

# FPGA acceleration of the Pulsar Search Pipeline

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# Overview

- ▶ UoA team actively contributing since 2013
  - ▶ Member of NZA
- ▶ Work on CSP (Central Signal Processor) package
  - ▶ Focus on PSS (Pulsar Search)
  - ▶ We are part of TDT team at Uni of Manchester led by Benjamin Stappers
  - ▶ Collaborate closely with Prabu Thiagaraj, Raman Research Institute (formerly Uni. of Manchester)
- ▶ Investigate and develop PSS pipeline stages on FPGAs
  - ▶ FPGAs used as high performance, low-power acceleration devices
- ▶ Using OpenCL, a high level (C-based) language
- ▶ Targeting high-end FPGAs, initially Intel/Altera, but now also Xilinx

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# Outline

## Pulsar Search and FPGAs

Pulsar search

Our contributions (so far)

Acceleration hardware and software

## Folding and Optimisation (FLDO)

Input data

Folding (FOLD)

Optimisation (OPT)

## FDAS and Matched Filter Group

FDAS Module

Matched Filter Group Optimisation

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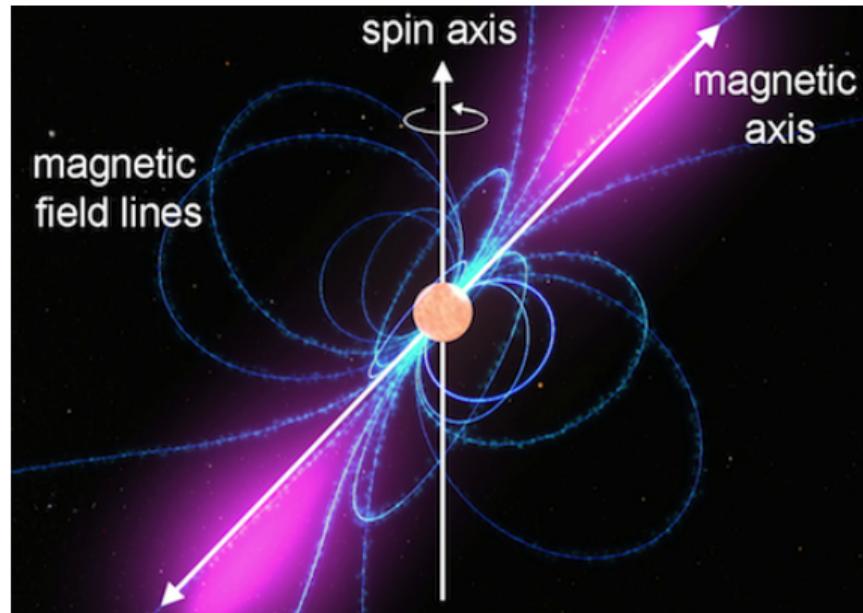
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# What Is A Pulsar?

- ▶ Pulsars are rapidly rotating collapsed stars.
- ▶ Emit highly directional EM radiation.
- ▶ Millisecond pulsars more accurate than (some) atomic clocks.
- ▶ Pulsar timing arrays for mapping the universe.
- ▶ Binary pulsars for testing general relativity.

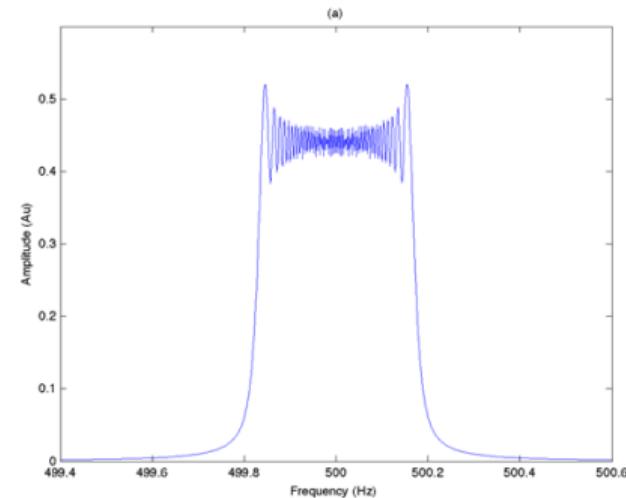
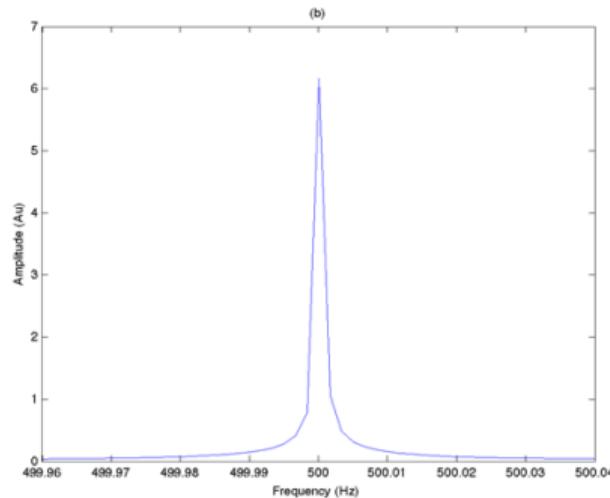


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<sup>1</sup>Image Credit: NASA/Tod Strohmayer (GSFC)/Dana Berry (Chandra X-Ray Observatory)

# Pulsar Search

- ▶ Acceleration search: 1) Time-domain 2) Frequency-domain



# Pulsar Search challenge

- ▶ Unknown position
  - ▶ Scan sky
- ▶ Unknown distance
  - ▶ Trying 1000-2000 different dedispersion measures (DM)
- ▶ Unknown period
- ▶ Unknown orbital frequency
  - ▶ Compensate with matched filtering (currently 85 different filters)
- ▶ Faint signal
  - ▶ Integrating for ca. 536 secs

=> Brute Force Search

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Our group contributed to SKA CSP pulsar search sub-element

- ▶ SKA1 CSP Pulsar Search Sub-element Prototype Plan document
  - ▶ Proto-PS\_SKA-TEL-CSP-0000116\_2\_PulsarSearchProtoPlan\_Armour\_2016-10-29.docx, contributed to Section 5.4.5
- ▶ SKA CSP Pulsar Search Sub-element Detailed Design Document
  - ▶ SKA-TEL-CSP-0000082\_2\_PSSDetDesDoc\_Karastergiou\_2018-05-29.docx, contributed to Section 16.3
- ▶ SKA-TEL-CSP-PSS\_0000007\_TestReportFTConvolution\_01.doc
- ▶ SKA-TEL-CSP-PSS\_0000008\_TestReportHarmonicSum\_01.doc
- ▶ SKA-TEL-CSP-PSS\_0000009\_TestReportMedianFiltering\_01.doc

# Implementations

Contributed OpenCL on FPGA implementations of

- ▶ Fourier Domain Acceleration Search
  - ▶ Multi-filter FT convolution
  - ▶ Harmonic summing
  - ▶ Threshold (candidate) detection
- ▶ Large-scale median-filtering

For all designs

- ▶ Tested and evaluated on real hardware
- ▶ Licensed under OpenSource (GPL)

- ▶ Wang, H., Thiagaraj, P., & Sinnen, O. (2019). FPGA-based Acceleration of FT Convolution for Pulsar Search Using OpenCL. *ACM Transactions on Reconfigurable Technology and Systems (TRETS)*, 11(4), 24.
- ▶ Wang, H., Thiagaraj, P., & Sinnen, O. (2018). Harmonic-summing Module of SKA on FPGA--Optimising the Irregular Memory Accesses. *IEEE Transactions on Very Large Scale Integration (VLSI) Systems*, Early Access, 10.1109/TVLSI.2018.2882238.
- ▶ Wang, H., Thiagaraj, P., & Sinnen, O. (2019). Combining Multiple Optimised FPGA-based Pulsar Search Modules Using OpenCL. *Journal of Astronomical Instrumentation*. (Accepted)

# Publications-Conference

- ▶ Wang, H., Sinnen, O. (2019, January). *Design Space Exploration of FPGA-based FIR Filter Group in Fourier Domain*. In 2019 International Conference on Electronics (ICEIC 2019). IEEE.
- ▶ Sherwin, K., Stappers, B., Thiagaraj, P., Wang, K., & Sinnen, O. (2018, December). *Investigating how hardware architectures are expressed in high-level languages for an SKA algorithm*. In Field-Programmable Technology (FPT), 2018 International Conference on. IEEE.
- ▶ Wang, H., Stappers, B., Thiagaraj, P., & Sinnen, O. (2018, December). *Optimisation of Convolution of Multiple Different Sized Filters in SKA Pulsar Search Engine*. In Field-Programmable Technology (FPT), 2018 International Conference on. IEEE.
- ▶ Sherwin, T., Kevin, I., Wang, K., Thiagaraj, P., & Sinnen, O. (2018, August). *Median filtering with very large windows: SKA algorithms for FPGAs*. In 2018 28th International Conference on Field Programmable Logic and Applications (FPL) (pp. 196-1965). IEEE.
- ▶ Wang, H., Zhang, M., Thiagaraj, P., & Sinnen, O. (2016, December). *FPGA-based acceleration of FDAS module using OpenCL*. In Field-Programmable Technology (FPT), 2016 International Conference on (pp. 53-60). IEEE.
- ▶ Wang, H., Gante, J., Zhang, M., Falcão, G., Sousa, L., & Sinnen, O. (2016, December). *High-Level Designs of Complex FIR Filters on FPGAs for the SKA*. In High Performance Computing and Communications (HPCC), 2016 IEEE 18th International Conference on (pp. 797-804). IEEE.
- ▶ Wang, H., & Sinnen, O. (2015, December). *FPGA based acceleration of FDAS module for Pulsar Search*. In Field Programmable Technology (FPT), 2015 International Conference on (pp. 240-243). IEEE.

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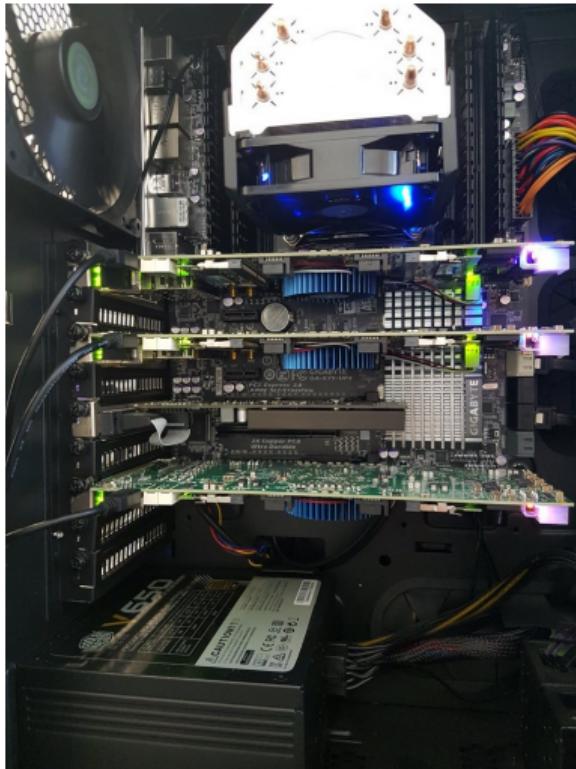
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# Hardware



- ▶ Terasic DE5-Net with Intel Stratix V x4
- ▶ Nallatech 385A with Intel Arria 10 x6
- ▶ Xeon+FPGA with Intel Arria 10  
(supported by Intel hardware accelerator research program, remote access)
- ▶ Xilinx VCU1525 with Virtex UltraScale+ XCVU9P x1
- ▶ Xilinx VCU1550 with Virtex UltraScale+ FPGA x1

# Hardware

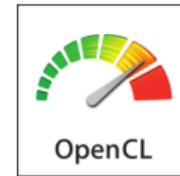


# Hardware Specifications

Device	Terasic DE5-Net <b>(S5)</b>	Nallatech 385A <b>(A10)</b>	Xeon+FPGA <b>(X+A10)</b>	Xilinx VCU1525 <b>(U9)</b>
Hardware	Intel Stratix V 5SGXA7	Intel Arria 10 GX1150	Intel Arria 10 GX1150	Xilinx Virtex UltraScale+ XCVU9P
Technology	<i>28nm</i>	<i>20nm</i>	<i>20nm</i>	<i>16nm</i>
Compute resource	622,000 LEs	1,506,000 LEs	1,506,000 LEs	2,586,000 LEs
	256 DSP blocks	1,518 DSP blocks	1,518 DSP blocks	6,840 DSP blocks
On-chip memory size	<i>50Mb</i>	<i>53Mb</i>	<i>53Mb</i>	<i>345.9 Mb</i>
Off-chip memory size	<i>2 × 2GB DDR3</i>	<i>2 × 4GB DDR3</i>	–	<i>4 × 16GB DDR4</i>
OpenCL global memory bandwidth	<i>25,600MB/s</i>	<i>35,128MB/s</i>	<i>25,000MB/s</i>	
Max clock frequency	<i>600MHz</i>	<i>1.5GHz</i>	<i>1.5GHz</i>	
Max power consumption	—	75W	–	225W

# OpenCL

- ▶ High level approach to parallel programming.
- ▶ Execution portable.
  - ▶ Same code, multiple platforms.

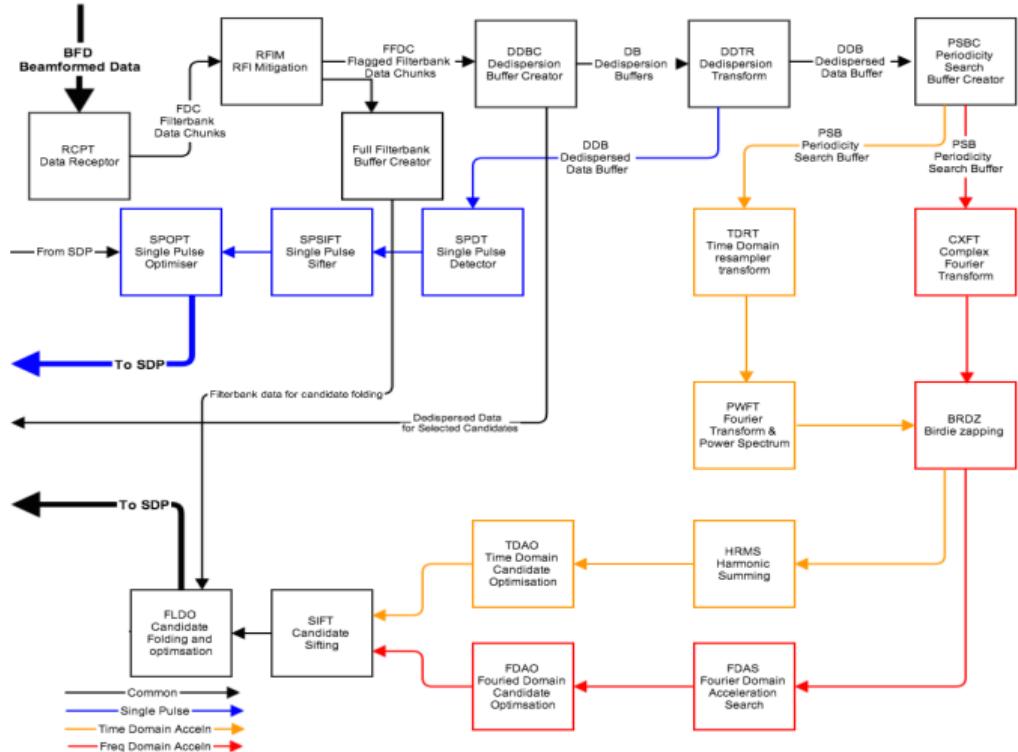


```
__kernel
void median(global const PIXEL_TYPE * restrict frame_in,
           global O_TYPE * restrict frame_out)
{
    #pragma unroll UNROLL_SIZE
    for (int col=OTHER_HALF; col<(COLS-OTHER_HALF); col+=(P_SIZE)) {
        // Start with a window full of 0's
```

# OpenCL differences: Intel and Xilinx

Optimization Features	Intel (AOC)	Xilinx (XOCC)
Pipe	Channel and Pipe	Pipe
Loop unroll	#pragma unroll	<code>__attribute__((opencl_unroll_hint))</code>
Pipelining	Automatically	<code>__attribute__((xcl_pipeline_loop))</code> <code>__attribute__((xcl_dataflow))</code>
Kernel Vectorization	<code>__attribute__((num_simd_work_item))</code>	<code>__attribute__((vec_type_hint))</code>
Floating-point operations	balanced tree (-fp-relaxed) rounding operations (-fpc)	No related attributes
Multiple memory banks	<code>-no-interleaving=&lt;global_memory_type&gt;</code>	<code>--max_memory_ports -sp ....</code>
Local memory partition	<code>__attribute__((&lt;memory_type&gt;, ...))</code>	<code>__attribute__((xcl_array_partition))</code>

# Pulsar Search Engine (PSS)



Complex system requiring more than 10POps.

- ▶ RFIM
- ▶ FLDO
- ▶ FDAS

# Specifications of Task

Parameter	Description	Value
$N_{beam}$	# of beams	1000 ~ 2000
$T_{obs}$	Observation period	540s
$N_{DM}$	# of de-dispersion measure (DM) trails	6000
$N_{channel}$	# of complex samples per group	$2^{21}$
$N_{template}$	# of templates/filter	85
$N_{tap}$	# of average template/filter length	211
$N_{hm}$	# of harmonic planes	8
$N_{candidate}$	# of candidates per harmonic plane	200

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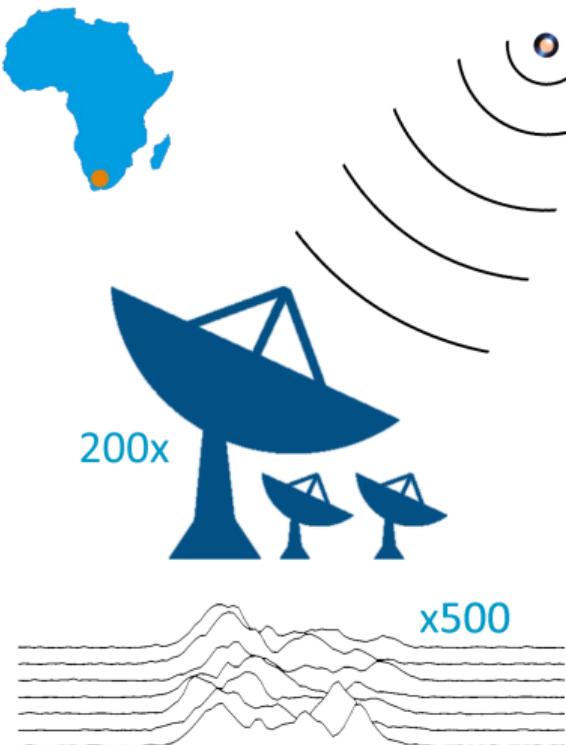
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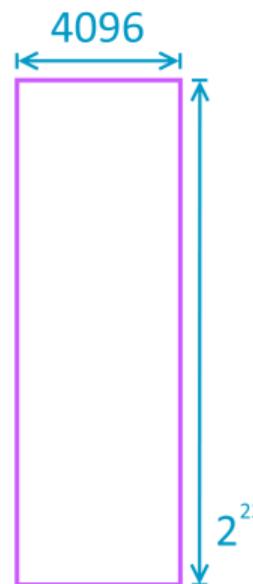
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# Input data



SKA-Mid



- ▶ 536 seconds in  $2^{23}$  time steps.
- ▶ 8bit samples from 4096 frequency channels every 64us.
- ▶ Up to 500 candidate profiles given for each of two beams.
  - ▶ Initial parameters & desired search space.
- ▶ 3 parameters to optimise.
- ▶ 256 **million trials/second/node**.

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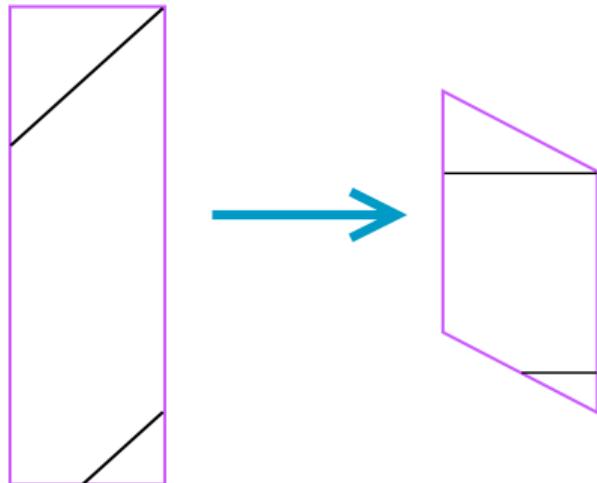
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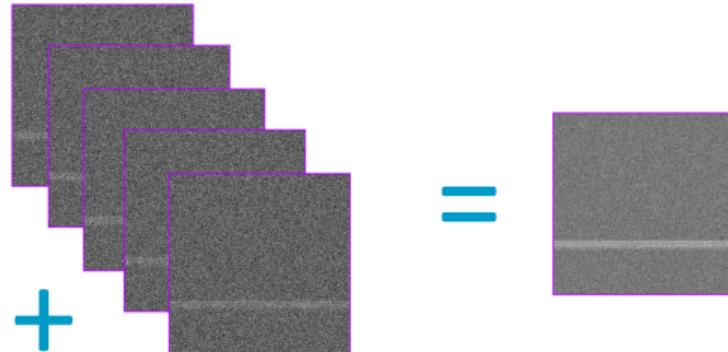
Matched Filter Group Optimisation

# FOLD - Integration & dedispersion



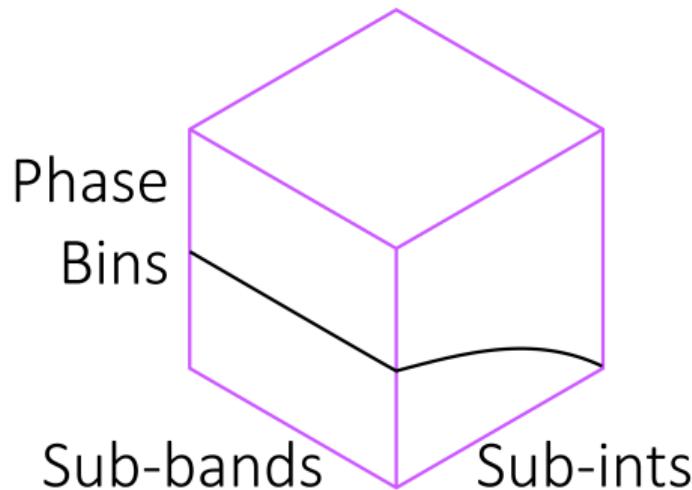
- ▶ Input may be integrated, depending on pulsar data.
- ▶ Dispersion causes time delays at lower frequencies which must be corrected.
  - ▶ Described by Dispersion Measure (*DM*).
  - ▶ Higher frequencies delayed more to compensate.
  - ▶ Integer bin shifts based on frequency.
- ▶ Frequency channels integrated to  $\leq 128$  sub-bands for folding.

# FOLD - Repeating pulses



- ▶ Combine multiple pulses.
- ▶ Random background noise cancels out.
- ▶ Repeating pulsar signal retains strength.
- ▶ Pulsar SNR improves.

# FOLD - 3D data cube



- ▶ Up to 64 sub-integrations.
- ▶ Each sub-integration containing series of pulses folded together.
- ▶ Each folded pulse up to 128 sub-bands across 128 phase bins.

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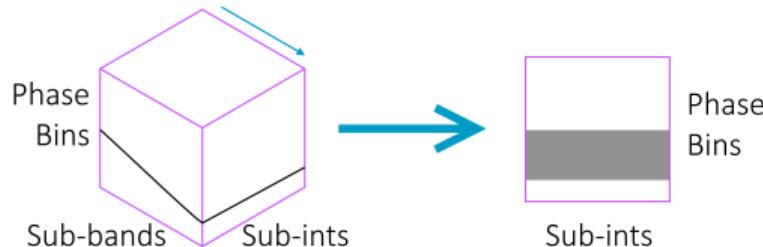
Optimisation (OPT)

## FDAS and Matched Filter Group

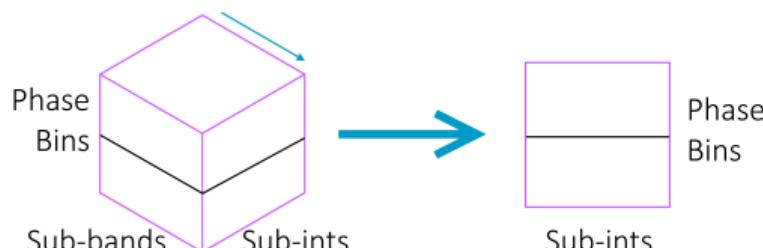
FDAS Module

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# OPT - Dispersion Measure (DM) trials

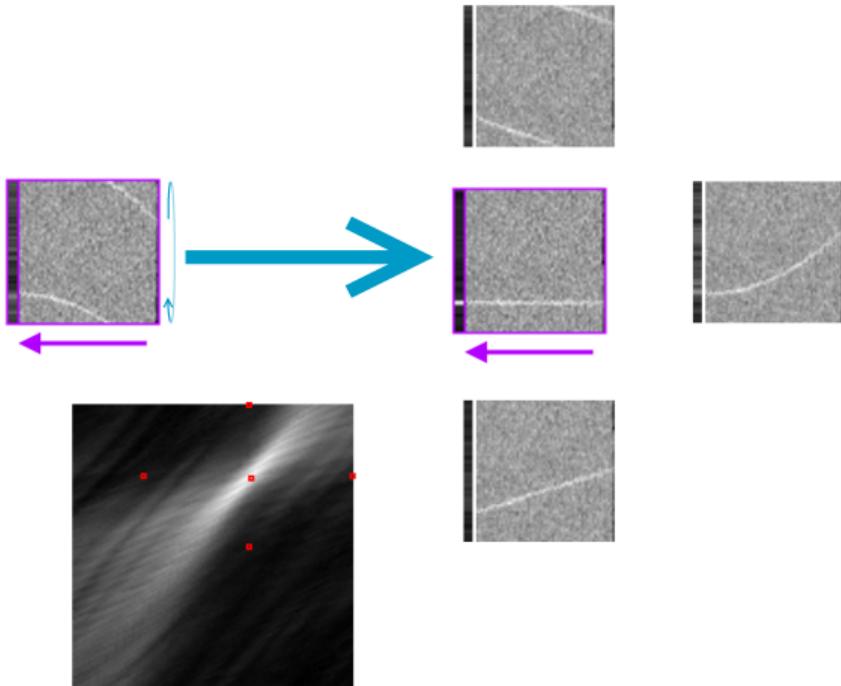


- ▶ Higher frequency values delayed in time.
- ▶ Shifting phase bins across all sub-integrations corrects delay.
- ▶ Each trial followed by integration of the sub-bands, giving a flattened 2D array.



- ▶ Period trials independent of sub-bands.
- ▶ Integration reduces necessary computation.

# OPT - Period trials



- ▶ Both  $P$  and  $\dot{P}$  varied, sweeping performed as a series of bin shifts.
- ▶ One dimension rotates while other fixed.
- ▶ Shifting phase bins across all sub-bands approximates refolding with new  $P$  and  $\dot{P}$ .
- ▶ Higher subintegrations (later pulses) shifted more.
- ▶ Values rotated out the bottom placed at the top.
  - ▶ Represents shift from end of one period to start of next.

# Output

- ▶ Initial data cube sent to SDP for further processing.
- ▶ Also outputs metadata including estimated pulse width and SNR for each trial.
- ▶ Optimal parameters then determined, as well as relationships such as harmonics.

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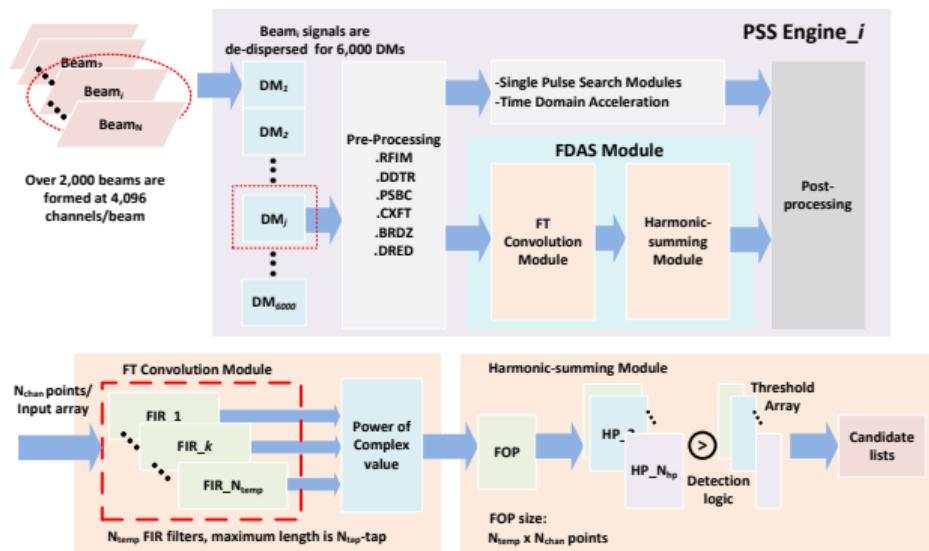
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# Fourier-domain Acceleration Search (FDAS)

FDAS module is applied to search for (binary) pulsars with constant frequency derivatives in Fourier-domain

## FT Convolution module + Harmonic-summing module



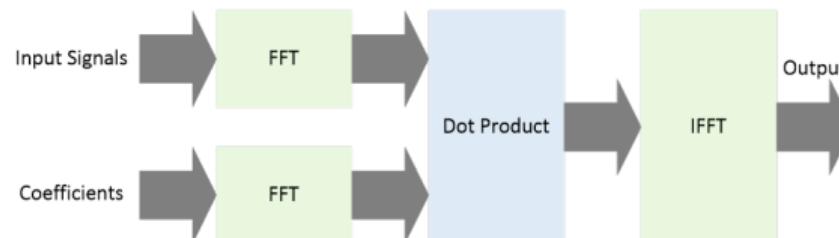
# FT Convolution Module

- ▶ Time-domain FIR Filter (TDFIR)

$$y_m[i] = \sum_{k=0}^{K-1} x_m[i - k] h_m[k], \text{ for } i = 0, 1, \dots N - 1$$

- ▶ Frequency-domain FIR Filter (FDFIR)

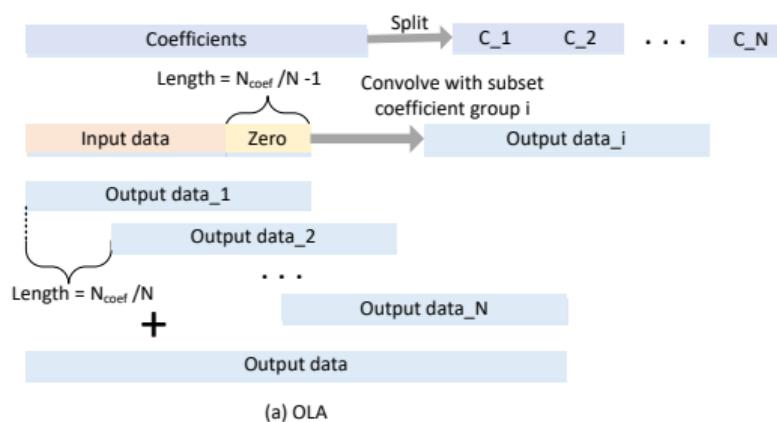
$$\mathcal{F}\{f * h\} = \mathcal{F}\{f\} \cdot \mathcal{F}\{h\}$$



# FT Convolution Module

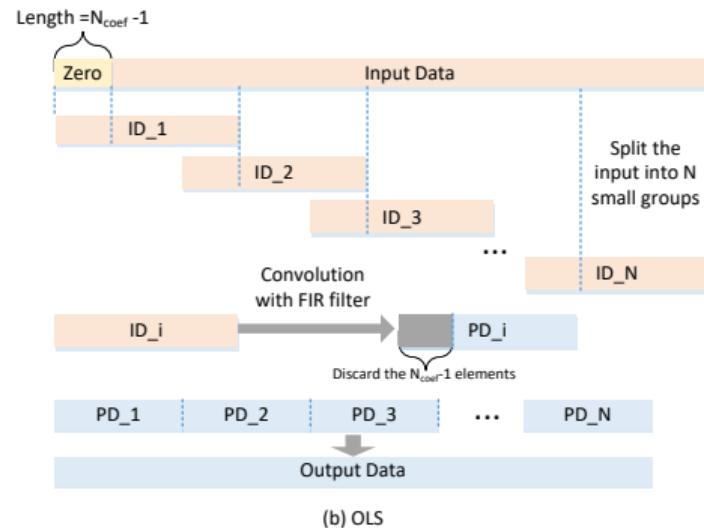
## Overlap-add Algorithm

- ▶ Split the coefficient array  
→ **OLA-TD**



## Overlap-save Algorithm

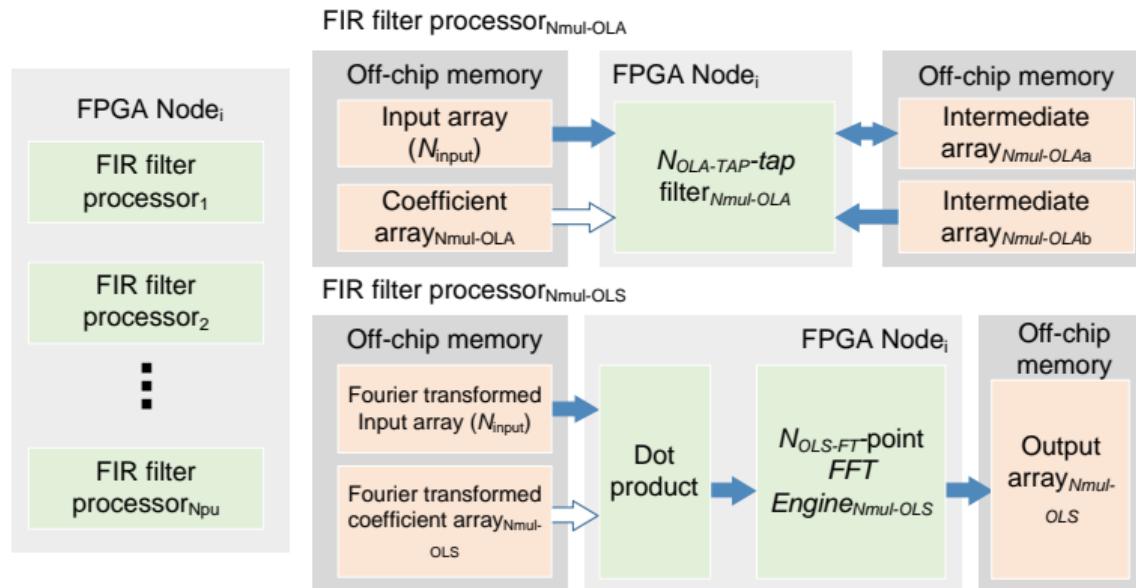
- ▶ Split the input array  
→ **OLS-FD**



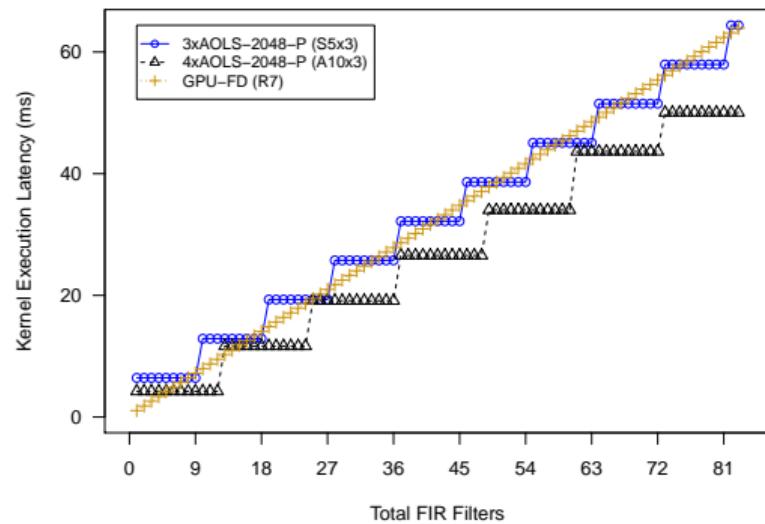
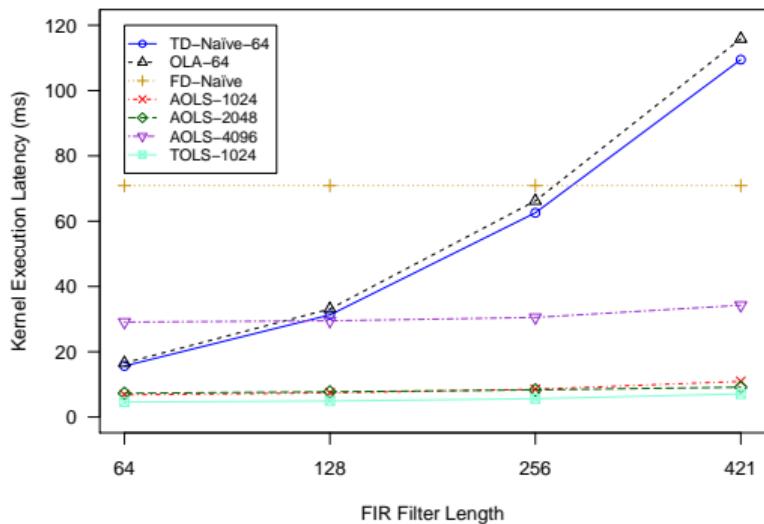
# FT Convolution Module

For general implementation, all filters are extended to the same length as the longest filter.

- ▶ Multiple processing units can be implemented in one FPGA image
- ▶ All processing units share the same input array or Fourier transformed input array



# FT Convolution Module-Evaluation



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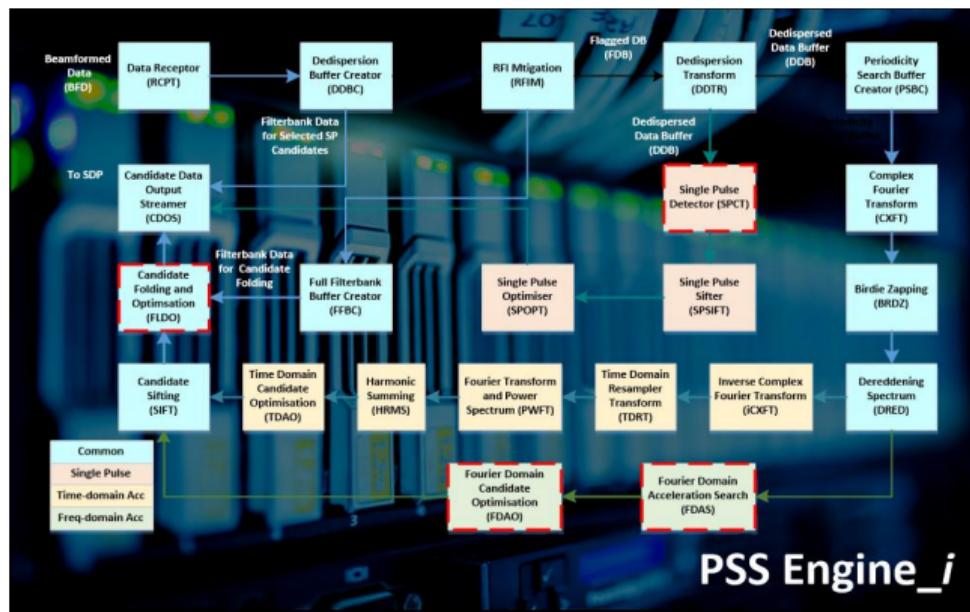
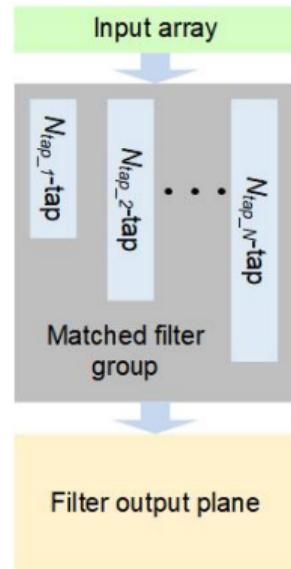
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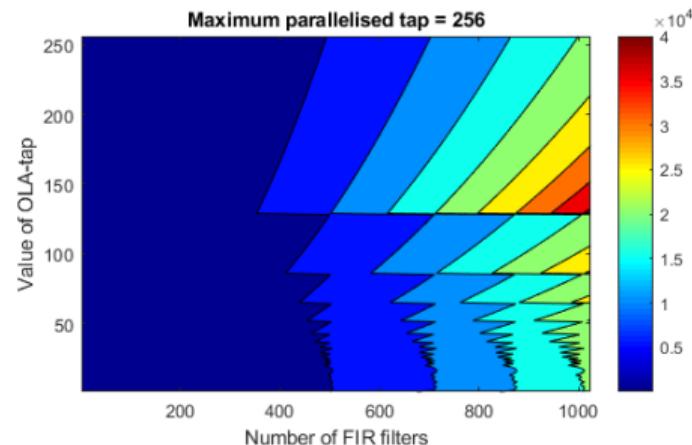
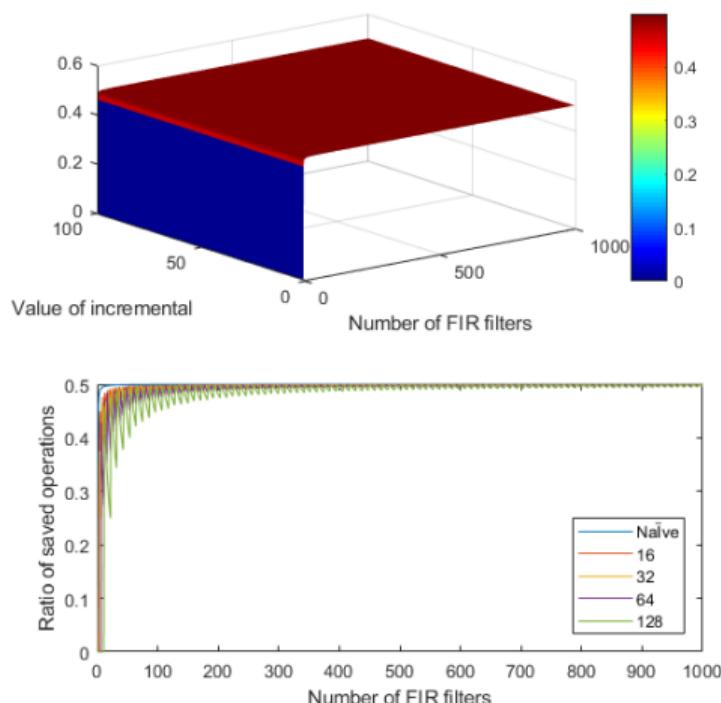
# Matched Filter Group Optimisation

*MF – (N<sub>filter</sub>, N<sub>inc</sub>, Tap1): A group of FIR filters with different sizes*

$$Tap_{i+1} = Tap_i + N_{inc}$$



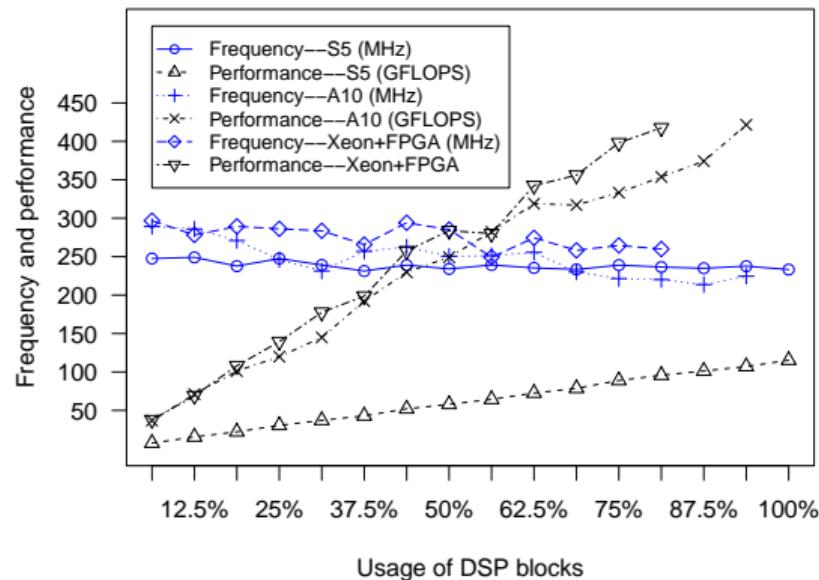
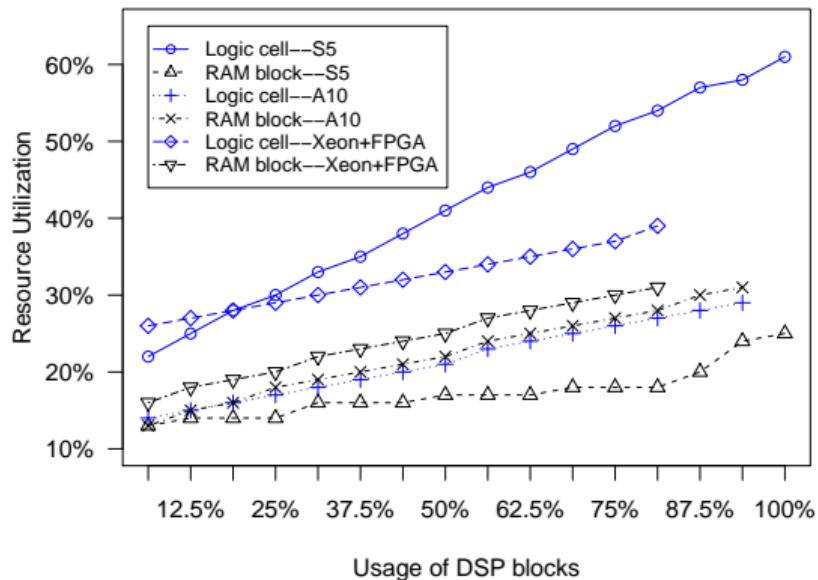
# Matched Filter Group Optimisation-TD



- ▶ The smaller the number of launch times the faster the design
- ▶ Search for the optimal  $N_{OLA-tap}$  for the OLA-TD based implementation on a specific FPGA

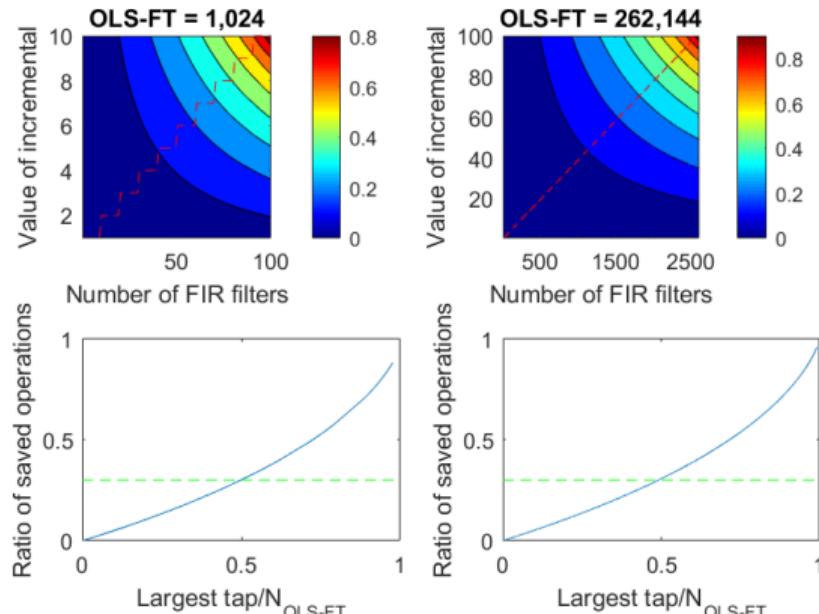
# Matched Filter Group Optimisation-TD

Main factors: 1) DSP blocks and 2) off-chip memory bandwidth



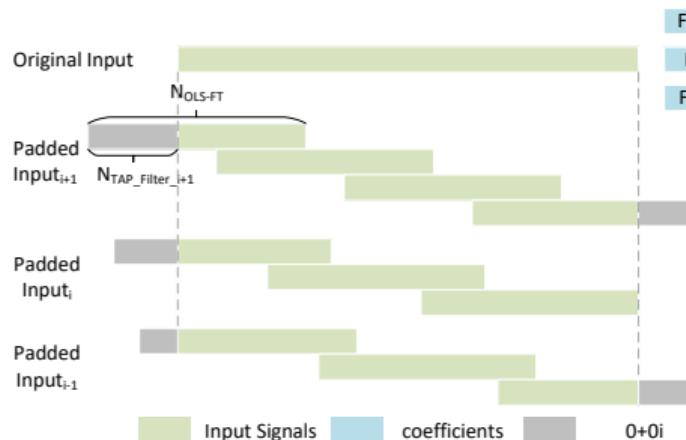
Load balancing—longest processing time rule (LPT)

# Matched Filter Group Optimisation-FD



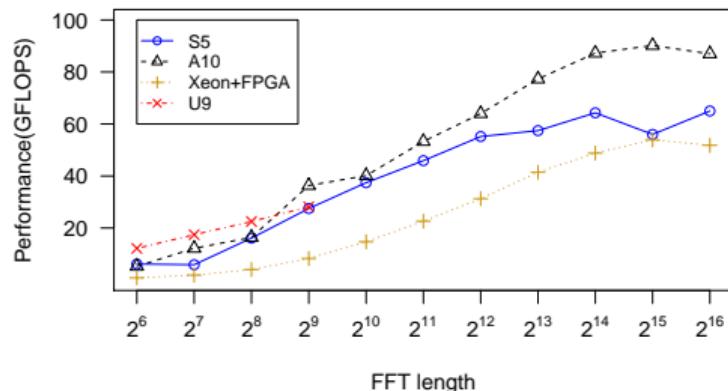
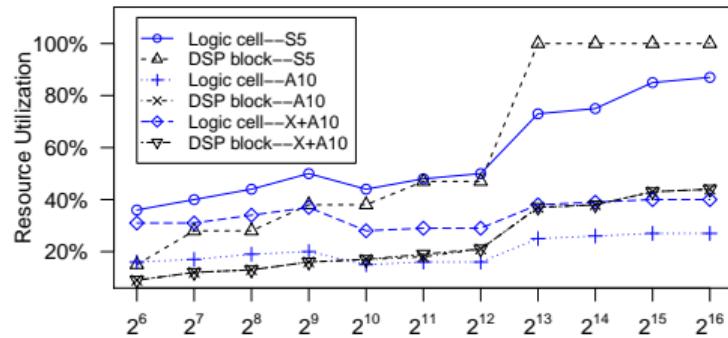
- ▶ The higher the value of  $Tap_{N_{filter}} / N_{OLS-FT}$ , the more operations can be saved (when  $Tap_{N_{filter}} / N_{OLS-FT} = 0.5$ , up to 30% operations can be saved)
- ▶ The larger the value of  $N_{OLS-FT}$ , the more logic resource is costs
- ▶ For a specific  $N_{OLS-FT}$ ,  $N_{pu}$  is decided by the logic resources on an FPGA

# Matched Filter Group Optimisation-FD



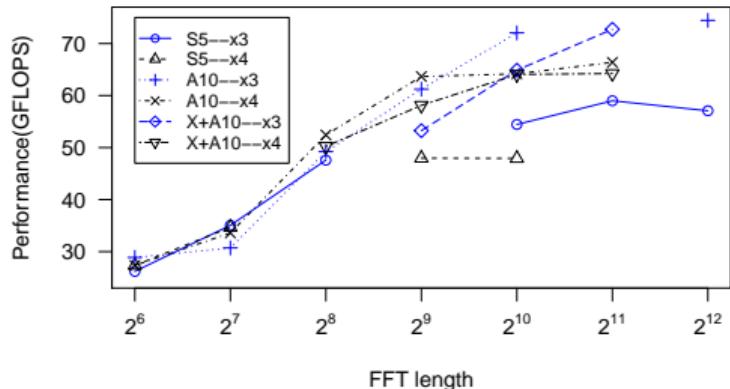
- ▶ Padded input group for small filters cannot be used for large FIR filters
- ▶ The higher  $N_{TAP\_filter}/N_{OLS-FT}$ , the larger the size of padded input group
- ▶ Multiple padded input groups lead to the increase of needed off-chip memory
- ▶ The goal is find the suitable  $N_{OLS-FT}$  and the total number of padded input group

# Matched Filter Group Optimisation-FD

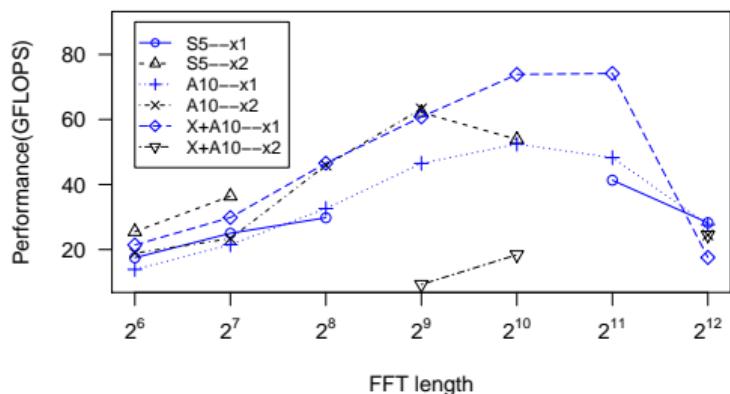


- ▶ processing multiple points in parallel such as 4, 8, and 16 (8 in Figure)
- ▶ Resource usage drops when FFT length increased from  $2^9$  to  $2^{10}$  and increases dramatically when FFT length is larger than  $2^{12}$ .
- ▶ The larger the FFT length, the higher the performance from  $2^6$  to  $2^{12}$

# Matched Filter Group Optimisation-FD

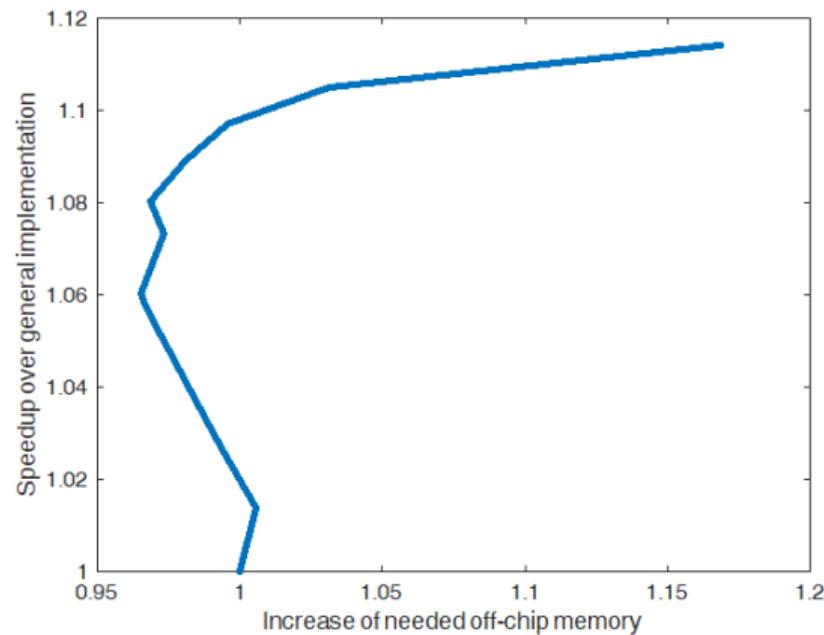
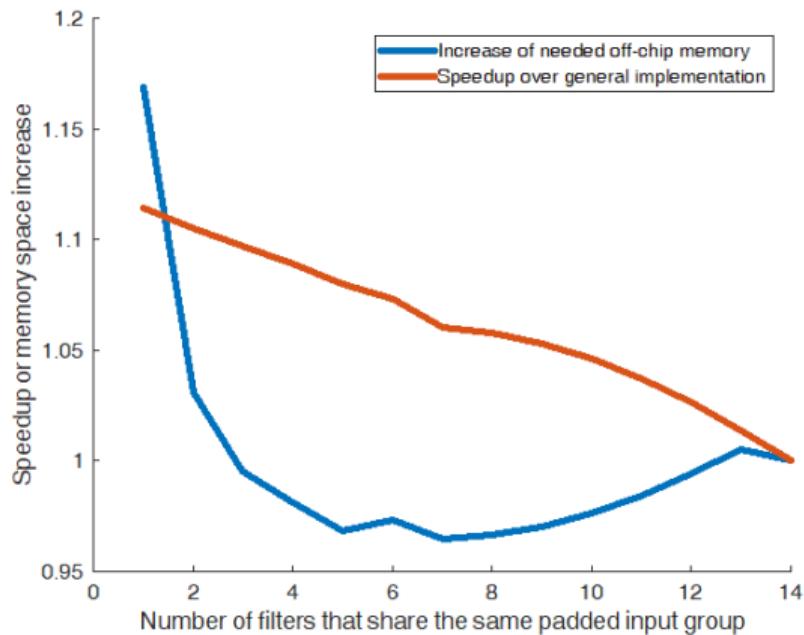


- ▶ 4-point FFT engine  $N_{pu} = 3$  and  $N_{pu} = 4$  are tested
  - ▶ S5:  $N_{OLS-FT} = 2048$ ,  $N_{pu} = 3$
  - ▶ A10:  $N_{OLS-FT} = 4096$ ,  $N_{pu} = 3$
  - ▶ X+A10:  $N_{OLS-FT} = 2048$ ,  $N_{pu} = 4$



- ▶ 8-point FFT engine  $N_{pu} = 1$  and  $N_{pu} = 2$  are tested
  - ▶ S5:  $N_{OLS-FT} = 1024$ ,  $N_{pu} = 2$
  - ▶ A10:  $N_{OLS-FT} = 512$ ,  $N_{pu} = 2$
  - ▶ X+A10:  $N_{OLS-FT} = 2048$ ,  $N_{pu} = 1$

# Matched Filter Group Optimisation-FD



# Summary

- ▶ Investigated the optimisation of matched filter group implementation for the pulsar search engine.
- ▶ For the optimisation in time-domain, the main task is to balance the size of  $N_{OLA-tap}$  and off-chip memory bandwidth
- ▶ For the optimisation in Fourier-domain, the main task is to balance the off-chip memory cost and the speedup in execution latency
- ▶ Duplicating more processing units cannot improve the performance when the required bandwidth outnumbers the device off-chip memory bandwidth

# Conclusions and future work

- ▶ Investigated high-performance implementations of PSS pipelines on FPGA
  - ▶ Meeting requirements using 'portable' high-level approach OpenCL

## Future

- ▶ Investigate other stages for implementation
- ▶ Investigate the Candidate sifting module (SIFT) that process the output from the Fourier Domain Candidate Optimisation (FDCAO) module

# OpenMP conference

International OpenMP conferences in Auckland in September 2019

- ▶ OpenMPCon – developer conference 9-10 Sept. – [www.openmpcon.org](http://www.openmpcon.org)



- ▶ Tutorial day 11 Sept.
- ▶ IWOMP - academic conference 12- 13 Sept. – [parallel.auckland.ac.nz/iwomp2019](http://parallel.auckland.ac.nz/iwomp2019)

