## Image Processing Chip for Planetary Navigation Applications

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# Introduction

- Purpose of Image Processing Chip
  - Feature Extraction (and Tracking) Integrated Circuit
  - FEIC
- FEIC Architecture
- Testing
- Test Results
- Problems and solutions
  - Wandering tracks
  - Short tracks
- Conclusions



### Partners

- ESA
- Astrium SAS, Selex Galileo, CGS, INETI, Astrium UK, GMV, TAS-I
- Contracts:
  - PANGU
  - NPAL,
  - HARVD,
  - VisNav,
  - ExoMars,
  - NEO-GNC,
  - VisNav-EM



# Purpose of FEIC

- Vision-based navigation
- Image co-processor
  - Offloads the image processing task from the host processor
- Applications:
  - Planetary landing
  - Asteroid station keeping, surveying and landing
  - Rover navigation



# Feature Extraction and Tracking IC

- Image processing
  - Extracts feature points using Harris detector
  - Tracks them from frame to frame
  - Reports tracked feature points to GNC computer
  - For integration with IMU using a Kalman filter
- Image processing chip
  - Chip to do image processing (FEIC)
  - Developed for ESA
  - Navigation for Planetary Approach and Landing (NPAL)
  - Operates at 20 Hz frame rate





## **Feature Extraction**

- Compute gradient products:
- Convolve with Gaussian-like kernel
- Apply the Harris corner detector
- Apply a 7×7 local maxima filter
- Sort and select the strongest features
- Uses fixed-point saturating arithmetic





# Harris Corner Detector

 $W = 7 \times 7$  convolution kernel  $\mathbf{M} = \begin{bmatrix} \left(\frac{\partial}{\partial x} \frac{\partial}{\partial x}\right) \otimes W & \left(\frac{\partial}{\partial x} \frac{\partial}{\partial y}\right) \otimes W \\ \left(\frac{\partial}{\partial x} \frac{\partial}{\partial y}\right) \otimes W & \left(\frac{\partial}{\partial y} \frac{\partial}{\partial y}\right) \otimes W \\ \left(\frac{\partial}{\partial y} \frac{\partial}{\partial y}\right) \otimes W & \left(\frac{\partial}{\partial y} \frac{\partial}{\partial y}\right) \otimes W \end{bmatrix}$  $H = \det(M) - k \cdot \operatorname{trace}(M)^2$  $k = \frac{5}{128} \approx 0.04$ 



# Local Maxima Unit

- 7x7 local maxima unit:
  - pass feature if stronger than all others ...
  - ... within a 7×7 neighbourhood
  - otherwise suppress the feature
  - all features output will be >3 pixels apart





0	5	6
3	6	3
4	1	0
FAIL		



# Select/Sort Unit

#### Purpose:

- stores feature points sorted by strength ...
- ... but only retains strongest N points (e.g. N=200)
- 8 fields (120 bits) of data per point

#### Performance constraint:

- feature points arrive every 8 cycles at 40 MHz
- must sort and store features within this time
- we use a single-cycle insertion sort
- uses lots of chip area so N<100 for V6000



# Feature Tracking

Using intensity insensitive correlation:
 – correlate 7×7 texture with N×M window

- OBC suggests window origin (guidance hint)

FEIC returns correlation strength(s)

- implemented using 32-bit fixed point arithmetic

$$c(x,y) = \frac{\sum_{i=1}^{N} \sum_{j=1}^{N} \left[ \left( I(x+i, y+j) - \bar{I}_{xy} \right) \left( T(i, j) - \bar{T} \right) \right]}{\sqrt{\sum_{i=1}^{N} \sum_{j=1}^{N} \left[ I(x+i, y+j) - \bar{I}_{xy} \right]^{2} \sum_{i=1}^{N} \sum_{j=1}^{N} \left[ T(i, j) - \bar{T} \right]^{2}}}$$



# Feature Tracking

Invoked for each tracked point:

load correlator with feature point texture

stream N×M pixel window into 7×7 average

– compute correlation measure (pipelined)

numerator/denominator computed on alternate cycles

recover 8-neighbours from correlator stream

- filter to identify the global maximum





# List Manager

- Provides storage and the interface to OBC:
  maintains the LN (new) and LT (tracked) points lists
  - responsible for all OBC feature point operations
  - drives the tracker and (partly) the extractor
- Provides interface between FEIC and OBC

# Development of FEIC by UoD

- Prototyped in C++:
  - started with "standard" algorithms
  - replaced with H/W data-flow algorithms
  - defined FEIC architecture
  - implemented FEIC in VHDL

Validation using PANGU image sequences:
 validated simulated VHDL against C++

- FPGA driver can validate output against C++
- Validation using Astrium acceptance tests:
   compares performance against floating-point



<u>NPAL Camera</u> EADS Astrium (France) Selex Galileo (Italy) University of Dundee

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1 1 55 4 28 2 14

No. of Concession, Name

## **FEIC Test Environment**

### PANGU

- Planet and Asteroid Natural-scene Generation Utility
- Software tool
  - Simulates planets and asteroids
  - Simulates cameras and other sensors viewing those bodies
  - Developed specifically to test vision based GNC algorithms
  - Realistic and high performance
  - Extensively vested and validated so that it can be used for testing flight systems

- Being used for lander, rover, orbiter simulation

# **Open and Closed Loop Testing**

- Open loop testing
  - PANGU image sequence from known positions
  - Fed to vision based navigation system
- Closed loop testing
  - Spacecraft dynamics simulation
  - Gives PANGU camera position and orientation
  - PANGU provides image of surface from that position
  - Position and orientation of target object
    - Can also be specified
    - Useful for asteroids
- Testing of real hardware
  - Using monitor and camera



#### PANGU Cratered Asteroid + Spacecraft Shadow





#### PANGU Martian Surface and Rover





#### PANGU Lunar South Pole





### **Empirical Results**



Tracking using the FPGA implementation.

Tracks start red then change colour depending on length of time tracked Only need to track six points successfully to recover motion of spacecraft





# Improvements to image processing

- Wandering Tracks
  - Possible for feature tracker to lock on to a similar feature close to the one being tracked.
  - Causes wandering features
  - Example in top left hand corner of image
  - True point is no longer in the image
  - Rather than being discarded a similar point close by has been picked up
- Proposed improvement
  - When a feature is extracted
  - Check that there are no similar features close by
  - That could cause confusion
  - If there are, reject the feature point











### Problem with rotation







# Image Processing Improvements

- Short tracks
  - With changes in the image perspective the feature template can become out of date
- Proposed improvement 1
  - Update feature template regularly
  - This causes feature wandering
  - Good tracking on a slightly incorrect feature template position propagated across many frames
- Proposed improvement 2
  - Update feature template regularly
  - Track with both old and new template
  - If feature positions are close over many frames
  - Accept new point and discard old one







## Currrent work

- Results shown are without aiding
- Assessment of robustness
- Known landmark tracking
- Dense correlation for stereovision
- Surface elevation determination
- Design of a generic image processing device for vision-based navigation applications



# Conclusions

Image processing co-processor

- Offloads computationally intense work from OBC
- Interfaced using SpaceWire
- Simulation using PANGU
  - very useful for developing and testing image processing algorithms
- Potential improvements to FEIC
- Shown, using PANGU, to give significant benefit
  - Avoid wandering tracks
  - Longer tracks
- Eases job of OBC