

## Through connectivity in applied computer systems – ADMOS and MARWIN projects

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# Through Connectivity in Applied Computer Systems – ADMOS and MARWIN Projects

#### Marcos Rodrigues





# The GMPR 3D scanning technologies 3D with single image



Each light plane is uniquely detected by original algorithms



#### The MARWIN Project FP7 Research for the Benefit of SMEs



Sheffield Hallam Universit



#### MARWIN: SHU work on various tasks Marcos Rodrigues, Mariza Kormann

3D Scanner Development Registration and fusion of 3D models Translating 3D welding sequence from CAD to scanned model Dissemination



GMPR scanner design A beam splitter allows for visible and nearinfrared cameras to be fitted







The prototype shown uses a MicroVision PicoP laser projector and an IDS CMOS camera (1280x1024)





#### Full design integrated into a robotic arm The actual robotic cell







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#### Scanning a part

Marwin Project - Computer vision based welding robot





#### Scanning a part with the robot As the robot moves the part is assembled





#### Automatic registration with CAD models Scanned patches are fused and registered



Registration goal is to estimate R,t:

$$F(\mathbf{R}, \mathbf{t}) = \sum_{i=1}^{m} \sum_{j=1}^{Ni} p_{i,j} d^{2}(\mathbf{R}p_{i,j} + \mathbf{t}, S_{k}) + \sum_{k=1}^{n} \sum_{l=1}^{Nk} q_{k,l} d^{2}(\mathbf{R}^{T} p_{k,l}^{'} - \mathbf{R}^{T} \mathbf{t}, S_{i})$$



Least squares minimisation:

$$f(\mathbf{R}, \mathbf{t}) = \frac{1}{N} \sum_{i=1}^{N} ||\mathbf{R}\mathbf{p}_i + \mathbf{t}, \mathbf{q}_i||^2$$



## Welding sequence Translated from CAD to scanned models







#### The ADMOS Project Gender classification and age estimation





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Privacy Regulations Modelling and System Design Hardware and Electronics Design Client Side Software Development: tracking, gender and age estimation Dissemination



#### Hardware and Electronics Design Optics and lighting

Optical lens and filters to ensure performance within various illuminating conditions





### Client Side Software Development Firmware and control s/w development

## Real time processing:

- 1. face detection and tracking
- 2. eye tracking
- 3. other feature tracking (mouth, nose)
- 4. cropping the various face-ROI
- 5. gender classification
- 6. age estimation
- 7. save statistical info to an xml file
- 8. transmit to server at periodic intervals





#### Binary patterns LBP, CT and MCT



MCT is similar to CT, except that it uses the average intensity of the kernel window as the intensity of the centre pixel.



#### Applying binary patterns to face images Visualizing the differences on images







LBP 3×3



Modified Census 3×3



Census 5×5



Census 3×3



Modified Census 5×5



### LBP processing Features are defined by the histogram





#### ROI sensitivity analysis

Male subjects tend to be classified with higher accuracy.

This agrees with all results reported in the literature.

No explanation for this behaviour is offered at this stage.

 TABLE I

 COMPARATIVE ANALYSIS OF IMAGE REGIONS

Image ROI	Gender	Classification results
ROI1	Male Female	92% 79%
ROI2	Male	83% 83%
ROI3	Male Female	88% 88%
ROI4	Male Female	88% 71%
ROI5	Male Female	$<\!$



#### Comparative analysis of binary patterns DCT-Discrete Cosine Transform

Decomposes a signal and defines it as a sum of cosines at different frequencies

$$y(k) = w(k) \sum_{n=1}^{N} z(n) \cos(\frac{\pi(2n-1)(k-1)}{2N})$$

$$k = 1, 2, \dots N$$

$$w(k) = \begin{cases} 1/\sqrt{N} & \text{ for } k = 1, \\ \sqrt{2/N} & \text{ for } 2 \le k \le N. \end{cases}$$

The length of coefficients y is the same size as the original signal z.



#### Comparative analysis of binary patterns DWT-Discrete Wavelet Transform

$$x(n) h(n) = \sum_{k=-\infty}^{\infty} x(k) h(n-k)$$

$$y(n) = \sum_{k=-\infty}^{\infty} h(k) x(2n-k)$$

$$y_{\text{high}} = \sum_{n} x(n) g(2k-n)$$

$$y_{\text{low}} = \sum_{n} x(n) h(2k-n)$$

$$x(n) = \sum_{k=-\infty}^{\infty} (y_{\text{high}}(k).g(-n+2k)) (y_{\text{low}}(k).h(-n+2k))$$



# Comparative analysis of binary patterns

Raw histograms Transformed histograms by DCT Transformed histograms by DWT















#### Tested on public databases FEI, AT&T, Sheffield-UMIST, and color FERET





### Classification results Average for 4 Regions of Interest

	LBP	СТ	MCT	LBP  CT	LBP  MCT
FEI Database					
LBP&Census	87.9	87.6	79.0	86.3	85.2
DWT	85.2	84.4	79.0	86.3	84.1
DCT	86.8	89.8	81.2	87.4	85.7
AT&T					
LBP&Census	66.4	65.3	50.0	77.9	50.0
DWT	59.1	50.0	50.0	69.6	50.0
DCT	84.2	87.0	90.1	83.2	80.3
Sheffield-UMIST					
LBP&Census	81.6	82.8	67.8	78.7	68.4
DWT	77.2	78.0	73.2	75.8	83.6
DCT	84.6	84.3	83.0	86.5	85.2
Color FERET					
LBP&Census	69.5	68.1	70.3	67.9	70.6
DWT	71.5	72.0	67.1	72.4	69.6
DCT	71.9	68.8	70.9	73.1	73.6



#### Real time and privacy requirements Define and track anonymous tags





#### Conclusions

LBP + Eigenvector decomposition: top half of the face most significant

<u>Binary patterns + SVM</u>: LBP is slightly superior to CT/MCT <u>Binary patterns + DCT + SVM</u>: CT is clearly the superior technique Also, bias towards male subjects is removed CT has the smallest standard deviation of all techniques <u>Real-time performance</u>: enabled by multiple threads using multi-level queues Testing in shopping malls in Hungary and UK in June and July 2015

