



SIGNAL & IMAGE PROCESSING LAB

3D Object Reconstruction using DaVinci DSP

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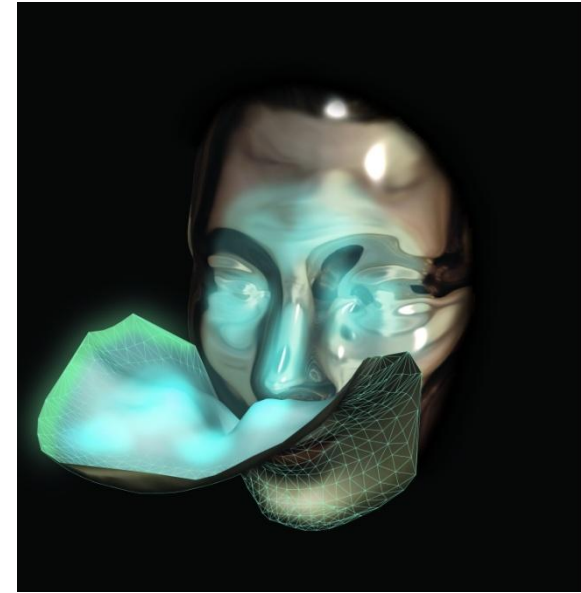
In association with GIP

Outline

- Background
- Project goal
- 3D reconstruction
- DaVinci platform
- Adapting the algorithm to the DaVinci
- Results
- Conclusion

Background – 3D Scanners

- A **3D scanner** acquires an image in which every pixel has 3 coordinates
- Used for:
 - Biometric recognition (face)
 - Medical uses
 - Comparison of 3D surfaces
 - Architecture and civil engineering
 - Virtual reality
 - And more...



Background - 3D Scanning

- Passive scanning
- Active scanning



Laser scanning

3D digital Escan



Time of light scanning

Leica ScanStation 2



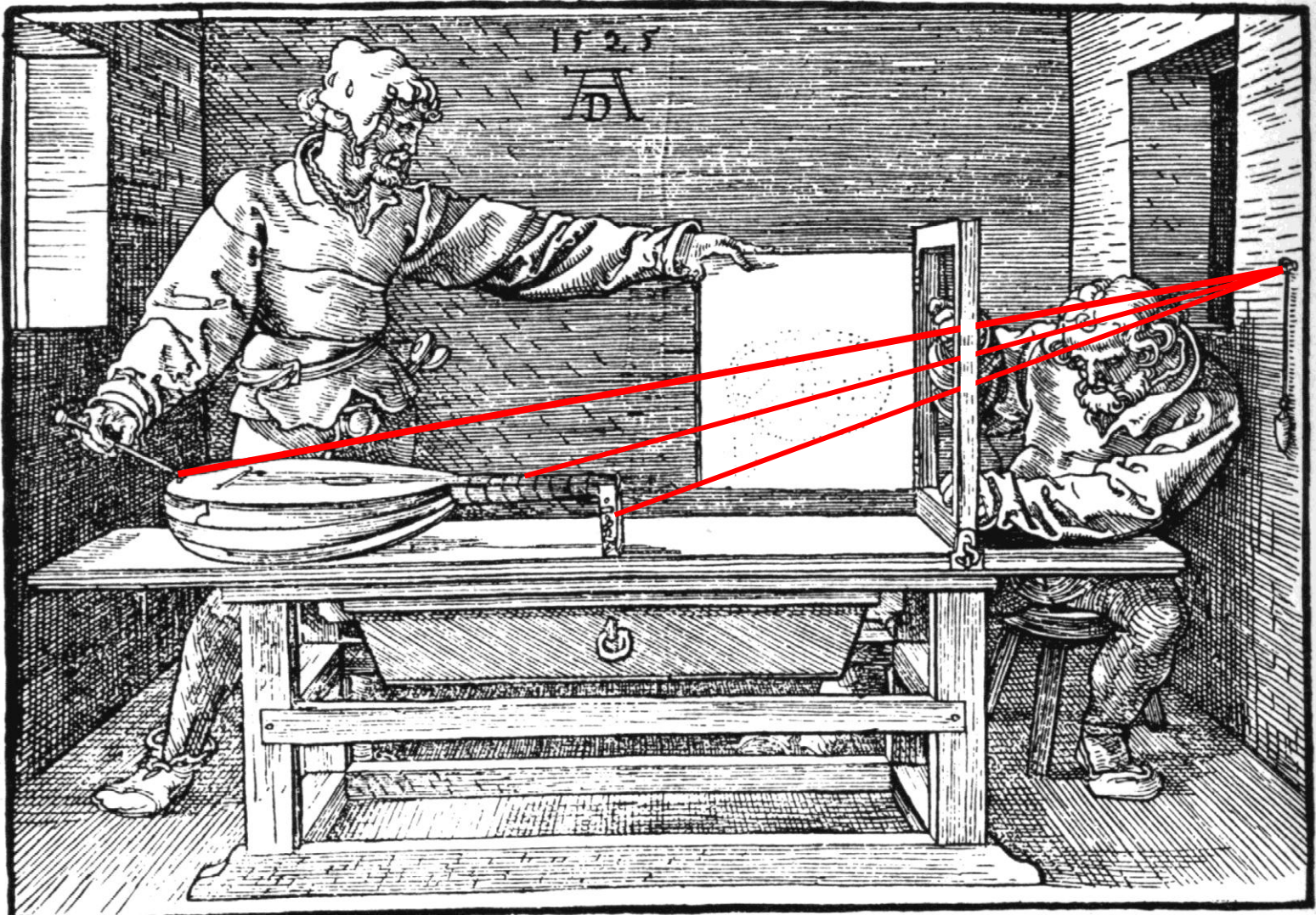
Structured light scanning

Cognitens optigo200

Project Goal

- Realtime implementation of 3D object reconstruction using the DaVinci platform
 - Structured light scanning
 - Using one standard camera and one standard projector
 - Low cost solution

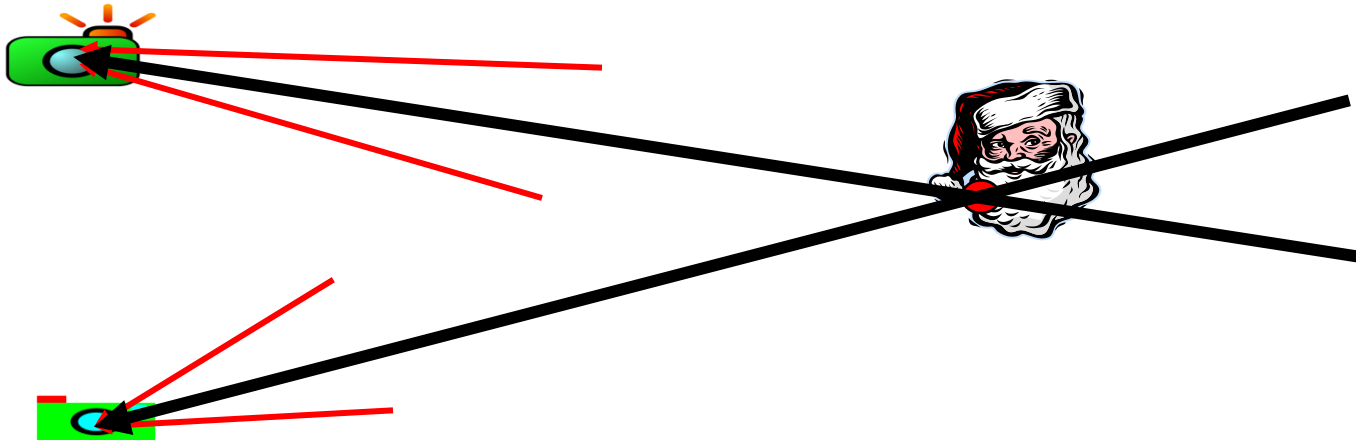
Perspective projection



The Draughtsman of the Lute, Albrecht Dürer

Image Reconstruction

- Given 2 cameras
- If we identify a point in both of them we can know its coordinates (if the lines that goes into the camera are not parallel)



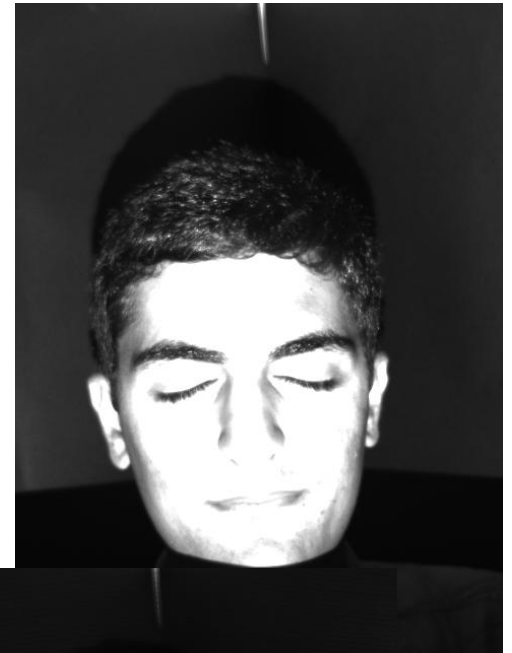
structured light method

- Using ‘active’ projector and a camera
- Projector cast light code on the object
- From the projected plane of the projector and the captured images in the camera we can make the reconstruction



Structured Light Method

scanned pictures



full darkness
image



Bit 3

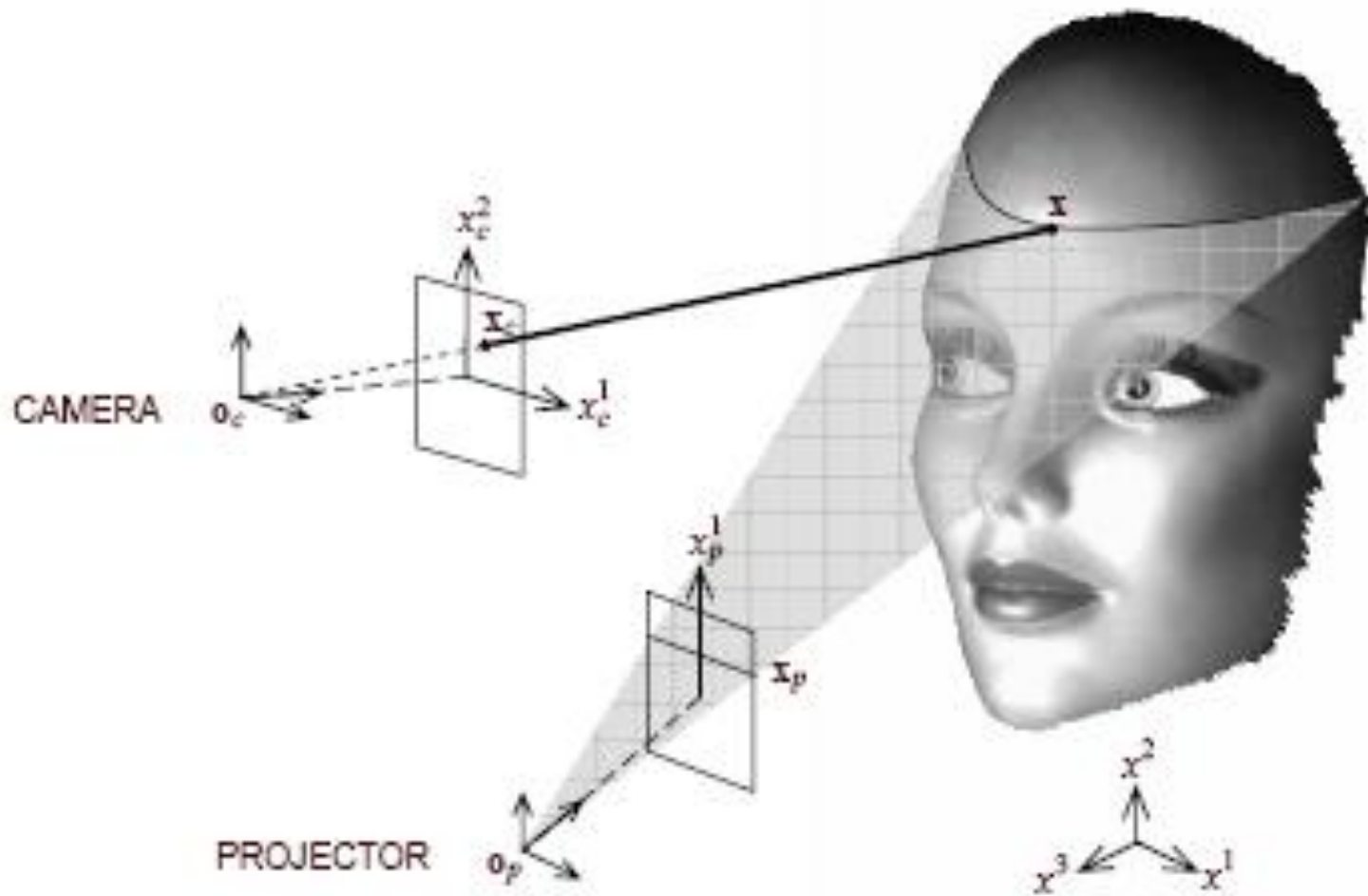


bit 6



bit 9

full
illumination
image



Reconstruction

PPM world to camera

$$X_c = C_c X_w \quad C_c = \alpha \begin{bmatrix} f_x & kf_y & x_c^0 \\ 0 & f_y & y_c^0 \\ 0 & 0 & 1 \end{bmatrix} [R_c \quad t_c].$$

PPM world to projector

$$X_p = C_p X_w \quad C_p = \alpha \begin{bmatrix} f_p & 0 & x_p^0 \\ 0 & 0 & 1 \end{bmatrix} [R_p \quad t_p].$$

Reconstruction 2

- We have $T : X_w \rightarrow (X_c, X_p)$
- But wants: $T^{-1} : (X_c, X_p) \rightarrow X_w$
- We can get that:

$$x_w = -R^{-1}s$$

- When:
- $[R, s] = \begin{bmatrix} x_c c_3 - c_1 \\ y_c c_3 - c_2 \\ x_p p_2 - p_1 \end{bmatrix}$

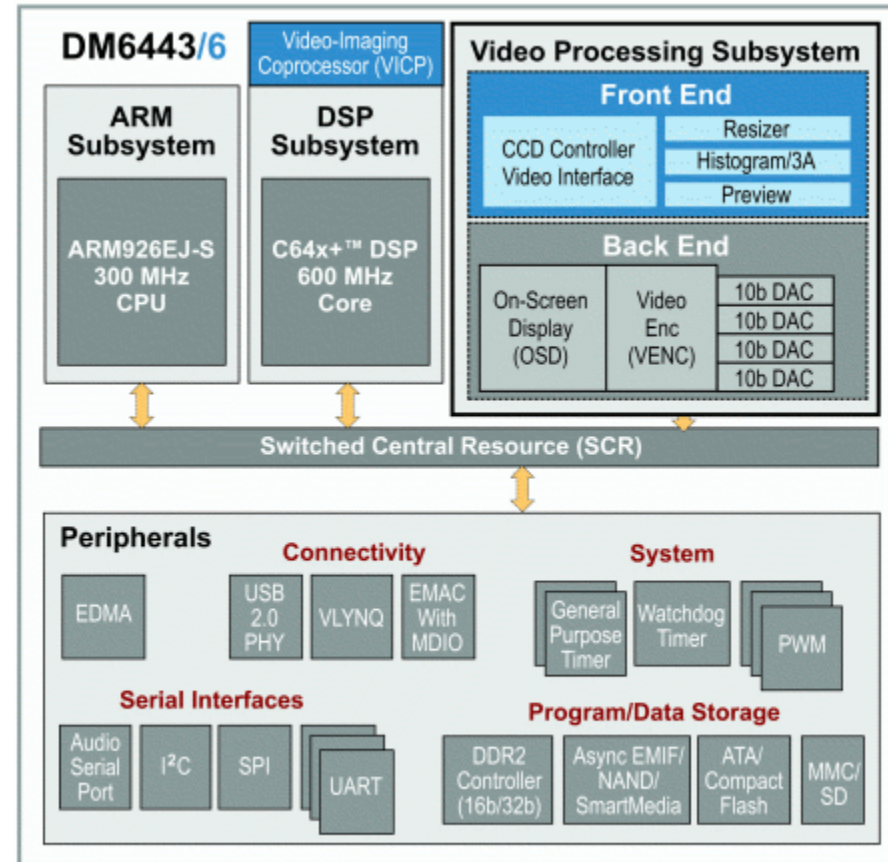
Previous implementation

- FireWire black and white CCD camera
- computer-controlled DLP projector
- 10 binary stripe images and additional full dark and full illumination images
- Code implemented in C language on PC
- 3 3D images per second
- Reconstruction using highly optimized Pentium IV code takes 280ms

DaVinci platform

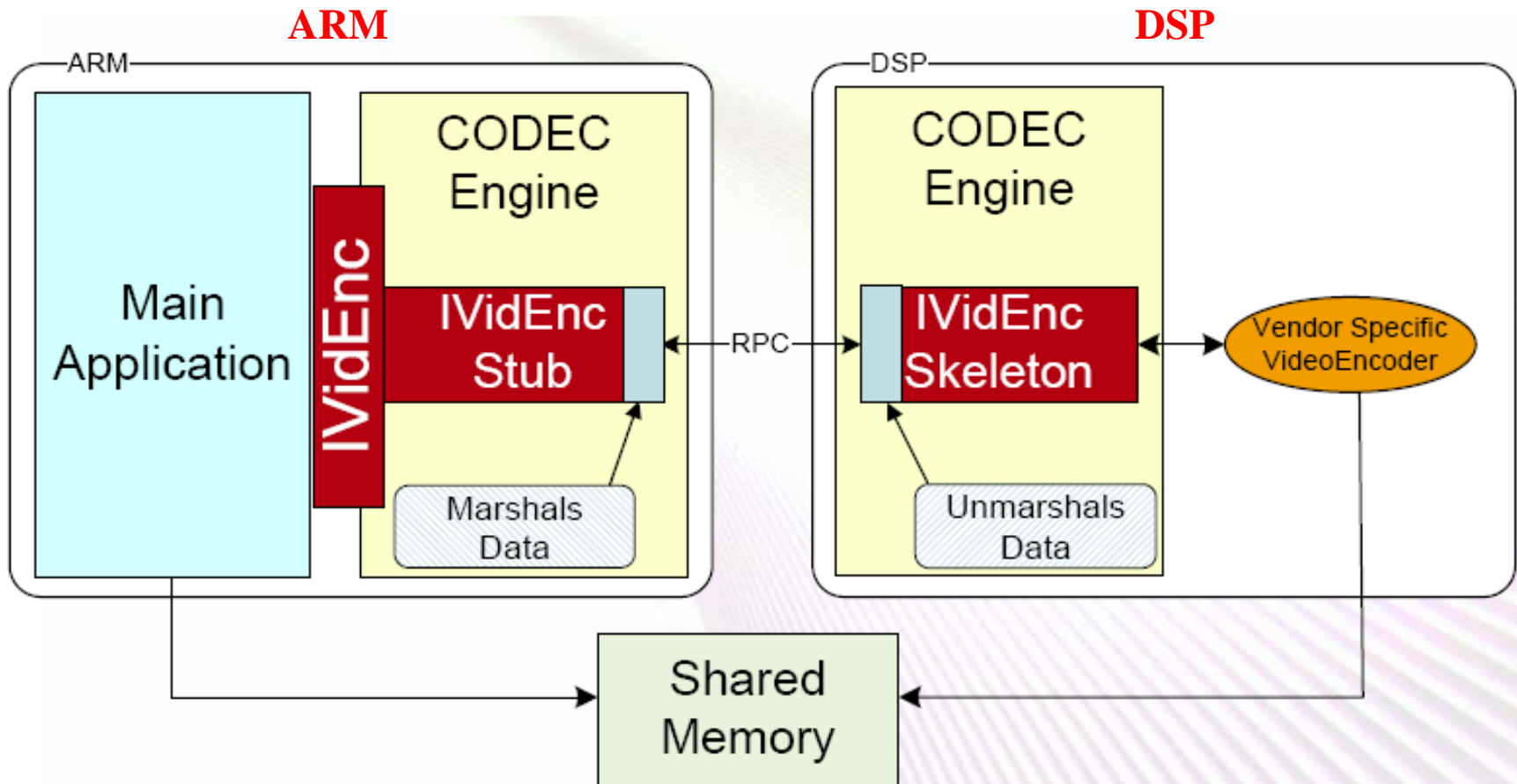


- **2 Cores:**
 - ARM9
 - C64x+ DSP
- **Memory**
 - On-Chip L1/SRAM -112KB DSP, 40 KB ARM
 - On-Chip L2/SRAM – 64KB DSP
- **TI's solution for Video processing.**
- contains large set of compatible software



■ = DM6446

DaVinci xDM

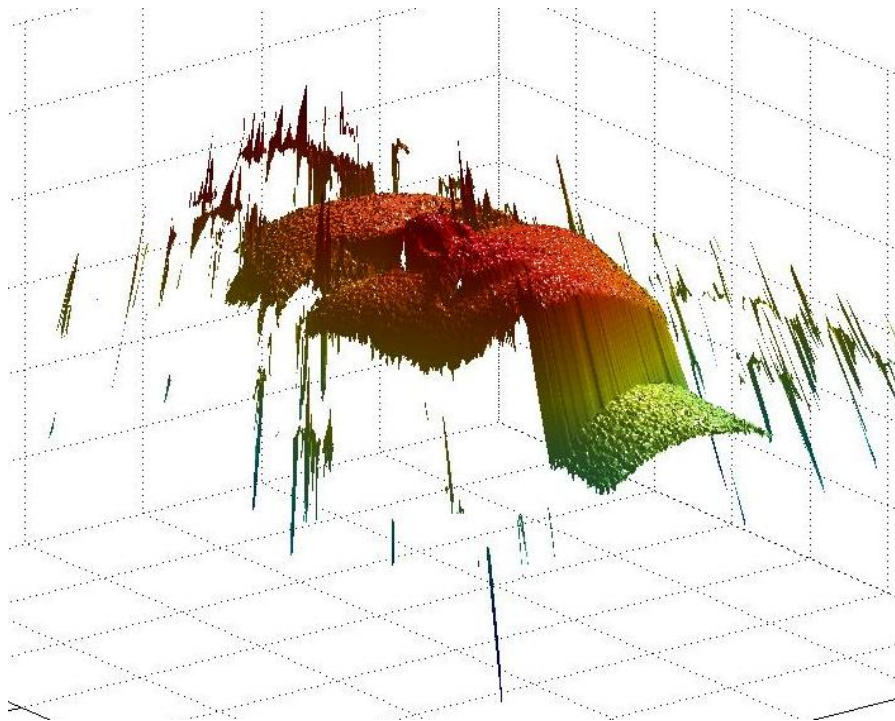


Adapting to the DaVinci

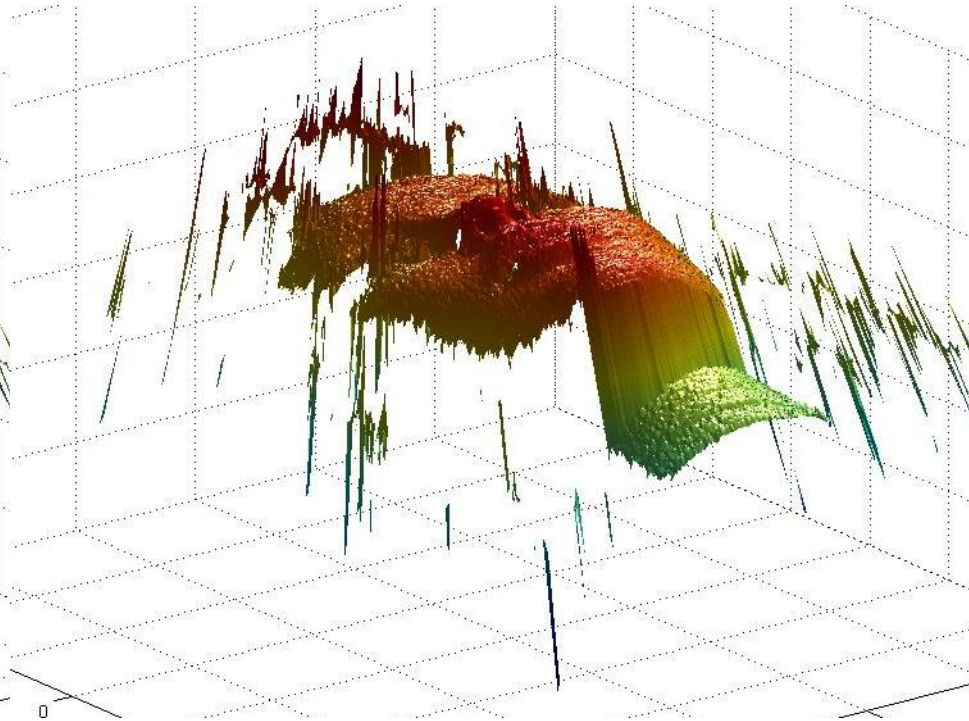
- Floating point to fixed point
 - Changing output format to homogenous coordinates
 - Fine-tuned scaling of all the values
 - Changing filter sizes for efficient division
 - Changing functions to look-up-tables
 -
- 1 core to 2 cores
 - Using TI's xDM standard and VISA interface for inter-core communication
 - ARM connects to peripherals and feeds the DSP
 - DSP makes the reconstruction and output the results back to the ARM

Reconstruction Results

Fixed point version



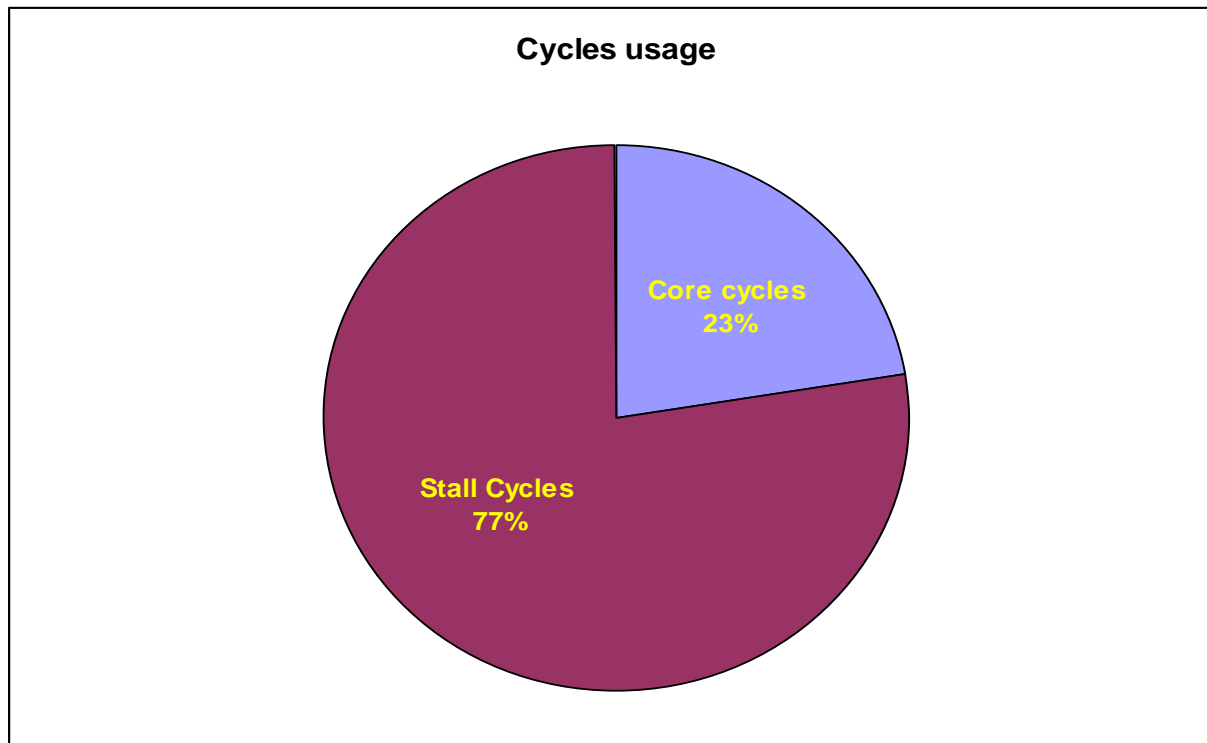
floating point version



RMS = 0.0271

Non-optimized Time Performance

- Reconstruction takes 2.5 sec



What's next

- Adding the camera and the projector and controlling them using the ARM
 - In progress
- Using the DMA controller for more efficient use of memory
- More code optimization
- Adding the calibration part

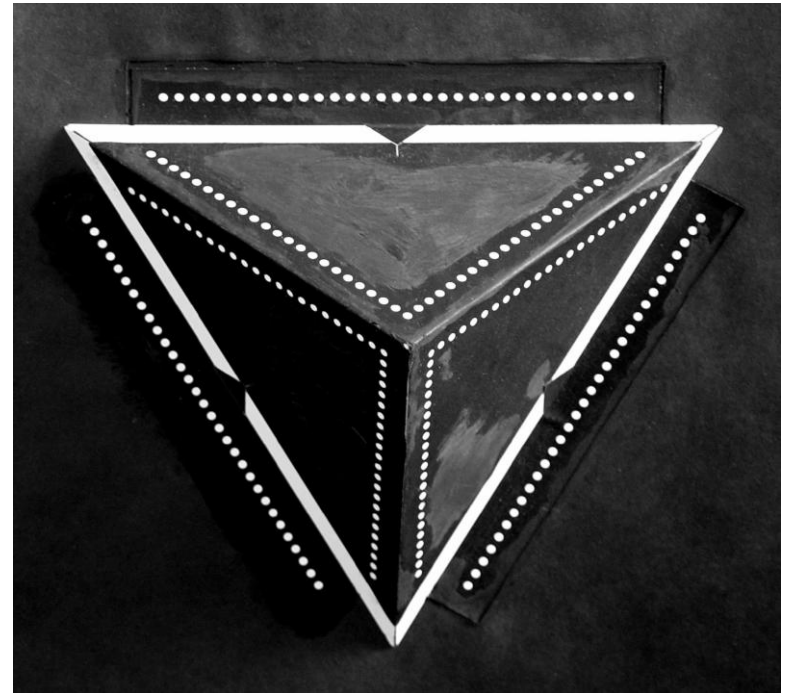
Thank you

Active Stereo techniques

- Gray level multiplexed
- Color multiplexed
- Space multiplexed
- Time multiplexed (we use this)

Calibration

- Known world coordinates
- Building C_c and C_p
- The calibration object:



Calibration 2

- Solving $(\mathbf{x}_c)_k = \mathbf{C}_c(\mathbf{x}_w)_k$
- Solving $(\mathbf{x}_p)_k = \mathbf{C}_p(\mathbf{x}_w)_k,$
- No accurate solution due to finite precision
- Instead we will solve:

$$\mathbf{C}_c = \operatorname{argmin} \sum_{k=1}^N \|\mathbf{C}_c(\mathbf{x}_w)_k - (\mathbf{x}_c)_k\|_2^2 \quad \text{s.t.} \quad \mathbf{C}_c \in \text{PPM}$$

$$\mathbf{C}_p = \operatorname{argmin} \sum_{k=1}^N \|\mathbf{C}_p(\mathbf{x}_w)_k - (\mathbf{x}_p)_k\|_2^2 \quad \text{s.t.} \quad \mathbf{C}_p \in \text{PPM}.$$

- But actually we are interested in \mathbf{T}^{-1} error:

$$\mathbf{T} = \operatorname{argmin} \sum_{k=1}^N \|\mathbf{T}^{-1}(\mathbf{x}_c, \mathbf{x}_p)_k - (\mathbf{x}_w)_k\|_2^2 \quad \text{s.t.} \quad \mathbf{C}_c, \mathbf{C}_p \in \text{textPPM}.$$

Technical details

	C64x+	ARM9
Peak MMACS	4752	
Freq	594	297
Memory on chip (KB)	16 (ROM) 40 (L1/SRAM)	112 (L1/SRAM) 64 (L2/SRAM)