (called accumulator). If we map all the possible points (which implies all lines by extension) into the Hough space then we would obtain an ordered space where each point would basically represent a line from the original space. Once, we have the Hough accumulator we can extract and print² the lines.

¹These gradients will be used again while using the optimised version of the Hough transform

²How to print a line is explained in proceeding chapters.

1.1 Document organisation

Next chapter covers used hardware equipment, its parameters, performances and software tools used for programming, compilation, and profiling measurements. Chapter 3 contain information and analysis of used algorithms. Chapters 4 and 5 cover a basic version of our processing chain - we deal with very basic implementation of given image processing chain on a PC and on a DSP. In chapter 5 we try to apply algorithm-architecture matching techniques to optimize performance of the basic implementation on a DSP.

Finally, Chapter 6 aims to optimise used algorithms while keeping algorithm-architecture techniques. Two versions are measured, they both share the same code, but one is without compiler optimizations and one is with `-O3`³ compiler optimizations.

In the last chapter we compare all versions, make conclusions and make some important remarks.

1.2 Measurement methodology

We have measured execution time of every used operator - Deriche smoother, Deriche derivator, Hough transform and printing of the result. To get statistically relevant results, 10 measurements ($x_0...x_9$) were done for each operator and average values were used for profiling (for calculation see \bar{x} in (1)). To transform results from instruction cycles to seconds, we've noted CPU clock frequency as $f_{CPU} = 594 \cdot 10^6 Hz$. From average processing time in instruction cycles we can calculate processing time in milliseconds (equation (2)) and in percent (equation (3)).

$$\bar{x} = \frac{1}{10} \sum_{i=0}^9 x_i \quad (1)$$

$$T[ms] = \frac{1}{1000} \frac{\bar{x}}{f_{CPU}} \quad (2)$$

$$T[\%] = \frac{\bar{x}}{f_{CPU}} \cdot 100 \quad (3)$$

Frames per second (FPS) indicates processing chain performance, how many images are processed per second. We have calculated fps of whole system and fps of every and each operator. It is a purely theoretical value which indicates algorithm performance. Operations per pixel (OPPX) indicates how many clock cycles were necessary to compute value of a single pixel at the output. We have calculated it again for whole system and for every operator separately. Gain was calculated as a ratio between operator performance in previous and actual version. As in previous cases, we have measured gain between operators separately and between two versions of the whole processing chain.

$$FPS = \frac{f_{CPU}}{\bar{x}} \quad (4)$$

$$OPPX = \frac{\bar{x}}{300 \cdot 200} \quad (5)$$

³Optimization aiming processing speed and allowing increase of the program code.

$$Gain = \frac{\bar{x}_{REF}}{\bar{x}} \quad (6)$$

To calculate performances of whole system (FPS, OPPX, CPU load), we had to measure processing time outside our processing chain. We have measured total time between two frames (denoted as T_P) and time required to process implemented processing chain (T_{CH}). With T_{CH} we can calculate system FPS and OPPX. CPU load is calculated as a ratio between T_{CH} and T_P , see equation (7). This indicator shows how much CPU time in percent is used by the image processing chain itself.

$$CPUload = \frac{T_{CH}}{T_P} \cdot 100 \quad (7)$$

2 Hardware & software resources

To fulfill the task, we have created a program running on x86 GPP (PC) which allowed us to quickly design first version of the processing chain. We wrote a PC version and then merged our results together to create a version running on a DSP.

Given DSP came with BIOS, a HAL called CSL (Chip-support library) and with demo-application which sampled data from camera and sent them directly to the LCD screen. This allowed us to overstep all the allocation, loading, storing and data transferring problems. Please note that the CSL uses dedicated DMA channel to transfer between camera/LCD and external memory through one of the two EMIF's⁴ [3].

2.1 Development kit & environment

The application was tested on a development board TMX320DM6437 from TI. The manufacturer emphasizes following features [8]:

- 600-MHz C64x+ DaVinci CPU (4800 MIPS)
- One Video input via NTSC/PAL or RAW data
- One Video output via NTSC/PAL and YpbPr/RGB
- Audio I/O: S/PDIF Interface, analog, and optical
- PCI, 10/100 Ethernet MAC
- UART, CAN I/O, and VLYNQ
- 16 Mbytes of non-volatile Flash memory,
- 64 Mbytes NAND Flash, 2 Mbytes SRAM
- 128 Mbytes of DDR2 DRAM

As a development environment we have been provided with TI's Code Composer Studio (in short CCS or CCStudio) which is IDE with editor, compiler, programmer and debugger. The debugging was done via USB, although there is on-board JTAG interface, but we haven't used it.

2.2 TMS320C64x+ processor architecture

TMS320C64x+ is a 600Mhz (4800 MIPS peak) fixed-point DSP with VLIW [5] architecture and 256 bits long instruction word. Single instruction word contain 8 fields for primitive instructions and therefore capability to execute up to 8 instructions in PP. CPU consists from 32 fixed point GPR's in two datapaths (64 GPR's in total) and eight functional units, each equipped with 6 ALU's and two multipliers. Each of the 8 functional units (.M1, .L1, .D1, .S1, .M2, .L2, .D2, and .S2) is capable of executing one instruction

⁴It uses external memory interface A (EMIFA, on-chip DSP peripheral).



Figure 2: TMX320DM6437 EVM DaVinci board

every clock cycle. The .M functional units perform all multiply operations. The .S and .L units perform a set of arithmetic, logical, and branch functions. The .D units are designed to load /store data from/to memory [3]. Every .M unit can perform 1 multiplication of two 32 bit words, or 2 multiplications of 16 bit shorts, or 4 multiplications of two octets per clock cycle.

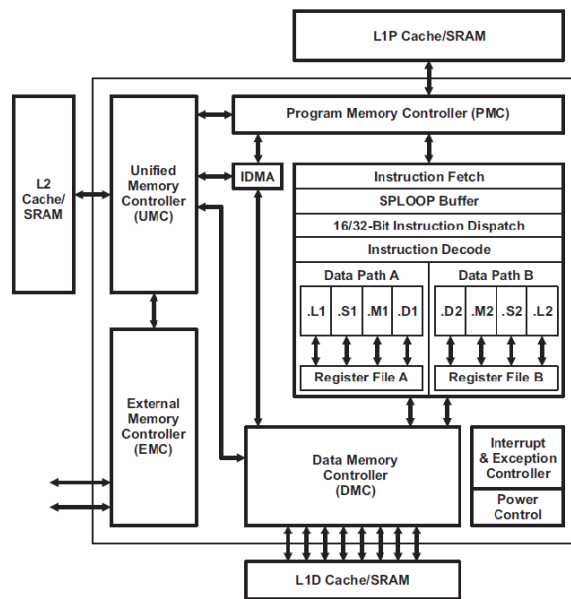


Figure 3: TMS320C64x+ DSP Block Diagram (taken from [4])

2.3 Memories description

The DSP datasheet says on the topic of memories: The C64x+ core uses a two-level cache-based architecture. The Level 1 Program memory/cache (L1P) consists of 32 KB memory space that can be configured as mapped memory or direct mapped cache. The Level 1 Data memory/cache (L1D) consists of 80 KB - 48 KB of which is mapped memory and 32 KB of which can be configured as mapped memory or 2-

way set associated cache. The Level 2 memory/cache (L2) consists of a 128 KB memory space that is shared between program and data space. L2 memory can be configured as mapped memory, cache, or a combination of both. (taken from [3]).

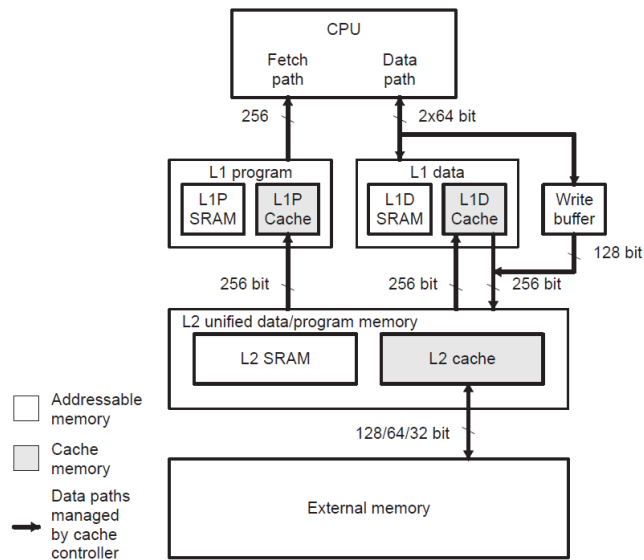


Figure 4: C64x+ Cache memory architecture

Access time to L1 cache and as well to L2 cache and RAM is 1 clock cycle. Both memories are SRAM running on the same frequency as the CPU. For further information see [6]. External memory is DDR2⁵ connected through EMIFA peripheral of the DSP. Data are transferred with EDMA peripheral support. Used DDR2 memory uses 166Mhz memory clock, so access times are at approximately 4x times higher than in case of on-chip cache memories.

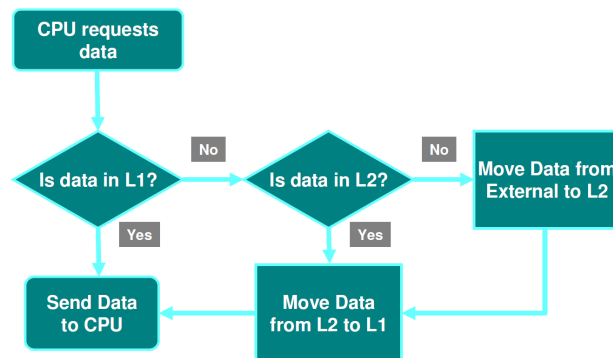


Figure 5: Cache data flow (see [7])

⁵Type DDR2-667 by JEDEC

2.4 Profiling techniques

For results analysis we have measured absolute CPU clock time between every operator. We always sample 10 consecutive frames and calculate average values per operator for analysis purposes. CSL library function `C64P_gettime()` returns actual CPU clock time as 32 bit integer.

Listing 1: Time measurement approach

```
1 unsigned int profiling[50];
2 for (i=0; i<NO_ITERATIONS; i++) {
3     /* Load the image from camera */
4     ...
5
6     /* Apply processing chain */
7     v=i%5;
8     profiling[5*v] = C64P_gettime();
9     deriche_gl(0.2);
10    profiling[5*v+1] = C64P_gettime();
11    roberts(30);
12    profiling[5*v+2] = C64P_gettime();
13    hough_transform();
14    profiling[5*v+3] = C64P_gettime();
15    print_lines(150);
16    profiling[5*v+4] = C64P_gettime();
17
18    /* Show the result on LCD screen */
19    ...
20 }
```

3 Algorithms description & analysis

In this chapter we analyze performance of used algorithms. All used operators are evaluated: deriche smoother, deriche derivator, tresholding for binarisation and a hough transform. The analysis is done for calculative operations like additions and multiplications and therefore it is not an absolute measure of algorithm running time. First of all, used architecture uses parallel processing, so number of instructions does not equal to number of cycles. Additionaly, and more importantly, calculations do not involve program flow control (loops, function calls, etc.) and memory accesses, which take most of the processing time. In practice, running time of used algorithms is by several orders higher than calculated.

3.1 Deriche smoother

Frederico Garcia Lorca version of Deriche smoother is using 2^{nd} order filter to smooth the image. We smooth horizontally and vertically in causal and anticausal way⁶. Complete algorithm analysis can be found at [9].

Listing 2: Deriche smoother

```
1 /* Horizontal smoother */
2 for(int i=0;i<height;i++){
3     /* Causal way */
4     for(int k=2;k<width;k++){
5         La[i][k] = g1*Image[i][k] + g2*La[i][k-1] - gg*La[i][k-2];
6     }
7     /* Anti-causal way */
8     for(int k=width-3;k>=0;k--){
9         Lb[i][k] = g1*La[i][k] + g2*Lb[i][k+1] - gg*Lb[i][k+2];
10    }
11 }
12
13 /* Transpose the image */
14 transpose(Lb, width, height);
15
16 /* Vertical smoother */ /* Note: it is horizontal smoother on transposed image */
17 for(int i=0;i<width;i++){
18     /* Causal way */
19     for(int k=2;k<height;k++){
20         Lc[i][k] = g1*Lb[i][k] + g2*Lc[i][k-1] - gg*Lc[i][k-2];
21     }
22     /* Anti-causal way */
23     for(int k=height-3;k>=0;k--){
24         Ld[i][k] = g1*Lc[i][k] + g2*Ld[i][k+1] - gg*Ld[i][k+2];
25     }
26 }
27 /* Transpose the image back */
28 transpose(Ld, width, height);
```

⁶In simple terms it means from left to right, from right to left, up to down and down to up.

For simplified version of horizontal smoothing implementation see listing above. Horizontal smoothing is done in a way that we go through all the image lines and smooth them in causal way (by going from left to right) and then in an anticausal way (by going from right to left). Vertical smoothing is done in the same way, only the image matrix has to be transposed. There are $320 \times 200 \times 4 = 240,000$ loop iterations in which we do 6 multiplications and 4 additions per pixel.

The horizontal smoothing accesses memory row-by-row, which provides better locality for cache memory performance than column-by-column approach manifested in vertical smoothing without transposition. The problem is the image transposition, we have to do it manually and therefore the problem with locality is not solved.

3.2 Deriche derivator

For edge magnitude and gradient detection we use Roberts operator made of convolution of two core matrices R_H and R_V (8) with the image. As a result we get two convoluted images N_H and N_V (9). Core matrices are moved by 90 degrees between each other, R_H core matrix detects edges in horizontal direction and the R_V in vertical direction. As a result, convoluted images N_H and N_V are moved by 90° to each other too. Therefore the image gradient is a by-product of the edge detection mechanism - horizontal matrix R_H provides X coordinate and vertical matrix R_V provides Y coordinate of the gradient. If we calculate magnitude (10) we get resulting intensity of the edges. If we count argument (11) we get the direction of the gradient.

$$R_H = \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix} R_V = \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \quad (8)$$

$$N(x, y) = \sum_{i=0}^1 \sum_{k=0}^1 R_X(i, j) I_{SRC}(x + i, y + j) \quad (9)$$

$$|E(x, y)| = \sqrt{N_H(x, y)^2 + N_V(x, y)^2} \quad (10)$$

$$\angle E(x, y) = \arctan \left(\frac{N_V(x, y)}{N_H(x, y)} \right) \quad (11)$$

ANSI-C implementation is listed below. We have 6 additions per pixel and $(height - 2)(width - 2)$ loop iterations which makes about 350,000 operations for image of size 300×200 pixels. Please note that our implementation of this algorithm is not creating new image matrices for both convolutions, but directly counts magnitude (euclidian distance in 2D space) and the argument (an angle).

Listing 3: Deriche derivator - Roberts

```

1 #define gpx(x,y) m[x+(y)*width]
2 float grady[width*height], gradx[width*height]
3
4 void roberts()
5 {
6     int j, k;
7     for(k=1; k<height-1;k++) /* Lines */
8         for(j=1; j<width-1;j++) { /* Columns */
9             /* ROBERTS */
10            gradx[j+k*width] = -gpx(j,k)-gpx(j+1,k)+gpx(j,k+1)+gpx(j+1,k+1);
11            grady[j+k*width] = -gpx(j,k)+gpx(j+1,k)-gpx(j,k+1)+gpx(j+1,k+1);
12        }
13 }

```

3.3 Thresholding & binarisation

This algorithm basically thresholds the image at some intensity level and says that every pixel below threshold have lowest intensity (0) and every pixel above threshold have highest intensity (255). C language implementation is listed below. We suppose that input image is resulting edge magnitude. This algorithm contain no additions or multiplications, just comparation at each pixel on the image, 1 read access and 1 write access per pixel.

Listing 4: Image binarisation

```

1 void binarize(float * abs, float treshold)
2 {
3     int j, k;
4     for(k=0; k<height;k++) /* Lines */
5         for(j=0; j<width;j++) /* Columns */
6             abs[j+k*width] = (abs[j+k*width]>treshold)?255.0:0;
7
8 }

```

3.4 Hough transform

This algorithm is supposedly most exhaustive amongst all the ones in use here, because it has algorithmic complexity of $O(n^3)$. The other algorithms are quadratic ($O(n^2)$). Hough transform takes every pixel of binarised image and looks whether it is an active pixel (white pixel). If the pixel is white the algorithm adds a vote to every possible line which goes through this pixel.

Votes are stored in the Hough accumulator space with two dimensions: ρ and ϕ . The ρ is a distance and ϕ is angle. In other words, Hough accumulator space is a space with polar coordinates. The transform

between polar and cartesian coordinates is given by:

$$\left. \begin{aligned} x &= \rho \cos(\phi) \\ y &= \rho \sin(\phi) \end{aligned} \right\} \quad (12)$$

$$\left. \begin{aligned} \rho &= \sqrt{x^2 + y^2} \\ \phi &= \operatorname{atan}\left(\frac{y}{x}\right) \end{aligned} \right\} \quad (13)$$

Therefore, Hough transform takes every white pixel and adds vote to every line with distance given by $\rho = x \cdot \cos(\phi) + y \cdot \sin(\phi)$ and with every possible angle (from 0 to 180). The algorithm in C language is listed below. Please note that $a \cdot \pi/180$ in the Listing 5 is a conversion of angle a in degrees to radians. This algorithm does two additions and two multiplications for 180 angles of each white pixel on the image. Therefore, in worst case of completely white image we get $300 \times 200 \times 180 \times 4$ operations (43,200,000 operations for 300×200 pixels image).

Listing 5: Standard Hough transform algorithm

```

1 void hough() {
2     /* run-in-vars */ int k, j, a; float rho;
3     /* maximum distance */ float maxrho = sqrt(width*width+height*height);
4     /* cleaning */ for(k=0; k<360*180;k++) h[k] = 0;
5     /* calculating accumulator space */
6     for(k=1; k<height;k++) /* Lines */
7         for(j=1; j<width;j++) /* Columns */
8             if(Image[j + k*width])
9                 for(a=0;a<180;a++) { // Angles
10                    rho = (j*cos(a*PI/180.0) + k*sin(a*PI/180.0));
11                    h[a+180*rho]++;
12                }
13 }
```

3.5 Printing detected lines

We use linear algebra to find the line. First of all we convert the line described as pair of $\{\rho, \phi\}$ into following form:

$$\begin{aligned} A_x + d_x k &= X \\ A_y + d_y k &= Y \end{aligned} \quad (14)$$

where $[A_x, A_y]$ is a known point on a line, (d_x, d_y) is a direction vector and k is multiplication constant. The known point can be retrieved simply from the hough accumulator space by transforming polar coordinates in accumulator to cartesian coordinates on the picture.

$$\begin{aligned} A_x &= \rho \cos(\phi) \\ A_y &= \rho \sin(\phi) \end{aligned} \quad (15)$$

To find a direction vector (d_x, d_y) we can take advantage of $[A_x, A_y]$. If we consider $[A_x, A_y]$ coordinates as a vector, we get a direction (A_x, A_y) , which is perpendicular to the line direction. To get the direction

vector, we just rotate the vector by 90° as it is done at (16). Resulting vector will have correct direction, but unknown size. In order to be able to draw lines pixel by pixel with k simply incremented by 1, we have to normalize the direction vector to have a size 1 (17).

$$(d_x, d_y) = (-A_y, A_x) \quad (16)$$

$$(d_x, d_y)' = \frac{(d_x, d_y)}{\|d\|} \quad (17)$$

With representation of a line in (14), we can look for an intersections with rectangle given by image boundaries . When looking for intersection with a line we are trying to solve following system of linear equations (18) to find k_1 and k_2 .

$$\begin{aligned} A_1x + d_{1x}k_1 &= A_2x + d_{2x}k_2 \\ A_1y + d_{1y}k_1 &= A_2y + d_{2y}k_2 \end{aligned} \quad (18)$$

To get intersections with a rectangle, we have to test intersection with all 4 lines on all sides. We can obtain following results: (note: we have decided to count number of intersections and print only lines with 2 intersections).

- 1 intersection when line is going through a corner
- 2 intersections between any two sides
- ∞ intersections for line overlapping with rectangle side

4 Basic implementation (version I)

This was a very first version we implemented, with no optimisation. The image processing chain was designed and tested in synthetic environment on a PC and then ported to a DSP.

4.1 Features

- All operations in floating point (IEEE 754)
- Classic version of Hough transform for lines detection
- Used trigonometric functions from C standard library
- Frederico Garcia Lorca version of Deriche filter
- Roberts operator used for Deriche derivator

4.2 Basic implementation on PC

In order to speed-up development process we decided to setup a PC environment in which we can develop our application. We have used a program for smoothing images given as a part of practicals related to a cache memory and changed it in a way in which it loads an image, applies processing chain and stores the result to a file. The program was written in C language, compiled with GCC. Profiling results are shown in Table 1. Appendix A contain source codes of this program⁷.

To be able to run tests quickly, we wrote a script which compile the source code, run the program and show the results. The script takes care about correctness of given arguments and also checks whether compilation was successful. It allowed us to make development noticeably quicker.

We have also created a script for profiling. It uses gprof profiling tool to generate performance statistics. Both scripts and profiling result are enclosed in Appendix A.

Function name	Time [%]	Total time [s]	Time/call [ms]	Calls [-]
deriche_gl	0	0.00	0.00	7
roberts	0	0	0	7
hough_lines_slow	93.75	0.15	21.43	7
print_lines	0	0	0	7
main	6.25	0.01	0.01	7

Table 1: PC version profiling results

The software was tested on a machine with Intel Core 2 Duo P8440 (2.26Ghz/3M L2 Cache) CPU under OS Debian 6.0.2. Compiler optimizations were disabled. Results clearly shows that hough transform consumes most of the computing time, compared to other operations in the image processing chain. Function main() takes about 6% of time, but that is not essential for the result.

⁷As well as profiling results and used shell scripts.

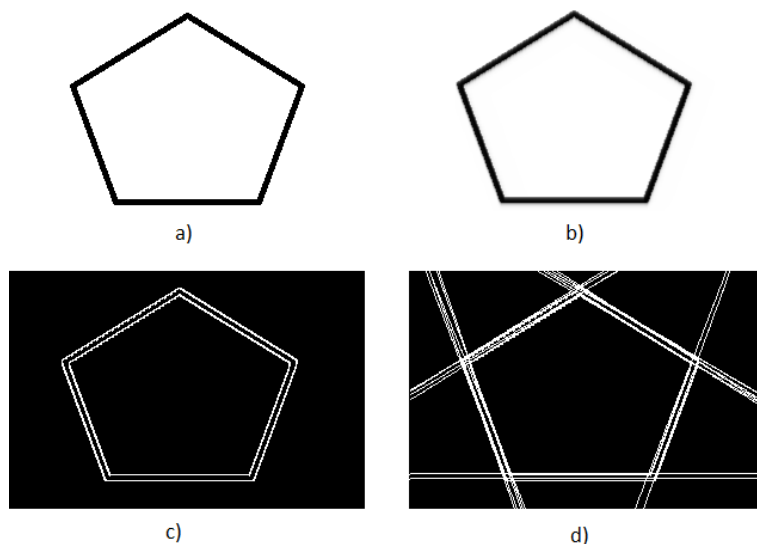


Figure 6: Example: (a) original image, (b) smoothed image, (c) binarized edges, (d) result

For the testing purposes our constants were set to 0.8 for the smoothing filter, edges were tresholded at level 160 for binarisation and all lines in hough accumulator space with more than a 125 votes were shown in the resulting image. For the result see Figure 6.

4.3 Basic implementation on DSP

The PC version was designed in a way that porting to a DSP is simple as possible, hence porting of our source code to a C64 series DSP was quite straight-forward. Program for DSP was changed for a fixed-size image (300px to 200px) and routines for loading/storing the image matrix were added. Image processing chain was placed between these two routines.

4.4 Implementation notes

We are working with image of size 300×200 in grayscale (light intensities) and using a floating point matrices (according to IEEE 754). All calculations are done in floating point, although our processor doesn't have hardware support for decimal numbers.

4.4.1 Basic version of Deriche smoother

Right from the beginning, we have run into troubles with implementing smoother of Garcia-Lorca version of a Deriche filter. Our result kept being unstable. There were several reasons:

- We misused constants gamma and alpha. Gamma is equal to minus exponent of alpha, but we took gamma directly as alpha
- There was a mistake in counting pixel values we put to the part of equation wrong operator (there should have been minus, but we put a plus)

- There were issues with storing data to correct matrices

Those mistakes accumulated overtime unknown algorithm proven to be problematic to identify. The second issue also created a problem with float overflowing to $+\infty$.

4.4.2 Roberts operator

Implementing a Roberts operator for edge detection and gradient value was quite straight-forward and unproblematic. Only drawback was that we had to be careful to do the calculations correctly. Originally we have interchanged N_X with N_Y and although that haven't changed result magnitude, the gradient direction was different. Directly after edge extraction, we did a thresholding to binarize the image.

4.4.3 Hough transform

We have prepared accumulator space made of unsigned short datatype, because longest possible line on our picture has 360 pixels, which is the maximum number of points which can vote for one line.

4.5 Performance analysis

Performance was evaluated using approach described in chapters 2.4 Profiling techniques and 1.2 Measurement methodology. We are making 10 tests and use average values to calculate processing times, frames per second (fps), operations per pixel (oppx). Memory usage and memory calls are evaluated from the source code.

Function name	time [%]	time [ms]	fps	oppx ⁸	mem. usage ⁹	mem. accesses ¹⁰
Deriche smoother	6.75	164.99	6.06	1633.47	1.44MB	12rd + 4wr
Deriche derivator	3.30	80.68	12.39	798.81	480kB	8rd + 1wr
Hough transform	89.29	2181.09	0.46	21592.79	369.6kB	1rd + 180wr
Printing result	0.13	3.19	312.63	31.66	369.6kB	-

Table 2: Basic version profiling results

- Resulting values are average calculated out of 10 measurements
- Performance result for the algorithm which prints the detected lines is highly dependent on actual number of detected lines.
- Memory accesses of printing line algorithm cannot be evaluated, because it depends on actual length of the line.
- Gains cannot be calculated, since this is a first version of our solution.

⁸Operations per pixel

⁹Memory usage refers to total amount of memory which is managed by the algorithm

¹⁰Theoretical value per pixel

From the result analysis we can see that majority of processing time is taken by the Hough transform, which slows down the system to 0.4 fps. Majority of memory is used by the Deriche filter, because it uses 4 additional image matrixes of floats for smoothing calculations. CPU is busy with the image processing chain in absolute majority of time (99.47%).

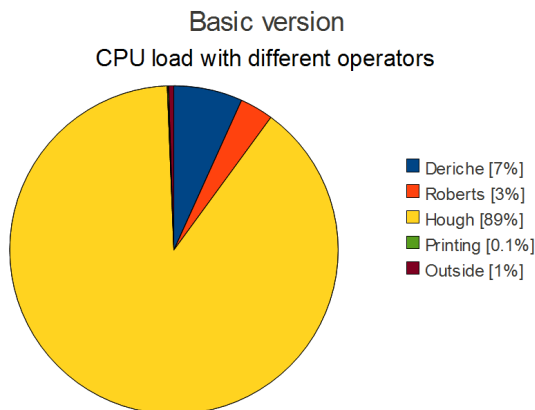


Figure 7: Basic version - CPU load by different operators

Summary of basic version shows most important indicators of this version's performance. It is resulting fps, total CPU load created by implemented image processing chain, amount of operations per pixel (oppx) and total memory usage.

Indicator	Value
FPS	0.412
OPPX	24056
CPU load	99.47%
Mem. usage	1569.6kB
Gain	-

Table 3: Basic version summary

4.6 Conclusions

Good thing is that this version works. On other hand, there are many problems. Used hough algorithm implementation takes just too much time to be used in real-life applications, big part of processing time is taken by floating point calculations and the Deriche smoother uses ridiculous amount of memory for smoothing.

There are many ways to optimize this version. Key to good performance is probably optimizing Hough algorithm itself, then move the calculations from floating point to fixed point and also we should take advantage of VLIW architecture and help compiler to be more effective.

5 Algorithm-architecture matched version (version II)

5.1 Features

- Algorithm-architecture matching techniques.
- Optimization with loop unrolling, software pipelining, and register rotations applied to the algorithms from basic version.

5.2 Hardware notions

From profiling results of the basic version we've seen that most of the processing time is taken by the Hough transform itself. This version doesn't change any algorithm, but it optimizes previously used algorithms to better utilize its hardware resources. The hardware solution uses following parameters [4]:

- Several multiplications can be done in one cycle (with hardware multiplier)
- Several additions can be done in one cycle
- Bit shift operation of arbitrary number of bits in one cycle (with barrel shifter)
- Floating point arithmetic is not supported (fixed-point CPU)
- Access to external memory is 6x slower than to internal SRAM with size of 128kB
- Thanks to parallel processing a simple operations (like incrementing a register) are executed often with no cost, together with other instructions.
- The processor has hardware support to process up to 3 nested for loops with no overhead (with SPLOOP)

DSP architecture is VLIW, which allows parallel processing of up to 8 instructions as was stated in the introduction part. Problem is that effective usage of this feature is completely dependent on the C compiler used. We can help the compiler a bit to find an optimal solution, but in the end a great deal of the resulting performance depends on the compiler, not on the programmer.

5.3 Implementation notes

Note: because in the project description was requested to move from floating point calculations to fixed point at the third version, this kind of optimisation is not used in here, although it belongs to the set of algorithm-architecture matching techniques

5.3.1 Lowering memory usage of the Deriche smoother

Previous version required 4 additional float matrices in order to compute the smoothed image. Some type of intermediate buffer is needed indeed, but 4 additional matrices makes the system slower just because of the DMA transfers between L2 cache and the DDR2 external memory.

As was described before, we have to smooth each line of the image in two directions: causal (left to right) and anticausal (right to left). When we analyze what the algorithm actually does, we find out that the intermediate buffer is required only for storing a line smoothed in one direction. So finally, we can take line of the original image, smooth it in causal way, store the result to the line buffer and then we smooth the line in anticausal way and store it back to the image.

In this way we managed to lower memory requirements from 960kB (4 float matrices 300×200 px) to 1200B (one line of floats).

5.3.2 Smart transpositions

The algorithm first does the horizontal smoothing, then transposes image matrix and finally does the horizontal smoothing again (which technically becomes vertical smoothing). When we have the result, we have to transpose image matrix for second time in order to get back original picture representation.

This optimization lies in reordering the smoothing algorithm so we need only one additional transposition. If we suppose that we get the source image matrix already transposed, we have to do the vertical smoothing first, then transpose the matrix and then we can do horizontal smoothing. The transposition is done only once. The optimization gain comes from the way in which the image matrix is loaded from the camera driver: we can load it one way or another, but if we load it directly transposed, we save time in the deriche smoother.

5.3.3 Optimizing memory accesses of the Deriche smoother

Deriche Garcia-Lorca smoother requires 12 read accesses to a memory and 4 write access in order to smooth one pixel. With the register rotation we optimized the algorithm that it need only 4 read accesses and 4 write accesses.

In order to count pixel value we need to read last three consecutive pixels. The idea behind register rotation is to store actual pixel and use it next loop iteration as previous pixel and loop after that as pixel before previous pixel. For example implementation see listing below:

Listing 6: Deriche smoother register rotation

```

1 register int p1, p2, p3;
2 p3=ig1*(mint2[ii++]); p2=(ig1*mint2[ii++]+ig2*l[0]);
3 for(k=2;k<width;k++){
4   p1 = p2; p2 = p3;
5   p3 = (g1*(mint2[ii++]) + g2*p2   gg*p1);
6   l[k] = p3;
7 }

```

5.3.4 Optimizing convolutions of the Deriche smoother

Robertson uses core matrices R_H and R_V and convolutes them with the source image. Result is used to get a gradient vector and intensity of detected edge. Original computation is shown in Listing 7 (for more

information see chapter 3.2 Deriche derivator).

Listing 7: Deriche derivator basic version

```
1 for(k=0; k<height-1;k++) // Lines
2   for(j=0; j<width-1;j++) { // Columns
3     gradx = -in[j+k*width]-in[j+1+k*width]+in[j+(k+1)*width]+in[j+1+(k+1)*width];
4     grady = -in[j+k*width]+in[j+1+k*width]-in[j+(k+1)*width]+in[j+1+(k+1)*width];
5   }
```

There are 8 read memory accesses per each convolution. But pixels on address $(j+1, k)$ and $(j+1, k+1)$ will be next loop iteration on address (j, k) , and $(j, k+1)$ respectively. The idea is to use actual pixel next loop iteration as a previous pixel without reading the memory. Altered algorithm is listed in Listing 8. It requires only 2 read memory accesses:

Listing 8: Deriche derivator optimised version

```
1 for(k=0; k<height-1;k++) // Lines
2   for(j=0; j<width-1;j++) { // Columns
3     p2 = mint[j+1+k*width]; p3 = mint[j+1+(k+1)*width];
4     convy = -p1+p4-p2+p3;
5     convx = -p1+p2-p4+p3;
6     p1 = p2; p4 = p3;
7   }
```

5.3.5 Loop unrolling and software pipelining

Loop unrolling is a technique which lowers overhead generated by loop control. Then, it copies the content of a loop several times into a single iteration. Software pipelining reduces overhead when next operation depends on finishing of current operation. Instead of waiting to finish, software pipeline prepares solution of the next iteration. This techniques are especially useful with VLIW architecture, which allows us to do up to 8 operations together during the same clock cycle.

Compiler optimizes loops automatically, it calculates optimal way of loop unrolling and applies software pipelining where possible. We can only help by describing minimum and maximum number of iterations of the loop, for this we have pragma `MUST_ITERATE(min, max, step)` [10]. For example see Listing 9 below.

Listing 9: Using of `MUST_ITERATE` pragma

```
1 #pragma MUST_ITERATE(width, width);
2 for(j=0; j<width-1;j++) { // Columns
3   p2 = mint[j+1+k*width]; p3 = mint[j+1+(k+1)*width];
4   convy = -p1+p4-p2+p3; convx = -p1+p2-p4+p3;
5   p1 = p2; p4 = p3;
6 }
```

We told the compiler additional information about loops where it was possible. Doing software pipelining to take advantage of PP in a C code is not possible, because C language was not designed to support this feature, the compiler takes care of that. In order to apply both optimisation methods, compiler optimisation `-O3` needs to be enabled. This optimisation (`-O3`) allows optimisation methods which make the program code larger in order to make the algorithm running faster.

5.4 Performance analysis

Performance was evaluated using approach described in chapter 2.4 Profiling techniques and 1.2 Measurement methodology. We made 10 tests and used average values to calculate processign times, frames per second (fps), operations per pixel (oppx). Memory usage and memory calls are evaluated from the source code.

Function name	time[%]	time[ms]	fps	oppx ¹¹	mem. usage ¹²	mem. access ¹³	gain ¹⁴
Deriche smoother	7.92	163.27	6.12	1616.40	481.2kB	4rd + 4wr	1.01
Deriche derivator	0.76	15.84	63.10	156.87	480.0kB	2rd + 1wr	5.09
Hough transform	88.58	1825.94	0.54	18076.82	369.6kB	1rd + 180wr	1.19
Printing result	0.18	3.72	268.47	36.87	369.6kB	-	0.86

Table 4: Profiling results of algorithm-architecture matched version

We can see that Hough transform keeps on being problematic, although performance was raised by 20%, the algorithm is still incredibly slow, resulting in slowing down the system to almost the same performance as in previous version.

Deriche smoother memory usage was lowered from 1.44MB to 1.2kB, which is improvement of more than 1000 times. Unfortunately speed performance of this algorithm didn't change much, only by 1%. We account this result to floating point operations, which make every loop iteration heavy and hence loop unrolling and software pipelining are not really effective.

Deriche derivator is 5.3 times faster than in previous version thanks to rotating registers. We lowered memory accesses from 8 to 2 reads per pixel and also managed to keep other 6 pixels in the GPR register bank, which made them quickly accessible.

Line printing mechanism seems to be bit slower than in the previous case, but it doesn't matter. The processing time varies greatly with the number of detected lines and therefore performance result depends upon image processing chain settings and upon image picture properties.

¹¹Operations per pixel

¹²Memory usage refers to total amount of memory which is managed by the algorithm

¹³Theoretical value per pixel

¹⁴Computed with respect to previous version

Algorithm-architecture matched version
CPU load with different operators

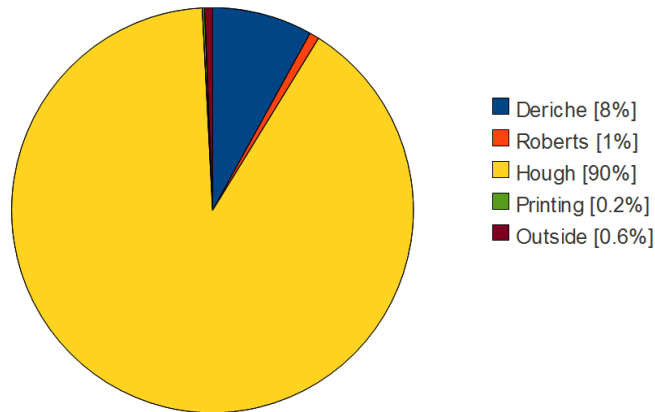


Figure 8: Algorithm-architecture matched version - CPU load by different operators

Summary of basic version shows most important indicators of this version’s performance. It is resulting fps, total CPU load created by implemented image processing chain, amount of operations per pixel (oppx) and total memory usage.

Indicator	Value
FPS	0.49
OPPX	19886
CPU load	97.45%
Mem. usage	610.8kB
Speed gain	1.21

Table 5: Algorithm-architecture matched version summary

5.5 Conclusions

This version isn’t much quicker than it’s predecessor. We have managed to take advantage of DSP architecture and raised performance by 21%. Although it is a good result indeed, yet the problem is still the Hough transform - the algorithm is too exhaustive.

We should note that the algorithm-architecture matching was not complete, because we haven’t moved computations from floating point to fixed point. This thing is asked to be done for the next version. We have big expectations, because it should allow the Deriche smoother optimizations (loop unrolling, software pipelining) to be more effective.

6 Algorithm-optimized version (version III)

In previous chapters we have seen that Hough transform algorithm is very exhaustive. We used floating point calculations on a fixed point CPU. We calculated values of $\sin(\phi)$ and $\cos(\phi)$ functions manually every call. This version optimizes all of that, changes algorithms and provides a real-time performance.

- Replace all floating point operations by fixed point
- Implement O’Gorman and Clowes version of Hough transform [2]
- Use LUT tables for trigonometric functions
- Use DMA for sending the images to the LCD screen
- Measure execution times of all used operators, profile this version and compare it to previous versions

6.1 Implementation notes

6.1.1 Image matrix represented in fixed point

When changing implementation to run completely in fixed point, first we had to change the datatype of image matrix from float (32 bit FP number) to integer format.

Our camera provides image in YUV color space [11]. Y component is color intensity called luminance, UV components are chrominances responsible for color information. Our camera encodes pixel to 16 bit number, where first 8 bit is Y component and next 2x4 bits are encoded as UV chrominance components. Since for B&W image we need only Y component, our image matrix can be encoded as a matrix of 8 bit unsigned integers.

By this conversion we have moved image matrix from floating point to fixed point, lowered memory occupation 4 times to 60kB, which allowed to move image matrix from DDR2 memory to internal SRAM (accessible part of L2 cache). Access to internal SRAM is approximately 4 times lower than external memory, SRAM is running on 594Mhz (CPU clock), while used DDR2 works on 166 Mhz.

6.1.2 Deriche filter towards fixed point

The deriche filter uses decimal constants for smoothing calculations. To make the filter working with unsigned char datatype, we had to manually convert result of floating point calculations to unsigned char. Converting resulting floats to integers worked properly with performance around 5 fps.

Thanks to this change, the Deriche filter was able to cooperate with unsigned char matrix, but inside it was still doing 12 multiplicatins and 8 summations per pixel in floating point calculated manually by processor. The problem were used floating point constants which are decimals.

The key is to multiply the constants by large number (1000 for example) and make them integer. With this approach we can store decimal number to up to three decimal places. For example number 1.2789 is FP, $1.2789 \cdot 10^3 = 1278.9$ is FP too, but after converstion to integer it becomes 1278. We can see that first three decimals are contained in the integer number. It is not a problem to do calculations all multiplied

by 10^3 , but drawback is that at some point we have to divide the result back. Division operation takes large amount of time and it is desirable to find a way around it.

It was said in chapter 2 that our CPU has a barrel shifter, so it can do bit shifts of arbitrary size in one clock cycle. It is also well known that bit shift to left by n places is the same as multiplication of the integer number by 2^n . Bit shift to right by n places divides the number by 2^n ¹⁵. In other words, we can multiply/divide in one clock cycle by 1, 2, 8, 16, 32, 64, 128, 256, 512, 1024 and so on..

In our solution we have precalculated floating point constants for smoothing and then multiplied them by 1024 (equals to left bit shift by 10 places, 2^{10} equals 1024). With the new constants we can do all the calculations in fixed point. In the end, the results have to be bit-shifted by 10 places to right in order to get proper results.

Listing 10: Converting floating point to a fixed point

```
1 float g = exp(-alpha);
2 int ig1 = ((1-g)*(1-g))*1024;
3 int ig2 = (2*g)*1024;
4 int igg = (g*g)*1024;
5 ...
6 // New fixed point calculations
7 p3 = (ig1*(matrix[ii++]) + ig2*p2 - igg*p1)>>10;
```

With this new approach the Deriche smoother manifests a performance of 234 fps, which is 40 times faster in comparison to it using the floating point calculations.

6.1.3 O’Gorman and Clowes version of Hough transform

The original version of Hough transform have $O(n^3)$ complexity. O’Gorman and Clowes proposed optimisation technique which reduces complexity to $O(n^2)$ [2]. The idea lies in approximating line direction from gradient information given by Deriche derivator. Instead of adding votes to all lines coming through a white point, we add a vote only to a line which is coming through that point with direction tangential to a gradient direction given by the Deriche derivator.

New algorithm fuses together Deriche derivator, binarisation and hough transform. For each point on the image we apply Roberts operator (posing as Deriche derivator), calculate magnitude of the edge and if it is above treshold, we add a vote to line in accumulator which is going through that point with direction tagential to the gradient.

¹⁵We consider MSB first

Listing 11: Optimized Hough transform implementation

```

1 void hough_lines(int threshold) {
2     /* Run-in-vars */ int k, j; float rho; int res, a;
3     /* gradients */ int convy, convx; float conv ;
4     register int p1,p2,p3,p4;
5
6     #pragma MUST_ITERATE(360*180, 360*180, 1);
7     /* clean the accumulator */ for(k=0; k<360*180;k++) h[k] = 0;
8     #pragma MUST_ITERATE(height, height);
9     for(k=0; k<height-1;k++) { // Lines
10        /* pre-load data to registers */ p1 = matrix[k*width]; p4 = matrix[(k+1)*width];
11        #pragma MUST_ITERATE(width, width);
12        for(j=0; j<width-1;j++) { // Columns
13            /* load next column */ p2 = mint[j+1+k*width]; p3 = mint[j+1+(k+1)*width];
14            /* calculate gradients */ convy = -p1+p4-p2+p3; convx = -p1+p2-p4+p3;
15            /* rotate registers */ p1 = p2; p4 = p3;
16            /* binarisation */
17            if((abs(convx) + abs(convy)) >= threshold) {
18                /* calculate grad angle */ a = (int)atanf(convy/(float)convx);
19                /* calculate point dist */ rho = j*cos(a) + k*sin(a);
20                /* store to acc */ h[a+90+180*((int)rho+360)>>1]++;
21            }
22        }
23    }

```

Please note that we have used accumulator with ϕ from -90° to 90° with coordinates encoded as $\phi = \text{array_index} - 90$. For example angle 0° degrees equal to column 90 in the accumulator space. This is a new feature which was used because of the `atan()` function. Arcus tangens is defined on interval $(-\infty, +\infty)$ and output range is $(\frac{-\pi}{2}; \frac{\pi}{2})$ which is equivalent to $(-90^\circ, +90^\circ)$

6.1.4 Lookup tables instead of trigonometric functions

Trigonometric function `sin()`, `cos()` and `arctan()` from C standard library are calculating functions using a Taylor series. Taylor series is generally complicated equation which requires time to be processed. With calculative approach to trigonometric functions we can calculate result very precisely, disadvantage is time required for the calculation. Aim of this optimisation technique is to pre-calculate resulting values for appropriate argument interval and store the result to a lookup table (LUT).

Lookup table is an array which acts as a mapping of $\mathbb{N}^+ \mapsto \mathbb{R}$, where \mathbb{N}^+ is array index and \mathbb{R} is the trigonometric function result. To get the \mathbb{N}^+ array index we need to do mapping of function input to array index. For that we have to choose how many function values we want to have on how long input interval.

We are using Hough accumulator space with angle precision of 1° (in other words $\phi \in \mathbb{Z}$ at interval $[-90^\circ, 90^\circ]$). This allows detected line angle precision of 1.7% at image size of 300×200 pixels. For full calculations see explanation below - diagonal (longest) line have offset 6 pixels between it's ends with

maximum error of 1° . That makes 1.7% of the line length.

$$\sqrt{300^2 + 200^2} \sin(1) = 360 \sin(1) = 6px \quad (19)$$

$$\frac{6px}{360px} = 1.7\%$$

This precision is considered as good enough for both Hough transform and LUT tables. Used lookup table have to contain angles at interval $[-90^\circ, 90^\circ]$ for $\sin()/\cos()$ and $(-\infty, +\infty)$ for $\arctan()$. At this point we can use advantage of $\sin()$ properties: it is an odd function. It means that $\sin(-\phi) = -\sin(\phi)$. Similarly $\cos()$ is an even function and therefore $\cos(-\phi) = \cos(\phi)$. With this knowledge we can redefine argument interval for both $\sin()$ and $\cos()$ as $[0^\circ, 90^\circ]$ and only setup rules to set sign of the output accordingly to the input. In case of $\sin()$ function the output has a same sign as an input. In case of $\cos()$ function the output doesn't change no matter positive or negative input. For used LUT implementation example of $\sin()$ and $\cos()$ see Listing 12.

Listing 12: Used implementation of LUT tables in floating point

```

1 float cosT[91]; float sinT[91];
2 float test; int a;
3
4 /* Initialize LUTs first */
5 for(int i=0;i<91;i++) {
6     /* i - number of degrees */
7     sinT[i] = sin(i*PI/180.0);
8     cosT[i] = cos(i*PI/180.0);
9 }
10 /* Now test it */
11 a = 60; /* angle */
12 test = ((a<0)?-1:1)*sinT[60]; /* test = sin(60 ) = 0.866 */
13 test = cosT[abs(a)]; /* test = cos(60 ) = 0.5 */
14
15 a = -60;
16 test = ((a<0)?-1:1)*sinT[60]; /* test = sin(-60 ) = -0.866 */
17 test = cosT[abs(a)]; /* test = cos(60 ) = 0.5 */

```

Creating a LUT table for $\arctan()$ function is more complicated, because function is defined on whole \mathbb{R} plane $(-\infty, +\infty)$. Additionally, the function changes dramatically functional values around zero. This led to implementation which uses two LUT tables, one with small step (0.1 between two records in a table) and high precision on interval from $[-10, 10]$ and second with long step (10 between two records in a table). In this way we can save memory while keeping high precision around zero input.

6.1.5 Using LUT tables in fixed point

At this point we have tried to solve already well known problem: FP operations. Functional values of all used trigonometric functions are decimal values, but the processor itself do not have support for floating point and hence it takes time to process them.

To solve this problem we used the same principle as with Deriche smoother. We wanted to preserve at least 4 decimal places to keep our calculations precise, so we decided to use bit shifting of 14 places which multiplies functional values by more than 10^4 . As a consequence, the results of all calculations have to be divided by 16384 (2^{14}).

Listing 13: Used implementation of LUT tables in fixed point

```

1 int cosT[91]; int sinT[91];
2
3 /* initialize LUTs */
4 for(int i=0;i<91;i++) {
5     /* i - number of degrees */
6     sinT[i] = sin(i*PI/180.0)<<14;
7     cosT[i] = cos(i*PI/180.0)<<14;
8 }
9
10 a = 60; /* angle */
11 test = (100*((a<0)?-1:1)*sinT[60])>>14; /* sinT[60] = 0.866*16384 = 14188 */
12                                     /* test = (14188*100)>>14 = 86 */

```

6.1.6 Utilizing DMA

We have decided not to utilize DMA channel. Our image matrix fits into DSP's internal SRAM memory and it's encoding is different from representation in camera/LCD driver. Conversion needs to be done anyway and therefore DMA utilisation is useless in our case.

6.2 Performance analysis

Performance was evaluated using approach described in chapter 2.4 Profiling techniques and 1.2 Measurement methodology. We are making 10 tests and use average values to calculate processign times, frames per second (fps), operations per pixel (oppx) and optimization gain. Memory usage and memory calls are evaluated from the source code.

Function name	time [%]	time [ms]	fps	oppx ¹⁶	mem. usage ¹⁷	mem. acc. ¹⁸	gain ¹⁹
Deriche smoother	23.72	18.97	52.70	187.85	241.2kB	4rd + 4wr	8.6
Deriche derivator	18.01	14.40	69.40	142.64	609.6kB	2rd + 1wr	127.83
Hough Transform							
Printing result	7.52	6.02	166.01	59.63	369.6kB	-	0.62

Table 6: Profiling results of algorithm optimized version (without compiler help)

¹⁶Operations per pixel

¹⁷Memory usage refers to total amount of memory which is managed by the algorithm

¹⁸Theoretical value per pixel

¹⁹Computed with respect to previous version

New version of Hough transform in combination with LUT tables and fixed point operations really gave it a boost. Specially new algorithm for Hough transform fused with Deriche derivator is over hundred times faster than it's predecessors. It is a great result and the key to get real-time performance out of the device. Algorithm manifests only 2 read accesses to count the gradient and 1 write access to give a vote in the Hough accumulator.

Deriche smoother shows improvement by almost 9 times. Algorithm is the same, only thing which changed is using of fixed point calculations and avoiding to division operations. Printing lines shows negative improvement, which is caused by image properties and by processing chain settings, as is explained in previous chapters. Significant part (50.74%) of processing time is outside image processing chain - it includes BIOS functions, transferring camera/screen²⁰ data and so on..

Function name	time [%]	time [ms]	fps	oppx ¹⁶	mem. usage ¹⁷	mem. acc. ¹⁸	gain ¹⁹
Deriche smoother	5.34	4.27	233.93	42.32	241.2kB	4rd + 4wr	4.44
Deriche derivator	13.51	10.81	92.52	106.99	609.6kB	2rd + 1wr	1.38
Hough Transform							
Printing result	7.01	5.60	178.35	55.50	369.6kB	-	1.07

Table 7: Profiling results of algorithm optimized version (with compiler help)

From Table 7 we can see that compiler managed to increase performance significantly. Version without compiler optimizations runs in real-time already, but this version reaches much further, almost to 50 fps. We don't know what compiler did to increase the performance, but now a majority of the processing time is outside our processing chain, which suggest that we are close to performance limit.

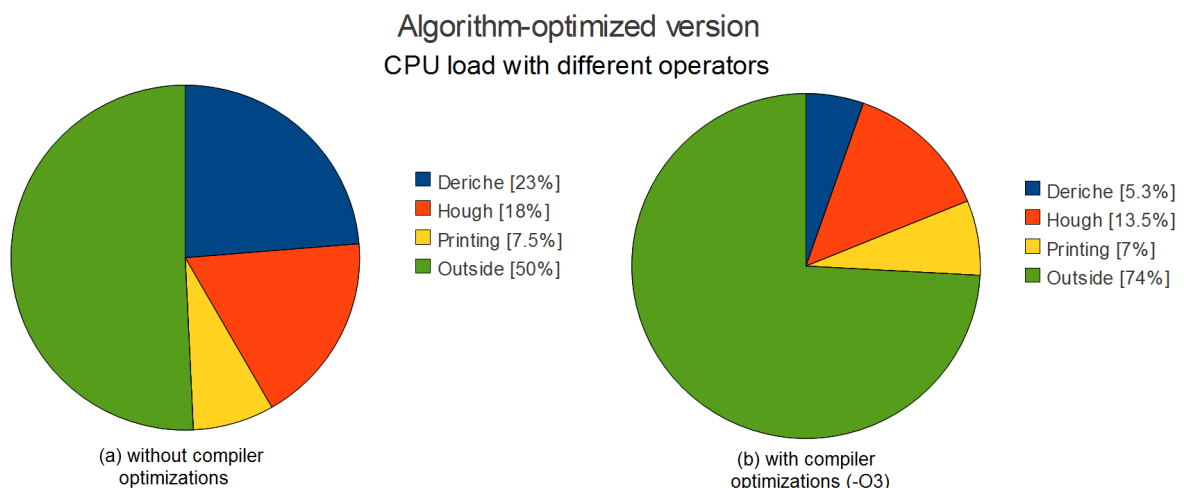


Figure 9: Algorithm-optimized version - CPU load by different operators

²⁰Transferring data between peripherals and external memory is done by EDMA without CPU intervention.

Following summary shows most important performance indicators of this version. It is resulting fps, total CPU load created by image processing chain, amount of operations per pixel (oppx), memory usage and optimization gain between two consecutive versions.

Indicator	Value	Optimized
FPS	25.37	48.33
OPPX	390	204
CPU load	49.25%	25.86%
Mem. usage	610.8kB	610.8kB
Speed gain	50.98	1.9

Table 8: Summary of optimized version

6.3 Conclusions

Because of better algorithms, lookup tables instead of trigonometric calculations and fixed point operations we get more than real-time performance. Optimized version manifests almost 50 fps, which is twice the real-time requirements. This results exceeds our expectations and satisfies our requirements.

7 Conclusion

In a nutshell, we have created 3 versions where we applied algorithm and architecture related optimizations. At the end we managed to get result running over 100 times faster than the basic version. There is tremendous difference between first and last version performances. Hough transform got over 200 times faster, Deriche smoother manifests almosts 40 times faster behaviour without changing the algorithm. Deriche derivator got 4 times faster thanks to register rotation. See Table 9 for complete list of optimization gains between first and last version. Please note that the term "gain" refers to speed performance. To complete description how values in Table 9 were calculated, see chapter 1.2 Measurement methodology.

Operator	Frames per second		Gain
	Basic version	Optimized version	
Deriche smoother	6.06	233.93	38.6
Deriche derivator	12.39	92.53	209.27
Hough Transform	0.46	178.36	0.57
Printing result	312.64	48.33	117.45
System	0.41	48.33	117.45

Table 9: Optimization gain

The very first version was running on a PC and we haven't really thought about optimization. Main goal was to write a starting point: working solution which can be easily ported to a DSP. When we did our first version on a DSP, the performance was so poor that the result was unusable in practice.

So, we decided to use the same algorithms and try to get advantage of the DSP capabilities. The results was algorithm-architecture matched version, which was faster by over 20% while keeping original algorithms and principles. Relatively speaking, it is a good result, but the performance was still quite poor. Partially, also because we haven't moved from floating point to fixed point with our calculations. This was requested for the very last version.

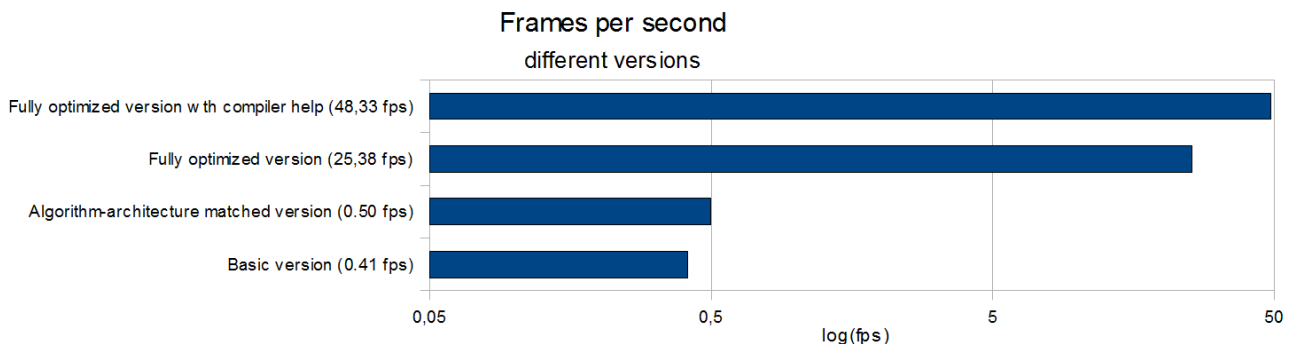


Figure 10: Compared FPS for different versions (on logarithmic scale)

The last version combines all optimization techniques known to us. Firstly, we have used $O(n^2)$ algorithm for Hough transform. Secondly, all calculations were moved from floating point to a fixed point and integer LUT tables replaced floating point trigonometric calculations. All together it created version over 50 times faster than previous. Additionally, compiler optimizations were able to double speed performance to almost 50 fps.

Personally, this project was a challenge for us. First of all used hardware equipment is very advanced. This was a first time we have got around parallel processing on instruction level. Then this was a first time we've been working on an embedded system related to image processing, which is for sure interesting topic for both of us. We have learned many things from this project, it is not only new programming and optimizing experience, we have also got around unexpected results - for example how big difference can relatively small amount of floating point operations do on performance, or how effective register rotations can be.

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8 Appendix A - Basic implementation on a PC

Listing 14: Used implementation of LUT tables in fixed point

```
1  /*****
2  /* Simple test program to exercise/demonstrate some functionality of IMPACT.
3  *
4  */
5  #include <stdio.h>
6  #include <sys/types.h>
7  #include <stdlib.h>
8  #include <stdint.h>
9  #include <float.h>
10 #include "mcimage.h"
11 #include "mccodimage.h"
12
13 #define FOR_EACH_PIXEL(inc, data) for(inc=0; inc<(image->row_size*image->col_size);inc++) {
14     data }
15
16 // it is a prototype, not a function
17 void processing(float *image, // (IN) Matrix with source image
18     float *result, // (OUT) Matrix with resulting image
19     unsigned int width, // (IN) Image width
20     unsigned int height, // (IN) Image height
21     int argc, char *argv[]);
22
23 // DO NOT MESS WITH THIS !!!
24 void store_me_fimage(char *filename, float *m, int width, int height)
25 {
26     struct xvimage *result; int i, max = 0;
27     result = allocimage(filename,height,width,1,1);
28
29     for(i=0; i<(height*width); i++)
30         if(max<*(m+i)) max=*(m+i);
31
32     for(i=0; i<(height*width); i++)
33         result->imagedata[i]=(unsigned char)((*(m+i)/max)*255);
34
35     writeimage(result, filename);
36     freeimage (result);
37 }
38
39 // DO NOT MESS WITH THIS !!!
40 int main (int argc, char *argv[])
41 {
42     // define image structure - PINK library
43     struct xvimage * image; struct xvimage * result;
44     float *fimage;float *fresult; int i; float max = 0;
45
46     // read image PGM - ASCII
47     image = readimage(argv[1]); // lecture d'une image au format PGM !
48
49     if (image == NULL)
50     {
51         fprintf(stderr, "addconst: readimage failed\n");
52         exit(0);
53     }
54
55     // result image allocation
56     result = allocimage("result",image->row_size,image->col_size,1,1);
57
58     // allocate float matrixes
59     fimage = malloc(sizeof(float)*image->row_size*image->col_size);
60     fresult = malloc(sizeof(float)*image->row_size*image->col_size);
61
62     // do the image processing with float matrixes and progressive image colordepth
63     FOR_EACH_PIXEL(i, { *(fimage+i)=(float)image->imagedata[i]; })
64     processing (fimage, fresult, image->row_size, image->col_size, argc, argv);
65     FOR_EACH_PIXEL(i, { if(max<*(fresult+i)) max=*(fresult+i); })
66     FOR_EACH_PIXEL(i, { result->imagedata[i]=(unsigned char)((*(fresult+i)/max)*255); })
67
68     // write result image
```

```

68 writeimage(result, argv[2]);
69 freeimage (image);
70 return (0);
71
72 }

```

Listing 15: project.c - image processing chain implementation

```

1 #include <math.h>
2 #include <stdlib.h>
3
4
5 #define gpx(x,y) m[x+(y)*width]
6 #define ADDR(in, x,y) in[(x) + (y)*m]
7 #define make_me_fimage(p, width, height) p = malloc(sizeof(float)*width*height)
8
9 #ifndef PI
10 #define PI 3.14159265
11 #endif
12
13 extern void store_me_fimage(char *filename, float *m, int width, int height);
14
15
16 void deriche_gl(float *Id, float *Ic, float a, int m, int n) {
17
18     float ga = exp(-a);
19     float gb = 1-ga;
20     float g1 = (1-a)*(1-a);
21     float g2 = 2*a;
22     float gg = g*a;
23
24     int i, j;
25     float Ir[m*n];
26
27     // Horizontal filtering
28     for(i = 0; i < n; i++) {
29         // Filter causal
30         ADDR(Ic,0,i) = ADDR(Id,0,i);
31         ADDR(Ic,1,i) = ADDR(Id,1,i);
32
33         for(j = 2; j < m; j++) {
34             ADDR(Ic,j,i) = (g1*ADDR(Id,j,i)+g2*ADDR(Ic,j-1,i)+gg*ADDR(Ic,j-2,i))/2;
35         }
36
37         // Filter anticausal
38         ADDR(Ir,m-1,i) = ADDR(Id,m-1,i);
39         ADDR(Ir,m-2,i) = ADDR(Id,m-2,i);
40
41         for(j = m-3; j >= 0 ; j--) {
42             ADDR(Ir,j,i) = (g1*ADDR(Ic,j,i) + (g2*ADDR(Ir,j+1,i) + (gg*ADDR(Ir,j+2,i))))/2;
43         }
44     }
45
46
47     // Vertical filtering
48     for(i = 0; i < m; i++) {
49         // Filter causal
50         ADDR(Ic,i,0) = ADDR(Ir,i,0);
51         ADDR(Ic,i,1) = ADDR(Ir,i,1);
52
53         for(j = 2; j < n; j++) {
54             ADDR(Ic,i,j) = (g1*ADDR(Ir,i,j) + (g2*ADDR(Ic,i,j-1) + (gg*ADDR(Ic,i,j-2))))/2;
55         }
56
57         // Filter anticausal
58         ADDR(Ir,i,n-1) = ADDR(Ic,i,n-1);
59         ADDR(Ir,i,n-2) = ADDR(Ic,i,n-2);
60
61         for(j = n-3; j >= 0 ; j--) {
62             ADDR(Ir,i,j) = (g1*ADDR(Ic,i,j) + (g2*ADDR(Ir,i,j+1) + (gg*ADDR(Ir,i,j+2))))/2;
63         }
64     }

```

```

65 }
66
67 void roberts(float *m, float *mout, float threshold, unsigned int width, unsigned int height)
68 {
69     float convy, convx; int j, k;
70     for(k=0; k<height-1;k++) // Lines
71         for(j=0; j<width-1;j++) { // Columns
72             //ROBERTSON
73             convx = -gpx(j,k)-gpx(j+1,k)+gpx(j,k+1)+gpx(j+1,k+1);
74             convy = -gpx(j,k)+gpx(j+1,k)-gpx(j,k+1)+gpx(j+1,k+1);
75             //SOBEL
76             //convy = -gpx(j-1,k-1)-2*gpx(j,k-1)-gpx(j+1,k-1)+gpx(j-1,k+1)+2*gpx(j,k+1)+gpx(j+1,k
77                 +1);
78             //convx = -gpx(j-1,k-1)-2*gpx(j-1,k)-gpx(j-1,k+1)+gpx(j+1,k-1)+2*gpx(j+1,k)+gpx(j+1,k
79                 +1);
80             if (convy<0) convy = -convy; if (convx<0) convx = -convx; // abs. val. done manually
81             mout[j+k*width] = (convy+convx >= threshold) ? 255 : 0;
82         }
83     }
84
85 float * hough_lines_slow(float *m, unsigned int width, unsigned int height) {
86     /* Run-in-vars */ int k, j, a; float rho; int mrho = (int)sqrt(width*width + height*height)
87     ;
88     /* Make the image */ float *h = malloc(sizeof(float)*mrho*180);
89     for(k=1; k<height-1;k++) // Lines
90         for(j=1; j<width-1;j++) // Columns
91             if(m[j + k*width]) for(a=0;a<180;a++) { // Angles
92                 rho = (j*cos(a*PI/180.0) + k*sin(a*PI/180.0));
93                 if(rho) h[a+180*((int)rho+mrho)/2]++; }
94     return h;
95 }
96
97 void print_line(float *m, float intensity, int x1, int y1, int x2, int y2, unsigned int
98     width, unsigned int height)
99 {
100     int i; int d = (int)sqrt((x2-x1)*(x2-x1)+(y2-y1)*(y2-y1)); /* line size */
101     float ax=(x2-x1)/(float)d, ay=(y2-y1)/(float)d; /* direction vectors */
102     for(i=0; i<d; i++) /* print pixel */ m[(x1+(int)(ax*i))+(y1+(int)(ay*i))*width] =
103         intensity;
104 }
105
106 void print_lines(float*m, float *h, unsigned int threshold, unsigned int width, unsigned int
107     height)
108 {
109     int j,k, x[2], y[2], hct; int mrho = (int)sqrt(width*width + height*height);
110     int sx=0, sy=0, ex=sx+width, ey=sy+height, is1_x, is2_y, is3_x, is4_y;
111     float x1, y1, ax, ay, rho;
112
113     for(k=0; k<mrho;k++) { rho = k*2-mrho;
114         for(j=0; j<180;j++) if(h[j+180*k] > threshold) { hct = 0;
115             x1 = rho*cos(j*PI/180.0); y1 = rho*sin(j*PI/180.0); /* get a point on line */
116             ax = y1/(float)j; ay = -x1/(float)j; /* get direction vector */
117             is1_x = x1-sx+(sy-y1)*ax/ay; is2_y = y1-sy+(ex-x1)*ay/ax; /* calculate intersections
118                 */
119             is3_x = x1-sx+(ey-y1)*ax/ay; is4_y = y1-sy+(sx-x1)*ay/ax;
120             if((is1_x>=sx)&&(is1_x<ex)) { x[hct]=is1_x; y[hct++]=sy; } /* find hits */
121             if((is3_x>=sx)&&(is3_x<ex)) { x[hct]=is3_x; y[hct++]=ey; }
122             if((is2_y>=sy)&&(is2_y<ey)) { x[hct]=ex; y[hct++]=is2_y; }
123             if((is4_y>=sy)&&(is4_y<ey)) { x[hct]=sx; y[hct++]=is4_y; }
124             if(hct==2) print_line(m, 255, x[0]-sx, y[0]-sy, x[1]-sx, y[1]-sy, width, height); /* print
125                 */
126         }
127     }
128 }
129
130 void processing(float *m, // (IN) Matrix with source image
131     float *mout, // (OUT) Matrix with resulting image
132     unsigned int width, // (IN) Image width
133     unsigned int height, // (IN) Image height
134     int argc, char *argv[])
135 {
136
137     float *h; float dercf, edgecf, trshcf; int j;

```



```

130
131     sscanf(argv[3], "%f", &dercf);
132     sscanf(argv[4], "%f", &edgecf);
133     sscanf(argv[5], "%f", &trshcf);
134
135     deriche_gl(m, mout, dercf, width, height);
136     store_me_fimage("deriche.pgm", mout, height, width);
137     roberts(mout, mout, edgecf, width, height);
138     store_me_fimage("robertson.pgm", mout, height, width)
139     h = hough_lines_slow(mout, mout, edgecf, width, height);;
140     store_me_fimage("hough.pgm", h, sqrt(width*width+height*height), 180);
141     print_lines(mout, h, trshcf, width, height);
142
143 }

```

Listing 16: testme.sh - script used to speedup development process

```

1 if [ $# = 5 ]; then
2   rm hough.pgm deriche.pgm robertson.pgm
3   gcc -g -lm main.c mcimage.c project.c -o r.out
4   if [ $? = 0 ]; then
5     ./r.out $1 $2 $3 $4 $5
6     display $1 deriche.pgm robertson.pgm hough.pgm $2
7   fi
8 else
9   echo "usage ./testme imgin.pgm imgout.pgm deriche_coef edge_binarisation_coef
10  trsholding_coef"
11 fi

```

Listing 17: profile-prj.sh - script used to profile program performance

```

1 rm a.out
2 rm gmon.out
3
4 gcc -pg -lm main.c mcimage.c project.c -o a.out
5
6 ./a.out test.pgm test2.pgm 0.8 160 120
7
8
9 rm gmon.sum
10 mv gmon.out gmon.sum
11
12 gprof -s a.out gmon.out gmon.sum
13 ./a.out test.pgm test2.pgm 0.8 160 120
14 gprof -s a.out gmon.out gmon.sum
15 ./a.out test.pgm test2.pgm 0.8 160 120
16 gprof -s a.out gmon.out gmon.sum
17 ./a.out test.pgm test2.pgm 0.8 160 120
18 gprof -s a.out gmon.out gmon.sum
19 ./a.out test.pgm test2.pgm 0.8 160 120
20 gprof -s a.out gmon.out gmon.sum
21 ./a.out test.pgm test2.pgm 0.8 160 120
22 gprof -s a.out gmon.out gmon.sum
23 ./a.out test.pgm test2.pgm 0.8 160 120
24 gprof -s a.out gmon.out gmon.sum
25
26 gprof a.out gmon.sum > $1

```

Listing 18: result-prof.txt - profiling results

```

1 Flat profile:
2
3 Each sample counts as 0.01 seconds.
4 % cumulative self self total name
5 time seconds seconds calls ms/call ms/call
6 93.75 0.15 0.15 7 21.43 21.43 hough_lines_slow
7 6.25 0.16 0.01 main
8 0.00 0.16 0.00 154 0.00 0.00 print_line
9 0.00 0.16 0.00 35 0.00 0.00 allocimage

```

```

10 0.00 0.16 0.00 28 0.00 0.00 freeimage
11 0.00 0.16 0.00 28 0.00 0.00 writeimage
12 0.00 0.16 0.00 28 0.00 0.00 writerawimage
13 0.00 0.16 0.00 21 0.00 0.00 store_me_fimage
14 0.00 0.16 0.00 7 0.00 0.00 deriche_gl
15 0.00 0.16 0.00 7 0.00 0.00 print_lines
16 0.00 0.16 0.00 7 0.00 21.43 processing
17 0.00 0.16 0.00 7 0.00 0.00 readimage
18 0.00 0.16 0.00 7 0.00 0.00 roberts

```

```

19
20 %           the percentage of the total running time of the
21 time        program used by this function.
22
23 cumulative a running sum of the number of seconds accounted
24 seconds    for by this function and those listed above it.
25
26 self       the number of seconds accounted for by this
27 seconds     function alone. This is the major sort for this
28             listing.
29
30 calls      the number of times this function was invoked, if
31             this function is profiled, else blank.
32
33 self       the average number of milliseconds spent in this
34 ms/call     function per call, if this function is profiled,
35             else blank.
36
37 total      the average number of milliseconds spent in this
38 ms/call     function and its descendents per call, if this
39             function is profiled, else blank.
40
41 name       the name of the function. This is the minor sort
42             for this listing. The index shows the location of
43             the function in the gprof listing. If the index is
44             in parenthesis it shows where it would appear in
45             the gprof listing if it were to be printed.

```

```

46
47           Call graph (explanation follows)

```

```

48
49
50 granularity: each sample hit covers 4 byte(s) for 6.25% of 0.16 seconds

```

```

51
52 index % time    self  children   called    name
53                                     <spontaneous>
54 [1]   100.0     0.01   0.15      7/7      main [1]
55                                     processing [3]
56                                     freeimage [6]
57                                     writeimage [7]
58                                     allocimage [5]
59                                     readimage [12]
60 -----
61                                     0.15   0.00      7/7      processing [3]
62 [2]   93.8      0.15   0.00      7        hough_lines_slow [2]
63 -----
64                                     0.00   0.15      7/7      main [1]
65 [3]   93.8      0.00   0.15      7        processing [3]
66                                     0.15   0.00      7/7      hough_lines_slow [2]
67                                     0.00   0.00     21/21     store_me_fimage [9]
68                                     0.00   0.00      7/7      print_lines [11]
69                                     0.00   0.00      7/7      roberts [13]
70                                     0.00   0.00      7/7      deriche_gl [10]
71 -----
72                                     0.00   0.00    154/154   print_lines [11]
73 [4]   0.0       0.00   0.00     154      print_line [4]
74 -----
75                                     0.00   0.00      7/35     main [1]
76                                     0.00   0.00      7/35     readimage [12]
77                                     0.00   0.00     21/35     store_me_fimage [9]
78 [5]   0.0       0.00   0.00      35      allocimage [5]
79 -----
80                                     0.00   0.00      7/28     main [1]
81                                     0.00   0.00     21/28     store_me_fimage [9]

```

82	[6]	0.0	0.00	0.00	28	freeimage [6]
83						
84			0.00	0.00	7/28	main [1]
85			0.00	0.00	21/28	store_me_fimage [9]
86	[7]	0.0	0.00	0.00	28	writeimage [7]
87			0.00	0.00	28/28	writerawimage [8]
88						
89			0.00	0.00	28/28	writeimage [7]
90	[8]	0.0	0.00	0.00	28	writerawimage [8]
91						
92			0.00	0.00	21/21	processing [3]
93	[9]	0.0	0.00	0.00	21	store_me_fimage [9]
94			0.00	0.00	21/28	freeimage [6]
95			0.00	0.00	21/28	writeimage [7]
96			0.00	0.00	21/35	allocimage [5]
97						
98			0.00	0.00	7/7	processing [3]
99	[10]	0.0	0.00	0.00	7	deriche_gl [10]
100						
101			0.00	0.00	7/7	processing [3]
102	[11]	0.0	0.00	0.00	7	print_lines [11]
103			0.00	0.00	154/154	print_line [4]
104						
105			0.00	0.00	7/7	main [1]
106	[12]	0.0	0.00	0.00	7	readimage [12]
107			0.00	0.00	7/35	allocimage [5]
108						
109			0.00	0.00	7/7	processing [3]
110	[13]	0.0	0.00	0.00	7	roberts [13]

111

112
113 This table describes the call tree of the program, and was sorted by
114 the total amount of time spent in each function and its children.

115

116 Each entry in this table consists of several lines. The line with the
117 index number at the left hand margin lists the current function.
118 The lines above it list the functions that called this function,
119 and the lines below it list the functions this one called.

120 This line lists:

121 index A unique number given to each element of the table.
122 Index numbers are sorted numerically.
123 The index number is printed next to every function name so
124 it is easier to look up where the function in the table.

125

126 % time This is the percentage of the 'total' time that was spent
127 in this function and its children. Note that due to
128 different viewpoints, functions excluded by options, etc,
129 these numbers will NOT add up to 100%.

130

131 self This is the total amount of time spent in this function.

132

133 children This is the total amount of time propagated into this
134 function by its children.

135

136 called This is the number of times the function was called.
137 If the function called itself recursively, the number
138 only includes non-recursive calls, and is followed by
139 a '+' and the number of recursive calls.

140

141 name The name of the current function. The index number is
142 printed after it. If the function is a member of a
143 cycle, the cycle number is printed between the
144 function's name and the index number.

145

146

147 For the function's parents, the fields have the following meanings:

148

149 self This is the amount of time that was propagated directly
150 from the function into this parent.

151

152 children This is the amount of time that was propagated from
153 the function's children into this parent.

154

```

155     called This is the number of times this parent called the
156 function '/' the total number of times the function
157 was called. Recursive calls to the function are not
158 included in the number after the '/'.
159
160     name This is the name of the parent. The parent's index
161 number is printed after it. If the parent is a
162 member of a cycle, the cycle number is printed between
163 the name and the index number.
164
165 If the parents of the function cannot be determined, the word
166 '<spontaneous>' is printed in the 'name' field, and all the other
167 fields are blank.
168
169 For the function's children, the fields have the following meanings:
170
171     self This is the amount of time that was propagated directly
172 from the child into the function.
173
174     children This is the amount of time that was propagated from the
175 child's children to the function.
176
177     called This is the number of times the function called
178 this child '/' the total number of times the child
179 was called. Recursive calls by the child are not
180 listed in the number after the '/'.
181
182     name This is the name of the child. The child's index
183 number is printed after it. If the child is a
184 member of a cycle, the cycle number is printed
185 between the name and the index number.
186
187 If there are any cycles (circles) in the call graph, there is an
188 entry for the cycle-as-a-whole. This entry shows who called the
189 cycle (as parents) and the members of the cycle (as children.)
190 The '+' recursive calls entry shows the number of function calls that
191 were internal to the cycle, and the calls entry for each member shows,
192 for that member, how many times it was called from other members of
193 the cycle.
194
195
196 Index by function name
197
198     [5] allocimage           [4] print_line           [9] store_me_fimage
199     [10] deriche_gl         [11] print_lines        [7] writeimage
200     [6] freeimage           [3] processing          [8] writerawimage
201     [2] hough_lines_slow    [12] readimage
202     [1] main                 [13] roberts

```

9 Appendix B - Basic implementation on a DSP

Listing 19: Basic implementation on a DSP

```
1 /**
2  Completely naive version of implementing given image processing chain.
3
4  Authors: Divij Babbar, Kubicka Matej (I4-IMC)
5  Date: 1/6/2012
6  Version: Naive
7
8  Properties:
9  -----
10 sin() LUT          NOT USED
11 cos() LUT          NOT USED
12 atan() LUT         NOT USED
13 hough version      STANDARD
14 matrix datatype    FLOAT
15 loop unrolling     NOT USED
16
17 */
18
19 #include <std.h>
20 #include <gio.h>
21 #include <log.h>
22 #include <math.h>
23
24 #include "psp_vpfe.h"
25 #include "psp_vpbe.h"
26 #include "fvid.h"
27
28 #include "psp_tvp5146_extVidDecoder.h"
29
30 #include <soc.h>
31 #include <cslr_ccdc.h>
32
33 #include <soc.h>
34 #include <cslr_sysctl.h>
35
36
37 //pour logger ce qui se passe avec log.h (voir DSPBIOS)
38 extern LOG_Obj trace; // BIOS LOG object
39
40
41 /* extrait de l'exemple EDMA3 :
42 // 48K L1 SRAM [0x10f04000, 0x10f10000), 0xc000 length
43 // 32K L1 Dcache [0x10f10000, 0x10f18000), 0x8000 length
44 // 128K L2 SRAM [0x10800000, 0x10820000), 0x20000 length
45 // 128M DDR2 [0x80000000, 0x88000000), 0x8000000 length are cacheable
46 */
47
48 #define width 300
49 #define height 200
50 #define ADDR(in, x,y) in[(x) + (y)*width]
51 #define gpx(x,y) m[(x)+(y)*width]
52
53 #ifndef PI
54 #define PI 3.14159265
55 #endif
56
57 #define NO_OF_BUFFERS          (2u)
58 #define width 300
59 #define height 200
60
61 // Global Variable Defined
62 static PSP_VPSSSurfaceParams *ccdcAllocFB[NO_OF_BUFFERS]={NULL};
63 static PSP_VPSSSurfaceParams *vidAllocFB[NO_OF_BUFFERS] = {NULL};
64
65 static FVID_Handle    ccdcHandle;
66 static FVID_Handle    vid0Handle;
67 static FVID_Handle    vencHandle;
68
```

```

69 static PSP_VPFE_TVP5146_ConfigParams tvp5146Params = {
70     TRUE, // enable656Sync
71     PSP_VPFE_TVP5146_FORMAT_COMPOSITE, // format
72     PSP_VPFE_TVP5146_MODE_AUTO // mode
73 };
74
75 static PSP_VPFECcdcConfigParams ccdcParams = {
76     PSP_VPFE_CCDC_YCBCR_8, // dataFlow
77     PSP_VPSS_FRAME_MODE, // ffMode
78     480, // height
79     720, // width
80     (720 *2), // pitch
81     0, // horzStartPix
82     0, // vertStartPix
83     NULL, // appCallback
84     {
85         PSP_VPFE_TVP5146_Open, // extVD Fxn
86         PSP_VPFE_TVP5146_Close,
87         PSP_VPFE_TVP5146_Control,
88     },
89     0 //segId
90 };
91
92 static PSP_VPBE0sdConfigParams vid0Params = {
93     PSP_VPSS_FRAME_MODE, // ffmode
94     PSP_VPSS_BITS16, // bitsPerPixel
95     //PSP_VPBE_RGB_888, //ajout TG
96     PSP_VPBE_YCbCr422, // colorFormat
97     (720 * (16/8u)), // pitch
98     0, // leftMargin
99     0, // topMargin
100    720, // width
101    480, // height
102    0, // segId
103    PSP_VPBE_ZOOM_IDENTITY, // hScaling
104    PSP_VPBE_ZOOM_IDENTITY, // vScaling
105    PSP_VPBE_EXP_IDENTITY, // hExpansion
106    PSP_VPBE_EXP_IDENTITY, // vExpansion
107    NULL // appCallback
108 };
109
110 static PSP_VPBEVencConfigParams vencParams = {
111     PSP_VPBE_DISPLAY_NTSC_INTERLACED_COMPOSITE // Display Standard
112 };
113
114 #pragma DATA_SECTION(m, ".ExtBuffer")
115 float m[width*height];
116
117 #pragma DATA_SECTION(m2, ".ExtBuffer")
118 float m2[width*height];
119
120 #pragma DATA_SECTION(h, ".ExtBuffer")
121 short h[360*180];
122
123 unsigned int profiling[50];
124
125
126
127 void robertson(float treshold)
128 {
129     float convy, convx; int j, k;
130     for(k=0; k<height-1;k++) // Lines
131         for(j=0; j<width-1;j++) { // Columns
132             //ROBERTSON
133             convx = -gpx(j,k)-gpx(j+1,k)+gpx(j,k+1)+gpx(j+1,k+1);
134             convy = -gpx(j,k)+gpx(j+1,k)-gpx(j,k+1)+gpx(j+1,k+1);
135             if (convy<0) convy = -convy; if (convx<0) convx = -convx; // abs. val. done manually
136             m2[j+k*width] = (convy+convx >= treshold) ? 255 : 0;
137         }
138 }
139
140
141 void transpose(float *in, float *out, int w, int h)

```

```

142 {
143     int n,m,s=0;
144     for(n=0; n<h;n++)
145         for(m=0; m<w;m++)
146             out[m*h+n] = in[s++];
147 }
148
149 float l[width];
150 float la[width*height];
151 float lb[width*height];
152 float lc[width*height];
153 float ld[width*height];
154
155 void deriche_gl(float g)
156 {
157     float g1 = (1-g)*(1-g);
158     float g2 = 2*g;
159     float gg = g*g;
160
161     int i,ii,k;
162
163     for(i = 0 ; i < width ; i++){ // lines
164         ii=i*height;
165
166         la[ii+0]=g1*m2[ii];
167         la[ii+1]=(g1*m2[1]+g2*la[ii+0]);
168         for(k=2;k<height;k++){
169             la[ii+k] = (g1*m2[ii+k] + g2*la[ii+k-1] - gg*la[ii+k-2]);
170         }
171
172         lb[ii+height-1]=la[ii+height-1];
173         lb[ii+height-2]=la[ii+height-2];
174         for(k=height-3;k>=0;k--){
175             lb[ii+k] = (g1*la[ii+k] + g2*lb[ii+k+1] - gg*lb[ii+k+2]);
176         }
177     }
178
179     transpose(lb, lc, height, width);
180
181     for(i = 0 ; i < height ; i++){ // lines
182         ii=i*width;
183         ld[ii+0]=g1*lc[ii+0];
184         ld[ii+1]=(g1*lc[ii+1]+g2*ld[ii+0]);
185         for(k=2;k<width;k++){
186             ld[ii+k] = (g1*lc[ii+k] + g2*ld[ii+k-1] - gg*ld[ii+k-2]);
187         }
188
189         m[ii+width-1]=ld[ii+width-1];
190         m[ii+width-2]=ld[ii+width-2];
191         for(k=width-3;k>=0;k--){
192             m[ii+k] = (g1*ld[ii+k] + g2*m[ii+k+1] - gg*m[ii+k+2]);
193         }
194     }
195 }
196 }
197
198 void print_line(float intensity, int x1, int y1, int x2, int y2)
199 {
200     int i; int d = (int)sqrt((x2-x1)*(x2-x1)+(y2-y1)*(y2-y1)); /* line size */
201     float ax=(x2-x1)/(float)d, ay=(y2-y1)/(float)d; /* direction vectors */
202     for(i=0; i<d; i++) /* print pixel */ m[(x1+(int)(ax*i))+(y1+(int)(ay*i))*width] =
        intensity;
203 }
204
205 void print_lines_slow(unsigned int threshold)
206 {
207     int j,k, x[2], y[2], hct;
208     int sx=0, sy=0, ex=sx+width, ey=sy+height, is1_x, is2_y, is3_x, is4_y;
209     float x1, y1, ax, ay, rho;
210
211     for(k=0; k<360;k++) { rho = k*2-360;
212         for(j=0; j<180;j++)
213             if(h[j+180*k] > threshold) {

```

```

214     hct = 0;
215     x1 = rho*cos(j*PI/180.0); y1 = rho*sin(j*PI/180.0); /* get a point on line */
216     ax = y1/(float)j; ay = -x1/(float)j; /* get direction vector */
217     is1_x = x1-sx+(sy-y1)*ax/ay; is2_y = y1-sy+(ex-x1)*ay/ax; /* calculate
        intersections */
218     is3_x = x1-sx+(ey-y1)*ax/ay; is4_y = y1-sy+(sx-x1)*ay/ax;
219     if((is1_x>=sx)&&(is1_x<ex
220     )) { x[hct]=is1_x; y[hct++]=sy; } /* find hits */
221     if((is3_x>=sx)&&(is3_x<ex)) { x[hct]=is3_x; y[hct++]=ey; }
222     if((is2_y>=sy)&&(is2_y<ey)) { x[hct]=ex; y[hct++]=is2_y; }
223     if((is4_y>=sy)&&(is4_y<ey)) { x[hct]=sx; y[hct++]=is4_y; }
224     if(hct==2) print_line(255, x[0]-sx, y[0]-sy, x[1]-sx, y[1]-sy); /* print */
225 }
226 }
227 }
228
229 void hough_lines_slow() {
230     /* Run-in-vars */ int k,j, a; float rho;
231
232     for(k=0; k<360*180;k++)
233         h[k] = 0;
234
235     for(k=1; k<height;k++) // Lines
236         for(j=1; j<width;j++) // Columns
237             if(m2[j + k*width]) for(a=0;a<180;a++) { // Angles
238                 rho = (j*cos(a*PI/180.0) + k*sin(a*PI/180.0));
239                 //rho = (j*cosT[a] + k*sinT[a]);
240                 if(rho) h[a+180*(((int)rho+360)/2)]++; }
241 }
242
243 void start_boucle() {
244     PSP_VPBChannelParams beinitParams;
245     PSP_VPFChannelParams feinitParams;
246     GPIO_Attrs gpioAttrs = GPIO_ATTRS;
247     PSP_VPSSSurfaceParams *FBAddr = NULL;
248     PSP_VPSSSurfaceParams *FBAddrOut = NULL;
249     int i,v = 0;
250     Uint32 j = 0;
251     Uint32 k = 0;
252
253     //Init CSL du DMA
254     edma3init();
255
256     // Create ccdc channel
257     feinitParams.id = PSP_VPFE_CCDC;
258     feinitParams.params = (PSP_VPFECcdcConfigParams*)&ccdcParams;
259     ccdcHandle = FVID_create( "/VPFE0", IOM_INOUT, NULL, &feinitParams,
260                             &gpioAttrs);
261     if ( NULL == ccdcHandle) {
262         return;
263     }
264
265     // Configure the TVP5146 video decoder
266     if( FVID_control( ccdcHandle,
267                     VPFE_ExtVD_BASE + PSP_VPSS_EXT_VIDEO_DECODER_CONFIG,
268                     &tvp5146Params) != IOM_COMPLETED ) {
269         return;
270     } else {
271         for ( i=0; i < NO_OF_BUFFERS; i++ ) {
272             if ( IOM_COMPLETED == FVID_alloc( ccdcHandle, &ccdcAllocFB[i] ) ) {
273                 if ( IOM_COMPLETED != FVID_queue(ccdcHandle, ccdcAllocFB[i] ) ) {
274                     return;
275                 }
276             }
277         }
278     }
279
280     // Create video channel
281     beinitParams.id = PSP_VPBE_VIDEO_0;
282     beinitParams.params = (PSP_VPBE0sdConfigParams*)&vid0Params;
283     vid0Handle = FVID_create( "/VPBE0", IOM_INOUT, NULL, &beinitParams,
284                             &gpioAttrs);
285     if ( NULL == vid0Handle ) {

```



```

286     return;
287 } else {
288     for ( i=0; i<NO_OF_BUFFERS; i++ ) {
289         if ( IOM_COMPLETED == FVID_alloc( vid0Handle, &vidAllocFB[i] ) ) {
290             if ( IOM_COMPLETED != FVID_queue( vid0Handle, vidAllocFB[i] ) ) {
291                 return;
292             }
293         }
294     }
295 }
296
297 // Create venc channel
298 beinitParams.id = PSP_VPBE_VENC;
299 beinitParams.params = (PSP_VPBEVencConfigParams *)&vencParams;
300 vencHandle = FVID_create( "/VPBE0", IOM_INOUT, NULL, &beinitParams,
301                          &gioAttrs);
302 if ( NULL == vencHandle ) {
303     return;
304 }
305
306 //Allocation memoire et la structure qui contiendra l'image
307 FVID_alloc( ccdcHandle, &FBAddr );
308 FVID_alloc( ccdcHandle, &FBAddrOut );
309
310 //=====BOUCLE ACQUISITION & COPIE & AFFICHAGE DESIMAGES=====
311 // 1)Acquisition
312 for( i = 0; i < 10000; i++ ) {
313
314     // Load image
315     if ( IOM_COMPLETED != FVID_exchange( ccdcHandle, &FBAddr ) ) {
316         return;
317     }
318 }
319
320 v = i%10;
321
322 // Make the Y matrix transposed
323 for(k=0; k<height;k++) // Lines
324     for(j=0; j<width;j++) // Columns
325         m2[k+j*height] = (float) *((unsigned char *)FBAddr->frameBufferPtr + (j*2 + k*2*720)*2
326                                 + 1);
327
328 // integer
329 profiling[5*v] = C64P_getltime();
330 deriche_gl(0.2);
331 profiling[5*v+1] = C64P_getltime();
332 robertson(30);
333 profiling[5*v+2] = C64P_getltime();
334 hough_lines_slow();
335 profiling[5*v+3] = C64P_getltime();
336 print_lines_slow(150);
337 profiling[5*v+4] = C64P_getltime();
338
339 // Print the Y matrix
340 for(k=0; k<height;k++) // Lines
341     for(j=0; j<width;j++) { // Columns
342         *((unsigned char *)FBAddrOut->frameBufferPtr + (j + k*720)*2) = 128;
343         *((unsigned char *)FBAddrOut->frameBufferPtr + (j + k*720)*2 + 1) = (unsigned char)m[j
344             +width*k];
345     }
346
347 LOG_printf( &trace, "    Affichage iteration = %u", i );
348
349 // Print changed image
350 if ( IOM_COMPLETED != FVID_exchange( vid0Handle, &FBAddrOut ) ) {
351     return;
352 }
353 }
354 //=====FIN BOUCLE ACQUISITION & COPIE & AFFICHAGE DESIMAGES
355 //=====
356 FVID_free(vid0Handle, FBAddr);

```

```
356 FVID_free(ccdcHandle, FBAddrOut);
357
358 // Free Memory Buffers
359 for( i=0; i< NO_OF_BUFFERS; i++ ) {
360     FVID_free( ccdcHandle, ccdcAllocFB[i] );
361     FVID_free( vid0Handle, vidAllocFB[i] );
362 }
363
364 // Delete Channels
365 FVID_delete( ccdcHandle );
366 FVID_delete( vid0Handle );
367 FVID_delete( vencHandle );
368
369 return;
370 }
```

10 Appendix C - Algorithm-architecture matched version

Listing 20: Algorithm-architecture matched version

```
1 /**
2  Algorithm-architecture matched version of given image processing chain.
3
4  Authors: Divij Babbar, Kubicka Matej (I4-IMC)
5  Date: 1/6/2012
6  Version: algorithm-architecture matched
7  */
8
9  #include <std.h>
10 #include <gio.h>
11 #include <log.h>
12 #include <math.h>
13
14 #include "psp_vpfe.h"
15 #include "psp_vpbe.h"
16 #include "fvid.h"
17
18 #include "psp_tvp5146_extVidDecoder.h"
19
20 #include <soc.h>
21 #include <cslr_ccdc.h>
22
23 #include <soc.h>
24 #include <cslr_sysctl.h>
25
26
27 //pour logger ce qui se passe avec log.h (voir DSPBIOS)
28 extern LOG_Obj trace; // BIOS LOG object
29
30
31 /* extrait de l'exemple EDMA3 :
32 // 48K L1 SRAM [0x10f04000, 0x10f10000), 0xc000 length
33 // 32K L1 Dcache [0x10f10000, 0x10f18000), 0x8000 length
34 // 128K L2 SRAM [0x10800000, 0x10820000), 0x20000 length
35 // 128M DDR2 [0x80000000, 0x88000000), 0x8000000 length are cacheable
36 */
37
38
39 #define ADDR(in, x,y) in[(x) + (y)*width]
40 #define absM(v) (v<0)?(-v):(v)
41
42 #ifndef PI
43 #define PI 3.14159265
44 #endif
45
46 #define NO_OF_BUFFERS          (2u)
47 #define width 300
48 #define height 200
49
50
51 // Global Variable Defined
52 static PSP_VPSSSurfaceParams *ccdcAllocFB[NO_OF_BUFFERS]={NULL};
53 static PSP_VPSSSurfaceParams *vidAllocFB[NO_OF_BUFFERS] ={NULL};
54
55 static FVID_Handle    ccdcHandle;
56 static FVID_Handle    vid0Handle;
57 static FVID_Handle    vencHandle;
58
59 static PSP_VPFE_TVP5146_ConfigParams tvp5146Params = {
60     TRUE, // enable656Sync
61     PSP_VPFE_TVP5146_FORMAT_COMPOSITE, // format
62     PSP_VPFE_TVP5146_MODE_AUTO        // mode
63 };
64
65 static PSP_VPFE_CcdcConfigParams ccdcParams = {
66     PSP_VPFE_CCDC_YCBCR_8, // dataFlow
67     PSP_VPSS_FRAME_MODE, // ffMode
68     480, // height
```

```

69 720, // width
70 (720 *2), // pitch
71 0, // horzStartPix
72 0, // vertStartPix
73 NULL, // appCallback
74 {
75     PSP_VPFE_TVP5146_Open, // extVD Fxn
76     PSP_VPFE_TVP5146_Close,
77     PSP_VPFE_TVP5146_Control,
78 },
79 0 //segId
80 };
81
82 static PSP_VPBEOsdConfigParams vid0Params = {
83     PSP_VPSS_FRAME_MODE, // ffmode
84     PSP_VPSS_BITS16, // bitsPerPixel
85     //PSP_VPBE_RGB_888, //ajout TG
86     PSP_VPBE_YCbCr422, // colorFormat
87     (720 * (16/8u)), // pitch
88     0, // leftMargin
89     0, // topMargin
90     720, // width
91     480, // height
92     0, // segId
93     PSP_VPBE_ZOOM_IDENTITY, // hScaling
94     PSP_VPBE_ZOOM_IDENTITY, // vScaling
95     PSP_VPBE_EXP_IDENTITY, // hExpansion
96     PSP_VPBE_EXP_IDENTITY, // vExpansion
97     NULL // appCallback
98 };
99
100 static PSP_VPBEEvencConfigParams vencParams = {
101     PSP_VPBE_DISPLAY_NTSC_INTERLACED_COMPOSITE // Display Standard
102 };
103
104 #pragma DATA_SECTION(mint, ".ExtBuffer")
105 float mint[300*200];
106 #pragma DATA_SECTION(mint2, ".ExtBuffer")
107 float mint2[300*200];
108 #pragma DATA_SECTION(h, ".ExtBuffer")
109 short h[360*180];
110
111 unsigned int profiling[50];
112
113 float sinT[360];
114 float cosT[360];
115 float arctanT[4000];
116
117 #pragma DATA_SECTION(l, ".L2Buffer")
118 float l[width];
119
120 void transpose(float *in, float *out, int w, int h)
121 {
122     int n,m,s=0;
123     for(n=0; n<h;n++)
124         for(m=0; m<w;m++)
125             out[m*h+n] = in[s++];
126 }
127
128 void deriche2(float g)
129 {
130     float ig1 = ((1-g)*(1-g));
131     float ig2 = (2*g);
132     float igg = (g*g);
133
134     int i,ii,k;
135     float p1, p2, p3;
136
137     #pragma MUST_ITERATE(width, width);
138     for(i = 0 ; i < width ; i++){ // lines
139         ii=i*height;
140         p3=ig1*(mint2[ii++]);
141         p2=(ig1*mint2[ii++]+ig2*p1[0]);

```

```

142     l[0]=p3; l[1]=p2;
143     #pragma MUST_ITERATE(198, 198,1);
144     for(k=2;k<height;k++){
145
146         p1 = p2;
147         p2 = p3;
148         p3 = (ig1*(mint2[ii++]) + ig2*(p2) - igg*(p1));
149         l[k] = p3;
150     }
151
152     mint2[ii--]=p3;
153     mint2[ii--]=p2;
154     #pragma MUST_ITERATE(198, 198,1);
155     for(k=height-3;k>=0;k--){
156         p1 = p2;
157         p2 = p3;
158         p3 = (ig1*(l[k]) + ig2*(p2) - igg*(p1));
159         mint2[ii--] = (unsigned char)p3;
160     }
161 }
162
163 transpose(mint2, mint, height, width);
164
165 #pragma MUST_ITERATE(height, height);
166 for(i = 0 ; i < height ; i++){ // lines
167     ii=i*width;
168
169     p3=(ig1*mint[ii++]);
170     p2=(ig1*mint[ii++]+ig2*l[0]);
171     l[0]=p3; l[1]=p2;
172     #pragma MUST_ITERATE(298, 298,1);
173     for(k=2;k<width;k++){
174         p1 = p2;
175         p2 = p3;
176         p3 = (ig1*(mint[ii++]) + ig2*(p2) - igg*(p1));
177         l[k] = p3;
178     }
179
180     mint[ii--]=p3;
181     mint[ii--]=p2;
182     #pragma MUST_ITERATE(298, 298,1);
183     for(k=width-3;k>=0;k--){
184         p1 = p2;
185         p2 = p3;
186         p3 = (ig1*(l[k]) + ig2*(p2) - igg*(p1));
187         mint[ii--] = (unsigned char) p3;
188     }
189 }
190 }
191
192 void print_line(float intensity, int x1, int y1, int x2, int y2)
193 {
194     int i; int d = (int)sqrt((x2-x1)*(x2-x1)+(y2-y1)*(y2-y1)); /* line size */
195     float ax=(x2-x1)/(float)d, ay=(y2-y1)/(float)d; /* direction vectors */
196     for(i=0; i<d; i++) /* print pixel */  mint[(x1+(int)(ax*i))+(y1+(int)(ay*i))*width] =
197         intensity;
198 }
199 void print_lines(unsigned int treshold)
200 {
201     int j,jk,k, x[2], y[2], hct;
202     int sx=0, sy=0, ex=sx+width, ey=sy+height, is1_x, is2_y, is3_x, is4_y;
203     float x1, y1, ax, ay, rho;
204
205     #pragma MUST_ITERATE(360, 360, 1);
206     for(k=0; k<360;k++) { rho = k*2-360;
207         #pragma MUST_ITERATE(180, 180);
208         #pragma UNROLL(20);
209         for(jk=0; jk<180;jk++)
210             if(h[jk+180*k] > treshold) {
211                 j=jk-90;
212                 hct = 0;
213                 x1 = rho*cosT[absM(j)]; y1 = ((j<0)?-1:1)*rho*sinT[absM(j)]; /* get a point on line

```

```

214     */
215     ax = y1; ay = -x1; /* get direction vector */
216     is1_x = x1-sx+(sy-y1)*ax/ay; is2_y = y1-sy+(ex-x1)*ay/ax; /* calculate intersections
217     */
218     is3_x = x1-sx+(ey-y1)*ax/ay; is4_y = y1-sy+(sx-x1)*ay/ax;
219     if((is1_x>=sx)&&(is1_x<ex)) { x[hct]=is1_x; y[hct++]=sy; } /* find hits */
220     if((is3_x>=sx)&&(is3_x<ex)) { x[hct]=is3_x; y[hct++]=ey; }
221     if((is2_y>=sy)&&(is2_y<ey)) { x[hct]=ex; y[hct++]=is2_y; }
222     if((is4_y>=sy)&&(is4_y<ey)) { x[hct]=sx; y[hct++]=is4_y; }
223     if(hct==2) print_line(255, x[0], y[0], x[1], y[1]); /* print */
224 }
225
226 void roberts(int treshold) {
227     /* Run-in-vars */ int k,j;
228     int convy, convx; int res;
229     register int p1,p2,p3,p4;
230
231     #pragma MUST_ITERATE(height, height);
232     for(k=0; k<height-1;k++) { // Lines
233         p1 = mint[k*width];
234         p4 = mint[(k+1)*width];
235         #pragma MUST_ITERATE(width, width);
236         for(j=0; j<width-1;j++) { // Columns
237             //ROBERTS
238             p2 = mint[j+1+k*width];
239             p3 = mint[j+1+(k+1)*width];
240             convy = -p1+p4-p2+p3;
241             convx = -p1+p2-p4+p3;
242             p1 = p2; p4 = p3;
243             res = ((absM(convx) + absM(convy)) >= treshold)?255:0;
244             mint2[j+k*width]=res;
245         }
246     }
247 }
248
249 void hough_lines_slow() {
250     /* Run-in-vars */ int k,j, a; float rho;
251
252     for(k=0; k<360*180;k++)
253         h[k] = 0;
254
255     for(k=1; k<height;k++) // Lines
256         for(j=1; j<width;j++) // Columns
257             if(mint2[j + k*width]) for(a=0;a<180;a++) { // Angles
258                 rho = (j*cos(a*PI/180.0) + k*sin(a*PI/180.0));
259                 if(rho) h[a+180*(((int)rho+360)/2)]++;
260             }
261 }
262
263 void start_boucle() {
264     PSP_VPBChannelParams beinitParams;
265     PSP_VPFChannelParams feinitParams;
266     GIO_Attrs gioAttrs = GIO_ATTRS;
267     PSP_VPSSSurfaceParams *FBAddr = NULL;
268     PSP_VPSSSurfaceParams *FBAddrOut = NULL;
269
270     Uint32 j = 0;
271     Uint32 k = 0;
272     Uint32 l = 0;
273
274     int i,v = 0;
275
276     for(i=0;i<359;i++) {
277         sinT[i] = sin(i*PI/180.0);
278         cosT[i] = cos(i*PI/180.0);
279     }
280
281     for(i=-2000;i<2000;i++)
282         arctanT[i+2000] = atan(i/10.0);
283
284     //Init CSL du DMA

```

```

285  edma3init();
286
287  // Create ccdc channel
288  feinitParams.id = PSP_VPFE_CCDC;
289  feinitParams.params = (PSP_VPFECCdcConfigParams*)&ccdcParams;
290  ccdcHandle = FVID_create( "/VPFE0", IOM_INOUT, NULL, &feinitParams,
291                          &gioAttrs);
292  if ( NULL == ccdcHandle) {
293      return;
294  }
295
296  // Configure the TVP5146 video decoder
297  if( FVID_control( ccdcHandle,
298                  VPFE_ExtVD_BASE + PSP_VPSS_EXT_VIDEO_DECODER_CONFIG,
299                  &tvp5146Params) != IOM_COMPLETED ) {
300      return;
301  } else {
302      for ( i=0; i < NO_OF_BUFFERS; i++ ) {
303          if ( IOM_COMPLETED == FVID_alloc( ccdcHandle, &ccdcAllocFB[i] ) ) {
304              if ( IOM_COMPLETED != FVID_queue(ccdcHandle, ccdcAllocFB[i] ) ) {
305                  return;
306              }
307          }
308      }
309  }
310
311  // Create video channel
312  beinitParams.id = PSP_VPBE_VIDEO_0;
313  beinitParams.params = (PSP_VPBE0sdConfigParams*)&vid0Params;
314  vid0Handle = FVID_create( "/VPBE0", IOM_INOUT, NULL, &beinitParams,
315                          &gioAttrs );
316  if ( NULL == vid0Handle ) {
317      return;
318  } else {
319      for ( i=0; i<NO_OF_BUFFERS; i++ ) {
320          if ( IOM_COMPLETED == FVID_alloc( vid0Handle, &vidAllocFB[i] ) ) {
321              if ( IOM_COMPLETED != FVID_queue( vid0Handle, vidAllocFB[i] ) ) {
322                  return;
323              }
324          }
325      }
326  }
327
328  // Create venc channel
329  beinitParams.id = PSP_VPBE_VENC;
330  beinitParams.params = (PSP_VPBEVencConfigParams *)&vencParams;
331  vencHandle = FVID_create( "/VPBE0", IOM_INOUT, NULL, &beinitParams,
332                          &gioAttrs);
333  if ( NULL == vencHandle ) {
334      return;
335  }
336
337  //Allocation memoire et la structure qui contiendra l'image
338  FVID_alloc( ccdcHandle, &FBAddr );
339  FVID_alloc( ccdcHandle, &FBAddrOut );
340
341
342  //=====BOUCLE ACQUISITION & COPIE & AFFICHAGE DESIMAGES=====
343  // 1)Acquisition
344  for( i = 0; i < 1000000; i++ ) {
345
346      // Load image
347      if ( IOM_COMPLETED != FVID_exchange( ccdcHandle, &FBAddr ) ) {
348          return;
349      }
350
351      v = i%10;
352
353      // Make the Y matrix transposed
354      l=0; for(k=0; k<height;k++) // Lines
355          for(j=0; j<width;j++) // Columns
356              mint2[k+j*height] = (float)(*((unsigned char *)FBAddr->frameBufferPtr +
357              (j*2 + k*2*720)*2 + 1));

```

```

358
359 // integer
360 profiling[5*v] = C64P_getltime();
361 deriche2(0.2);
362 profiling[5*v+1] = C64P_getltime();
363 roberts(75);
364 profiling[5*v+2] = C64P_getltime();
365 hough_lines_slow();
366 profiling[5*v+3] = C64P_getltime();
367 print_lines(100);
368 profiling[5*v+4] = C64P_getltime();
369
370
371 // Print the Y matrix
372 l=0; for(k=0; k<height;k++) // Lines
373     for(j=0; j<width;j++) { // Columns
374         *((unsigned char *)FBAddrOut->frameBufferPtr + (j + k*720)*2) = 128;
375         *((unsigned char *)FBAddrOut->frameBufferPtr + (j + k*720)*2 + 1) =
376             (unsigned char)(mint[l]);
377         l++;
378     }
379
380     LOG_printf( &trace, "    Affichage iteration = %u", i );
381
382     // Print changed image
383     if ( IOM_COMPLETED != FVID_exchange( vid0Handle, &FBAddrOut ) ) {
384         return;
385     }
386 }
387 //=====FIN BOUCLE ACQUISITION & COPIE & AFFICHAGE DESIMAGES=====
388 FVID_free(vid0Handle, FBAddr);
389 FVID_free(ccdcHandle, FBAddrOut);
390
391 // Free Memory Buffers
392 for( i=0; i< NO_OF_BUFFERS; i++ ) {
393     FVID_free( ccdcHandle, ccdcAllocFB[i] );
394     FVID_free( vid0Handle, vidAllocFB[i] );
395 }
396
397 // Delete Channels
398 FVID_delete( ccdcHandle );
399 FVID_delete( vid0Handle );
400 FVID_delete( vencHandle );
401
402 return;
403 }

```

11 Appendix D - Optimized version

Listing 21: Optimized version

```
1 /**
2  Optimized version of given image processing chain.
3
4  Authors: Divij Babbar, Kubicka Matej (I4-IMC)
5  Date: 1/6/2012
6  Version: Naive
7
8  */
9
10 #include <std.h>
11 #include <gio.h>
12 #include <log.h>
13 #include <math.h>
14
15 #include "psp_vpfe.h"
16 #include "psp_vpbe.h"
17 #include "fvid.h"
18
19 #include "psp_tvp5146_extVidDecoder.h"
20
21 #include <soc.h>
22 #include <cslr_ccdc.h>
23
24 #include <soc.h>
25 #include <cslr_sysctl.h>
26
27
28 //pour logger ce qui se passe avec log.h (voir DSPBIOS)
29 extern LOG_Obj trace; // BIOS LOG object
30
31
32 /* extrait de l'exemple EDMA3 :
33 // 48K L1 SRAM [0x10f04000, 0x10f10000), 0xc000 length
34 // 32K L1 Dcache [0x10f10000, 0x10f18000), 0x8000 length
35 // 128K L2 SRAM [0x10800000, 0x10820000), 0x20000 length
36 // 128M DDR2 [0x80000000, 0x88000000), 0x80000000 length are cacheable
37 */
38
39
40 #define ADDR(in, x,y) in[(x) + (y)*width]
41 #define absM(v) (v<0)?(-v):(v)
42
43 #ifndef PI
44 #define PI 3.14159265
45 #endif
46
47 #define NO_OF_BUFFERS (2u)
48 #define width 300
49 #define height 200
50
51
52 // Global Variable Defined
53 static PSP_VPSSSurfaceParams *ccdcAllocFB[NO_OF_BUFFERS]={NULL};
54 static PSP_VPSSSurfaceParams *vidAllocFB[NO_OF_BUFFERS] ={NULL};
55
56 static FVID_Handle ccdcHandle;
57 static FVID_Handle vid0Handle;
58 static FVID_Handle vencHandle;
59
60 static PSP_VPFE_TVP5146_ConfigParams tvp5146Params = {
61     TRUE, // enable656Sync
62     PSP_VPFE_TVP5146_FORMAT_COMPOSITE, // format
63     PSP_VPFE_TVP5146_MODE_AUTO // mode
64 };
65
66 static PSP_VPFE_CcdcConfigParams ccdcParams = {
67     PSP_VPFE_CCDC_YCBCR_8, // dataFlow
68     PSP_VPSS_FRAME_MODE, // ffMode
```

```

69 480, // height
70 720, // width
71 (720 *2), // pitch
72 0, // horzStartPix
73 0, // vertStartPix
74 NULL, // appCallback
75 {
76 PSP_VPFE_TVP5146_Open, // extVD Fxn
77 PSP_VPFE_TVP5146_Close,
78 PSP_VPFE_TVP5146_Control,
79 },
80 0 //segId
81 };
82
83 static PSP_VPBEosdConfigParams vid0Params = {
84 PSP_VPSS_FRAME_MODE, // ffmode
85 PSP_VPSS_BITS16, // bitsPerPixel
86 //PSP_VPBE_RGB_888, //ajout TG
87 PSP_VPBE_YCbCr422, // colorFormat
88 (720 * (16/8u)), // pitch
89 0, // leftMargin
90 0, // topMargin
91 720, // width
92 480, // height
93 0, // segId
94 PSP_VPBE_ZOOM_IDENTITY, // hScaling
95 PSP_VPBE_ZOOM_IDENTITY, // vScaling
96 PSP_VPBE_EXP_IDENTITY, // hExpansion
97 PSP_VPBE_EXP_IDENTITY, // vExpansion
98 NULL // appCallback
99 };
100
101 static PSP_VPBEvencConfigParams vencParams = {
102 PSP_VPBE_DISPLAY_NTSC_INTERLACED_COMPOSITE // Display Standard
103 };
104
105 #pragma DATA_SECTION(mint, ".L2Buffer")
106 unsigned char mint[300*200];
107 #pragma DATA_SECTION(mint2, ".ExtBuffer")
108 unsigned char mint2[300*200];
109 #pragma DATA_SECTION(h, ".ExtBuffer")
110 short h[360*180];
111
112 unsigned int profiling[50];
113
114 float sinT[360];
115 float cosT[360];
116 float arctanT[4000];
117
118 #pragma DATA_SECTION(l, ".L2Buffer")
119 int l[width];
120
121 void transpose(unsigned char *in, unsigned char *out, int w, int h)
122 {
123 int n,m,s=0;
124 for(n=0; n<h;n++)
125 for(m=0; m<w;m++)
126 out[m*h+n] = in[s++];
127 }
128
129 void deriche2(float g)
130 {
131 int ig1 = ((1-g)*(1-g))*1024;
132 int ig2 = (2*g)*1024;
133 int igg = (g*g)*1024;
134
135 int i,ii,k;
136 int p1, p2, p3;
137
138 #pragma MUST_ITERATE(width, width);
139 for(i = 0 ; i < width ; i++){ // lines
140 ii=i*height;
141 p3=ig1*(mint2[ii+])>>10;

```

```

142     p2=(ig1*mint2[ii++]+ig2*l[0])>>10;
143     l[0]=p3; l[1]=p2;
144     #pragma MUST_ITERATE(198, 198,1);
145     for(k=2;k<height;k++){
146         p1 = p2;
147         p2 = p3;
148         p3 = (ig1*(mint2[ii++]) + ig2*(p2) - igg*(p1))>>10;
149         l[k] = p3;
150     }
151
152     mint2[ii--]=p3;
153     mint2[ii--]=p2;
154     #pragma MUST_ITERATE(198, 198,1);
155     for(k=height-3;k>=0;k--){
156         p1 = p2;
157         p2 = p3;
158         p3 = (ig1*(l[k]) + ig2*(p2) - igg*(p1))>>10;
159         mint2[ii--] = (unsigned char)p3;
160     }
161 }
162
163 transpose(mint2, mint, height, width);
164
165 #pragma MUST_ITERATE(height, height);
166 for(i = 0 ; i < height ; i++){ // lines
167     ii=i*width;
168     p3=(ig1*mint[ii++])>>10;
169     p2=(ig1*mint[ii++]+ig2*l[0])>>10;
170     l[0]=p3; l[1]=p2;
171     #pragma MUST_ITERATE(298, 298,1);
172     for(k=2;k<width;k++){
173         p1 = p2;
174         p2 = p3;
175         p3 = (ig1*(mint[ii++]) + ig2*(p2) - igg*(p1))>>10;
176         l[k] = p3;
177     }
178
179     mint[ii--]=p3;
180     mint[ii--]=p2;
181     #pragma MUST_ITERATE(298, 298,1);
182     for(k=width-3;k>=0;k--){
183         p1 = p2;
184         p2 = p3;
185         p3 = (ig1*(l[k]) + ig2*(p2) - igg*(p1))>>10;
186         mint[ii--] = (unsigned char) p3;
187     }
188 }
189 }
190
191 void print_line(float intensity, int x1, int y1, int x2, int y2)
192 {
193     int i; int d = (int)sqrt((x2-x1)*(x2-x1)+(y2-y1)*(y2-y1)); /* line size */
194     float ax=(x2-x1)/(float)d, ay=(y2-y1)/(float)d; /* direction vectors */
195     for(i=0; i<d; i++) /* print pixel */  mint[(x1+(int)(ax*i))+(y1+(int)(ay*i))*width] =
196         intensity;
197 }
198 void print_lines(unsigned int treshold)
199 {
200     int j,jk,k, x[2], y[2], hct;
201     int sx=0, sy=0, ex=sx+width, ey=sy+height, is1_x, is2_y, is3_x, is4_y;
202     float x1, y1, ax, ay, rho;
203
204     #pragma MUST_ITERATE(360, 360, 1);
205     for(k=0; k<360;k++){ rho = k*2-360;
206         #pragma MUST_ITERATE(180, 180);
207         #pragma UNROLL(20);
208         for(jk=0; jk<180;jk++){
209             if(h[jk+180*k] > treshold) {
210                 j=jk-90; hct = 0;
211                 x1 = rho*cosT[absM(j)]; y1 = ((j<0)?-1:1)*rho*sinT[absM(j)]; /* get a point on line
212                 */
213                 ax = y1; ay = -x1; /* get direction vector */

```

```

213     is1_x = x1-sx+(sy-y1)*ax/ay; is2_y = y1-sy+(ex-x1)*ay/ax; /* calculate intersections
214     */
215     is3_x = x1-sx+(ey-y1)*ax/ay; is4_y = y1-sy+(sx-x1)*ay/ax;
216     if((is1_x>=sx)&&(is1_x<ex)) { x[hct]=is1_x; y[hct++]=sy; } /* find hits */
217     if((is3_x>=sx)&&(is3_x<ex)) { x[hct]=is3_x; y[hct++]=ey; }
218     if((is2_y>=sy)&&(is2_y<ey)) { x[hct]=ex; y[hct++]=is2_y; }
219     if((is4_y>=sy)&&(is4_y<ey)) { x[hct]=sx; y[hct++]=is4_y; }
220     if(hct==2) print_line(255, x[0], y[0], x[1], y[1]); /* print */
221 }
222 }
223
224 void hough_lines(int treshold) {
225     /* Run-in-vars */ int k,j; float rho;
226     int convy, convx; float conv ; int res, a;
227     register int p1,p2,p3,p4;
228     int v;
229
230     #pragma MUST_ITERATE(360*180, 360*180, 1);
231     for(k=0; k<360*180;k++)
232         h[k] = 0;
233
234     #pragma MUST_ITERATE(height, height);
235     for(k=0; k<height-1;k++) { // Lines
236         p1 = mint[k*width];
237         p4 = mint[(k+1)*width];
238         #pragma MUST_ITERATE(width, width);
239         for(j=0; j<width-1;j++) { // Columns
240
241             //ROBERTS
242             p2 = mint[j+1+k*width];
243             p3 = mint[j+1+(k+1)*width];
244             convy = -p1+p4-p2+p3;
245             convx = -p1+p2-p4+p3;
246             p1 = p2; p4 = p3;
247             res = ((absM(convx) + absM(convy)) >= treshold)?255:0;
248
249             if(res!=0) {
250                 conv = (convy/(float)convx);
251                 if(conv >= 200) a = 90;
252                 else if(conv < -200) a=-90;
253                 else a = (int) (arctanT[(int) (conv*10)+2000]*57.295);
254                 rho = (j*cosT[absM(a)] + ((a<0)?-1:1)*k*sinT[absM(a)]);
255                 if(rho) h[a+90+180*((int)rho+360)>>1]++;
256             }
257         }
258     }
259 }
260
261 void start_boucle() {
262     PSP_VPBChannelParams beinitParams;
263     PSP_VPFEChannelParams feinitParams;
264     GIO_Attrs gioAttrs = GIO_ATTRS;
265     PSP_VPSSSurfaceParams *FBAddr = NULL;
266     PSP_VPSSSurfaceParams *FBAddrOut = NULL;
267
268     Uint32 j = 0;
269     Uint32 k = 0;
270     Uint32 l = 0;
271
272
273     int i,v = 0;
274
275     for(i=0;i<359;i++) {
276         sinT[i] = sin(i*PI/180.0);
277         cosT[i] = cos(i*PI/180.0);
278     }
279
280     for(i=-2000;i<2000;i++)
281         arctanT[i+2000] = atan(i/10.0);
282
283     //Init CSL du DMA
284     edma3init();

```

```

285
286 // Create ccdc channel
287 feinitParams.id = PSP_VPFE_CCDC;
288 feinitParams.params = (PSP_VPFEccdcConfigParams*)&ccdcParams;
289 ccdcHandle = FVID_create( "/VPFE0", IOM_INOUT, NULL, &feinitParams,
290                          &gioAttrs);
291 if ( NULL == ccdcHandle) {
292     return;
293 }
294
295 // Configure the TVP5146 video decoder
296 if( FVID_control( ccdcHandle,
297                 VPFE_ExtVD_BASE + PSP_VPSS_EXT_VIDEO_DECODER_CONFIG,
298                 &tpv5146Params) != IOM_COMPLETED ) {
299     return;
300 } else {
301     for ( i=0; i < NO_OF_BUFFERS; i++ ) {
302         if ( IOM_COMPLETED == FVID_alloc( ccdcHandle, &ccdcAllocFB[i] ) ) {
303             if ( IOM_COMPLETED != FVID_queue(ccdcHandle, ccdcAllocFB[i] ) ) {
304                 return;
305             }
306         }
307     }
308 }
309
310 // Create video channel
311 beinitParams.id = PSP_VPBE_VIDEO_0;
312 beinitParams.params = (PSP_VPBE0sdConfigParams*)&vid0Params;
313 vid0Handle = FVID_create( "/VPBE0", IOM_INOUT, NULL, &beinitParams,
314                          &gioAttrs );
315 if ( NULL == vid0Handle ) {
316     return;
317 } else {
318     for ( i=0; i<NO_OF_BUFFERS; i++ ) {
319         if ( IOM_COMPLETED == FVID_alloc( vid0Handle, &vidAllocFB[i] ) ) {
320             if ( IOM_COMPLETED != FVID_queue( vid0Handle, vidAllocFB[i] ) ) {
321                 return;
322             }
323         }
324     }
325 }
326
327 // Create venc channel
328 beinitParams.id = PSP_VPBE_VENC;
329 beinitParams.params = (PSP_VPBEVencConfigParams *)&vencParams;
330 vencHandle = FVID_create( "/VPBE0", IOM_INOUT, NULL, &beinitParams,
331                          &gioAttrs);
332 if ( NULL == vencHandle ) {
333     return;
334 }
335
336 //Allocation memoire et la structure qui contiendra l'image
337 FVID_alloc( ccdcHandle, &FBAddr );
338 FVID_alloc( ccdcHandle, &FBAddrOut );
339
340
341 //=====BOUCLE ACQUISITION & COPIE & AFFICHAGE DESIMAGES=====
342 // 1)Acquisition
343 for( i = 0; i < 1000000; i++ ) {
344
345     // Load image
346     if ( IOM_COMPLETED != FVID_exchange( ccdcHandle, &FBAddr ) ) {
347         return;
348     }
349
350     v = i%10;
351
352     // Make the Y matrix transposed
353     l=0; for(k=0; k<height;k++) // Lines
354         for(j=0; j<width;j++) // Columns
355             mint2[k+j*height] = (unsigned char)((unsigned char *)FBAddr->frameBufferPtr
356             + (j*2 + k*2*720)*2 + 1));
357

```

```

358 // integer
359 profiling[5*v] = C64P_gettime();
360 deriche2(0.2);
361 profiling[5*v+1] = C64P_gettime();
362 profiling[5*v+2] = C64P_gettime();
363 hough_lines(10);
364 profiling[5*v+3] = C64P_gettime();
365 print_lines(20);
366 profiling[5*v+4] = C64P_gettime();
367
368
369 // Print the Y matrix
370 l=0; for(k=0; k<height;k++) // Lines
371     for(j=0; j<width;j++) { // Columns
372         *((unsigned char *)FBAddrOut->frameBufferPtr + (j + k*720)*2) = 128;
373         *((unsigned char *)FBAddrOut->frameBufferPtr + (j + k*720)*2 + 1) =
374             (unsigned char)(mint[l]);
375         l++;
376     }
377
378 LOG_printf( &trace, "    Affichage iteration = %u", i );
379
380 // Print changed image
381 if ( IOM_COMPLETED != FVID_exchange( vid0Handle, &FBAddrOut ) ) {
382     return;
383 }
384 }
385 //=====FIN BOUCLE ACQUISITION & COPIE & AFFICHAGE DESIMAGES=====
386 FVID_free(vid0Handle, FBAddr);
387 FVID_free(ccdcHandle, FBAddrOut);
388
389 // Free Memory Buffers
390 for( i=0; i< NO_OF_BUFFERS; i++ ) {
391     FVID_free( ccdcHandle, ccdcAllocFB[i] );
392     FVID_free( vid0Handle, vidAllocFB[i] );
393 }
394
395 // Delete Channels
396 FVID_delete( ccdcHandle );
397 FVID_delete( vid0Handle );
398 FVID_delete( vencHandle );
399
400 return;
401 }

```
