

Title: Uses of HPC in radar data processing and analysis

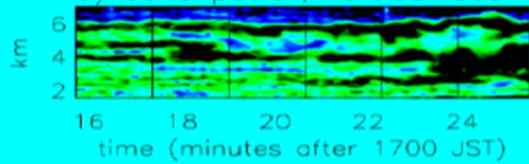
Date: May 07, 2014 03:50 PM

URL: <http://pirsa.org/14050056>

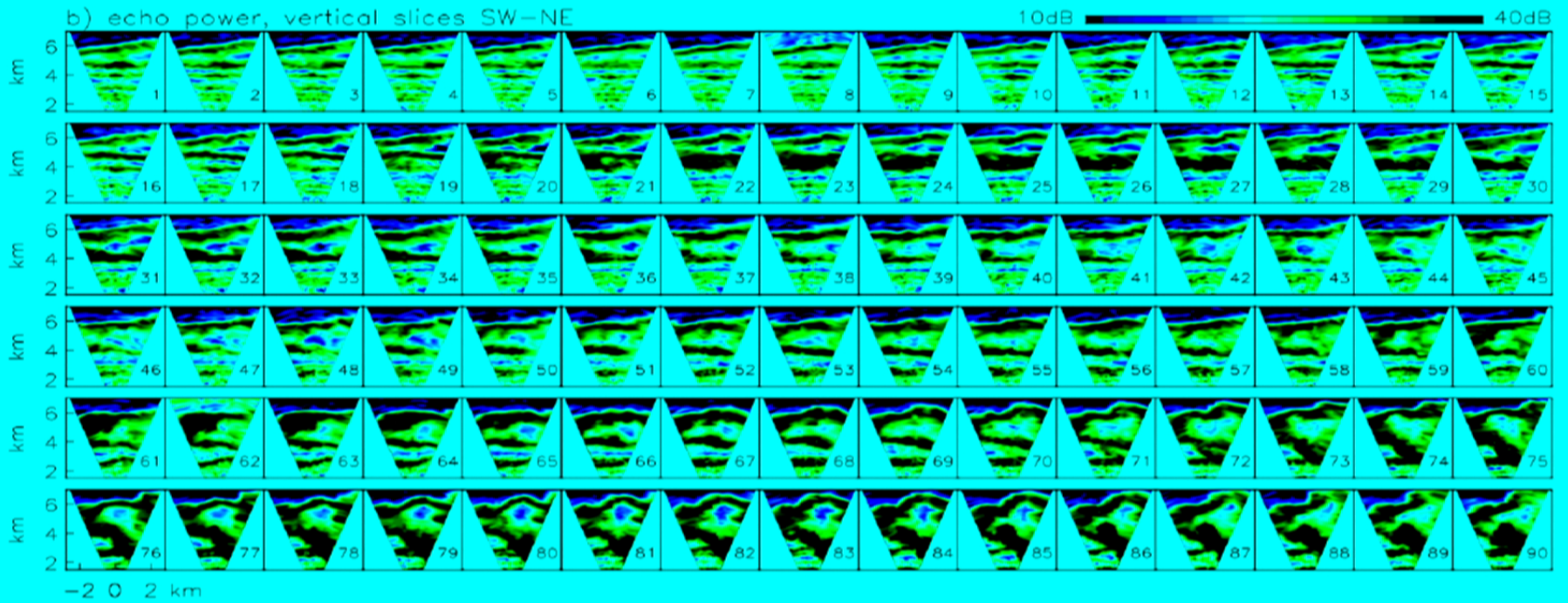
Abstract: A low troposphere MST type radar located in Costa Rica was used to gather information up to 6 km. With the digital radar technique used thousands of sweeps can be recorded every second. Challenges in processing spectral analysis and radar imaging were addressed with tools provided by HPC.

MU Radar

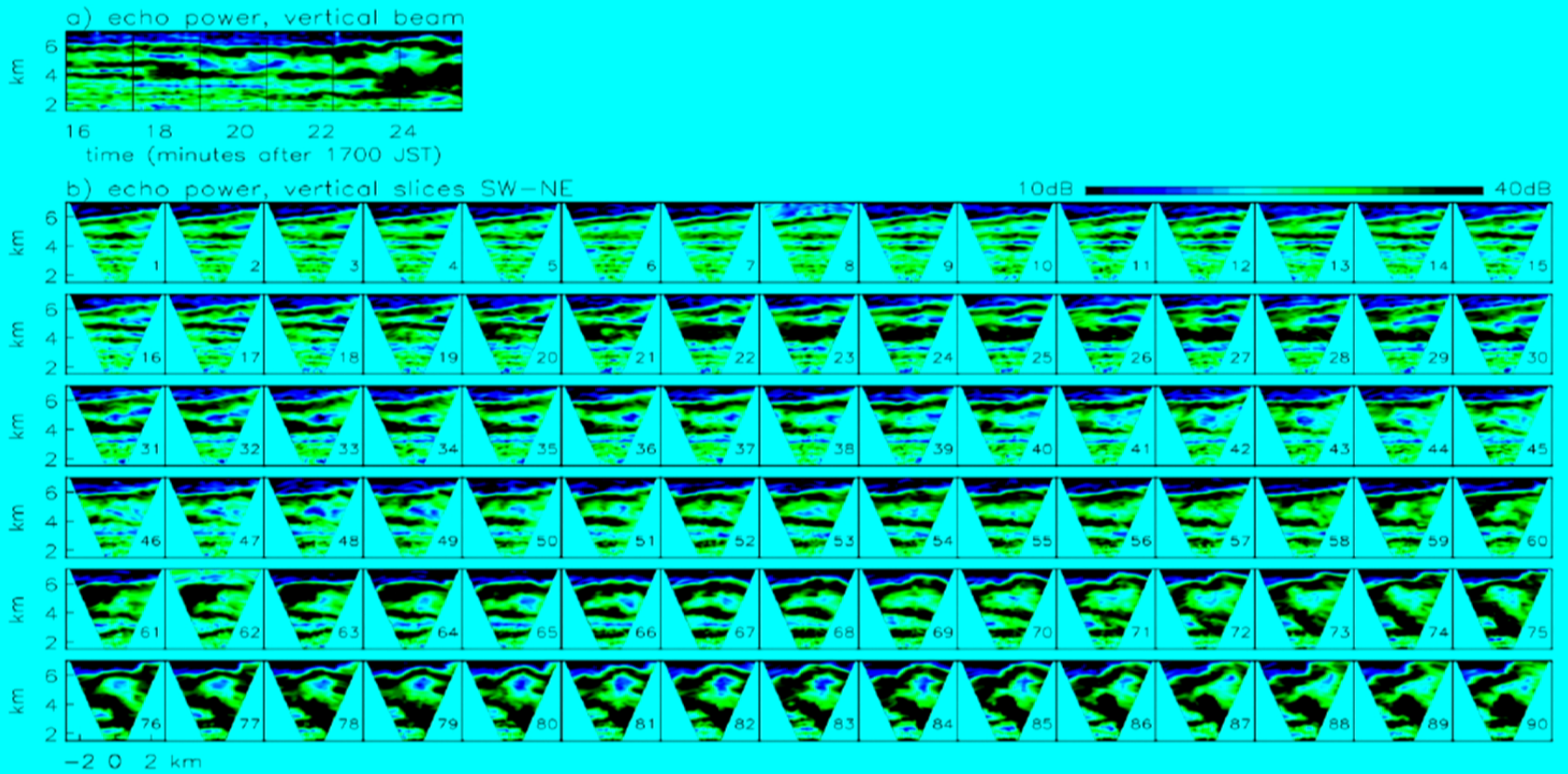
a) echo power, vertical beam



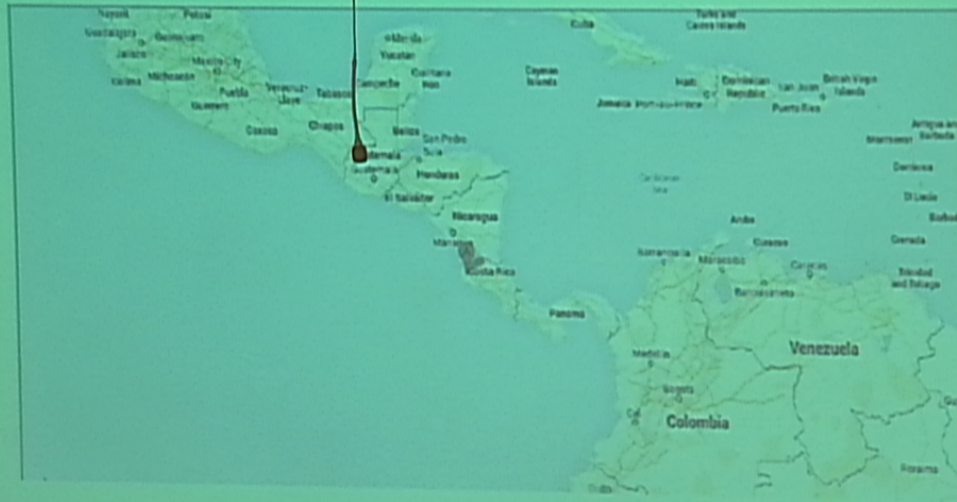
b) echo power, vertical slices SW-NE



MU Radar



Welcome to Costa Rica's radar

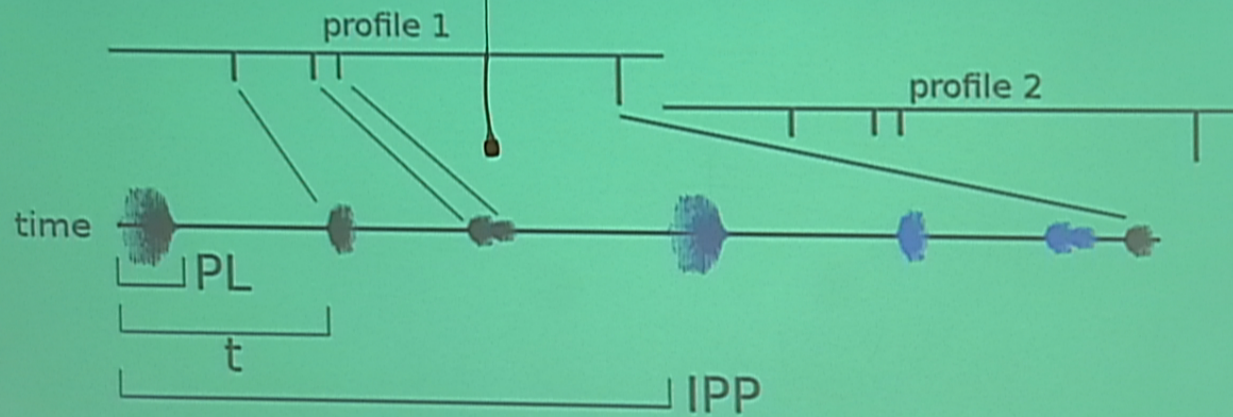


MST radar look

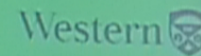


Back to Radars

Radar basics



Pulse Length
Inter Pulse Period



Back to Radars

Radar equations

Distance to target

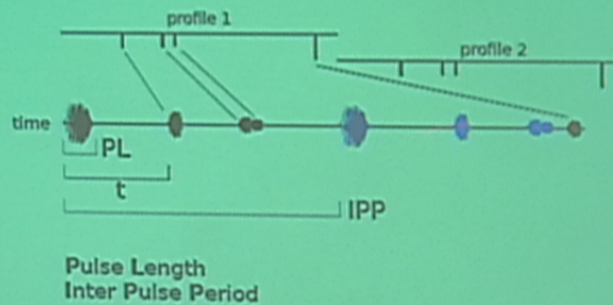
$$R = \frac{c_0 t}{2}$$

Maximum unambiguous distance

$$R_{amb} = \frac{c_0 \cdot IPP}{2}$$

Radar resolution

$$R_{res} = \frac{c_0 \cdot PL}{2}$$



Ideal conditions

What are the ideal conditions to probe the atmosphere?

An ideal method to probe the atmosphere would require to have, high vertical resolution (meters) and high time resolution (minutes).
Radars can provide both under adequate conditions.

- TIME: With a PRF of 3 kHz the time factor should be covered.
- HEIGHT:
 - τ : Resolution of 1 km.
 - Small BW: Resolution of 600 m [Canada].
 - Large BW: Resolution of 60 m [Costa Rica].

How do we do this?

Atmospheric physics takes care of the convolution.

$$T_p(t) * P(t) = R(t) + \epsilon$$

And we take care of the deconvolution. It can be done in the time domain, but there is a nice trick in the frequency domain that we can use.

$$T_p(t) * P(t) \xrightarrow{\mathcal{F}} T_p(f) \cdot P(f) = \mathcal{F}(R(t) + \epsilon)$$

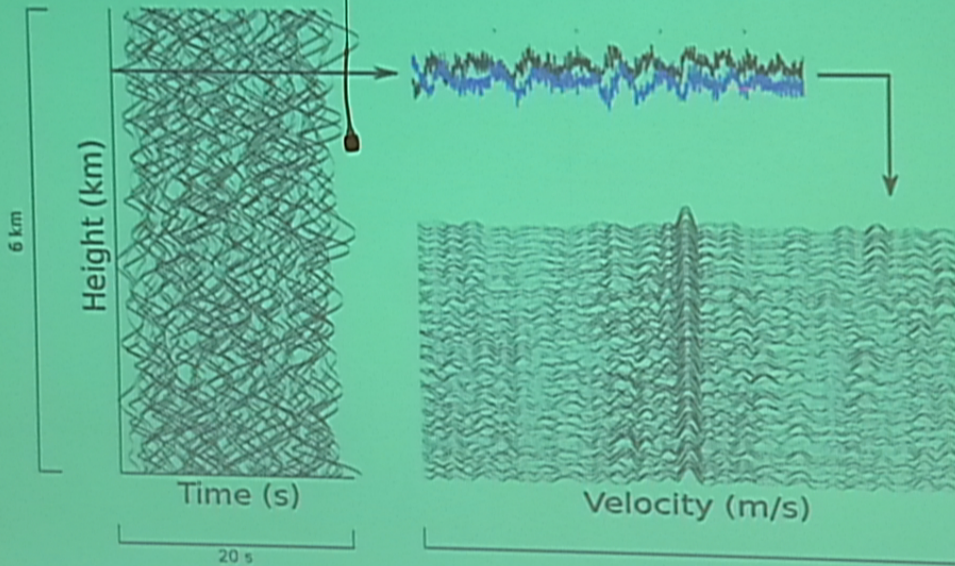
If the noise factor is small enough we can calculate the profile.

$$P(f) = \frac{\mathcal{F}(R(t) + \epsilon)}{T(f)}$$

Western

Deconvolution process

One more step for the energy information



Here is where HPC comes in!

We gather 60032 sweeps of the atmosphere per measurement.

$$60032 \text{ sweeps} \frac{1 \text{ measurement}}{64 \text{ sweeps}} = 938 \text{ profiles}$$

$$60032 \text{ sweeps} \frac{1 \text{ s}}{3071.76 \text{ sweeps}} = 19.54 \text{ s}$$

Under optimal conditions we could gather more than 4×10^6 profiles per day. This becomes 6.5 Gb of RAW data. After the deconvolution process and the height analysis this becomes more than 50 Gb.

Here is where HPC comes in!

We gather 60032 sweeps of the atmosphere per measurement.

$$60032 \text{ sweeps} \frac{1 \text{ measurement}}{64 \text{ sweeps}} = 938 \text{ profiles}$$

$$60032 \text{ sweeps} \frac{1 \text{ s}}{3071.76 \text{ sweeps}} = 19.54 \text{ s}$$

Under optimal conditions we could gather more than 4×10^6 profiles per day. This becomes 6.5 Gb of RAW data. After the deconvolution process and the height analysis this becomes more than 50 Gb.

HPC

Spectral estimation and long data sets

Fourier transform is very fast now days. Even for very long data series it will take just a short time to process. What consumes a lot of time is processing the signals after zero padding and using alternative methods for spectral estimation.

$$S_F = w^\dagger R w$$

$$S_C = \frac{R^{-1} w}{w^\dagger R^{-1} w}$$

The other area where the analysis can take a lot of time of when analysis long data series (months of data) to generate images of the atmosphere.



Identify the challenges

3 tasks require HPC [1/3]

- **Initial Processing.** When the raw data are captured they are gathered in a multicore PC. They are synchronized to Orca's storage. The raw format needs to be processed in order to deconvolve the signals and obtain profiles that can be analyzed. If the process change (often) or a test needs to be performed all the data will need to be reprocessed. **Large numbers of files - I/O intensive.**

West

Fourier Method - 10 files in 25.8 seconds - one core

17630	9.535	37.4	basic.py:155(fft)	FFTs
10	9.778	38.4	{cPickle.load}	LOADING FILES
10	24.494	96.1	spectralAnalysisMPI.py:36(ProcessFile)	
1	25.487	100	spectralAnalysisMPI.py:3(<module >)	

Capon's Method - 10 files in 54.5 minutes - one core

8250	9.064	0.3	basic.py 155(fft)	FFTs
11	9.799	0.3	{cPickle.load}	LOADING FILES
9380	66.113	2.0	defmatrix.py:817(getI)	Inverting
38758170	162.497	5.0	{method 'transpose'}	Transpose
9380	623.521	19.0	lags.py:167(CorrelationMatrix)	Correlation
23253020	905.741	27.6	defmatrix.py:327(__mul__)	Multiplication
9380	3253.49	99.1	getSpectra.py:7(Capon)	
1	3281.525	100	spectralAnalysisMPI-Capon.py:3	

Identify the challenges

Using MPI for one day analysis

Using 24 hours of moderate density data with 250 cores.

Fourier:

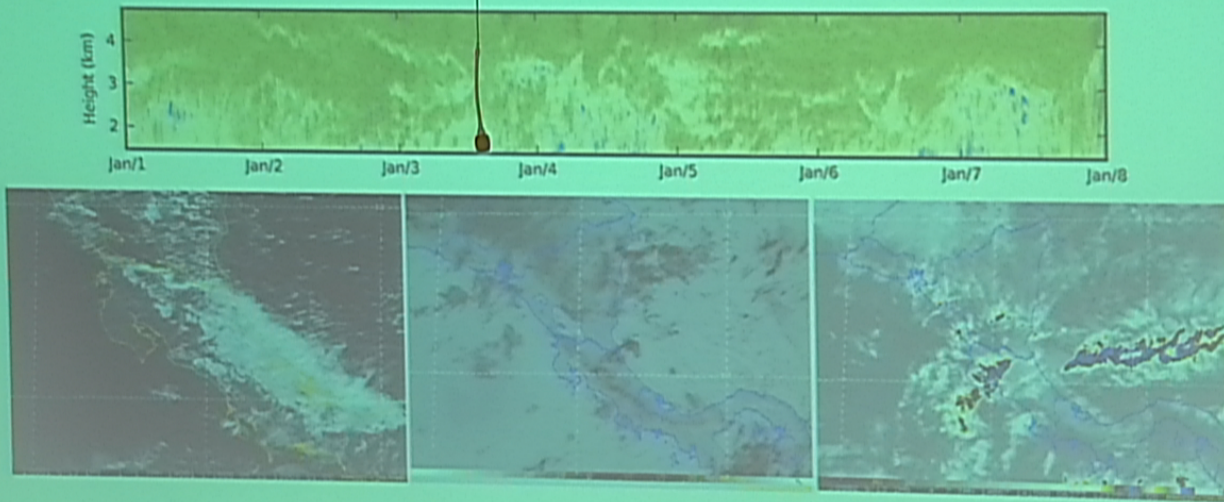
- cpu time: 1.3h
- elapsed time: 47s
- virtual memory: 84.6G

Capon:

- cpu time: 80.2h
- elapsed time: 1204s [approx 20 minutes]
- virtual memory: 184.0G

Products

Jan, 2014. 1st - 7th. cpu time: 16.4h. elapsed time: 269s. memory: 177.6G.



- Algorithm for waves and oscillations location.
- Make better use of HPC tools.
- Identify those sections where GP-GPU could be useful.
- Processing of long term data sets.

THE END

Questions?

