

HI source finding algorithms

Comparing the general purpose
Duchamp algorithm to a purpose built HI
source finding algorithm

Talk Outline

- Common elements of source finding algorithms
- The Duchamp algorithm
 - Algorithm
 - Strengths
 - Draw-backs
 - Improvements
- An alternative HI source finder algorithm
 - Key differences
 - Algorithm
- Preliminary work
- Conclusion

Common elements of general source finding algorithms

- Define/calculate detection and growing criteria
 - Thresholds or False Detection Rate
- Pre-condition data
- Scan through data and apply detection criterion
- Grow detections using growing criterion
- Merge detections
- Apply size criterion

The Duchamp Algorithm

Duchamp: Algorithm

- Pre-condition data (optional)
 - Blank pixel removal
 - Baseline removal using wavelet reconstruction
 - Define channels to ignore
- Wavelet reconstruction using a' trous wavelet procedure (priority)
 - OR
 - Smooth in frequency space
 - OR
 - Smooth spatially
- Set detection and growing criteria
 - User specified (priority)
 - OR
 - FDR or calculated from globally determined mean and rms values

Duchamp: Algorithm

- Raster scan data
 - Travel along planes or channels and apply detection criterion
 - If a voxel satisfies the detection criterion
 - Flag it
 - Check it's proximity to all previous detections and merge accordingly
 - Can be turned off for efficiency, but default is ON.
- Merge detections
 - Apply proximity test (**again**) to all detections
- Grow detections
- Merge detections again
- Apply size criterion
 - Can be done prior to first round of merging

Duchamp: Strengths

- A truly general source finding algorithm
- Makes minimal assumptions
- Extremely flexible source detection
- **IT EXISTS! and IT WORKS!**
- Output is feature rich

Duchamp: Draw-backs

- Efficiency decreases with the number of detections
 - Searching for faint sources is very inefficient
- Default is to run a merging routine every time a detection is made
 - Compared to **every!** previous detection
- Merging is carried out multiple times
- Size criterion is applied at the very end
 - Inefficient but necessary
- Global detection and growing criteria are used
 - Noise varies throughout the cube
 - Detect 'crud' in some regions, miss detections in others
- Side-lobes + high dynamic range = problems
- Multiple detections of single source

Duchamp: Improvements

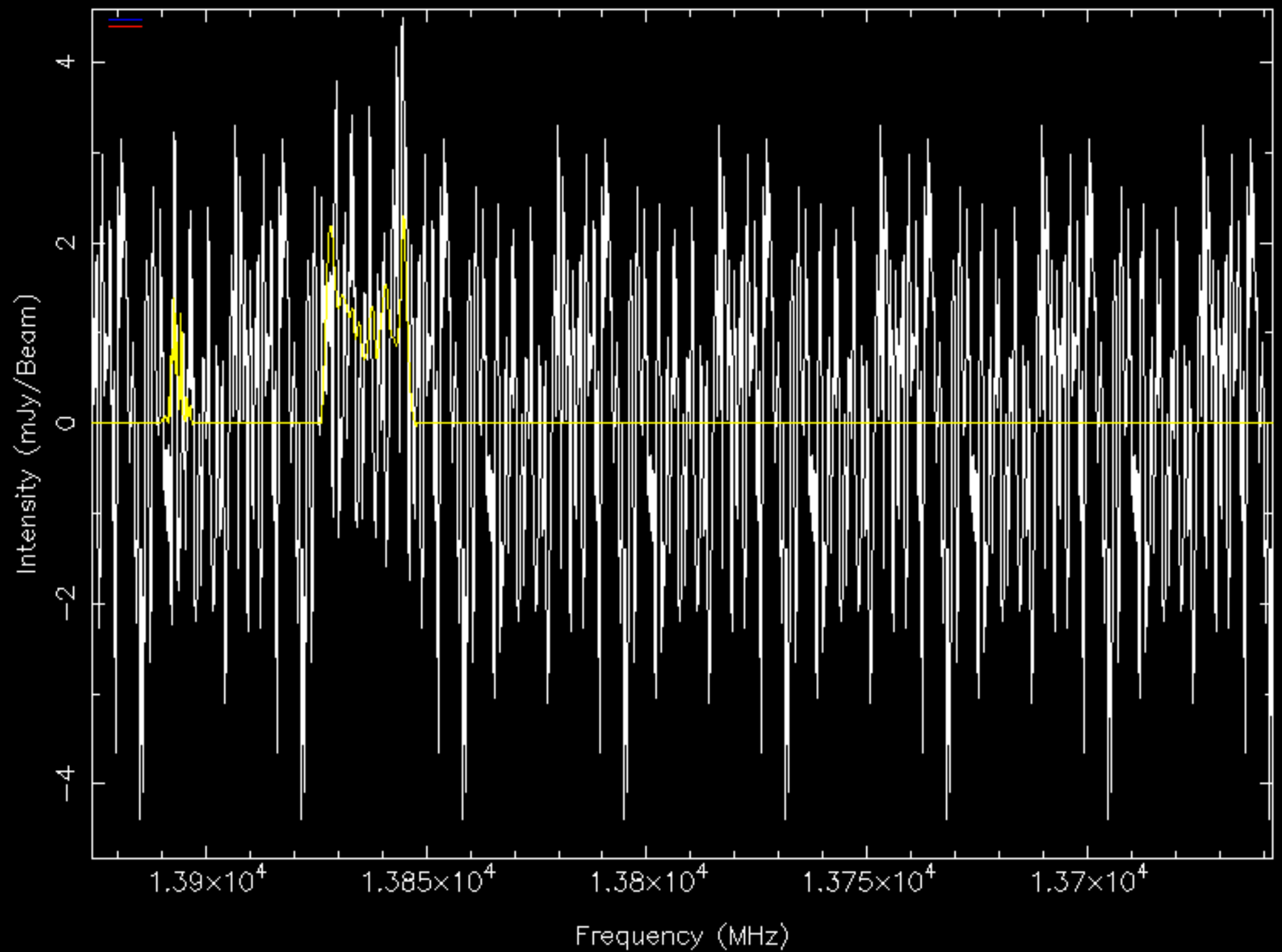
- Sub-sample channels when raster scanning
 - Sampling set to size criterion
 - Minimise detections that eventually would fail size criterion
- Define a data volume to check for previous detections
 - To be used when initial merging not turned off
- Grow detections, merge (just the once), apply size and detection threshold criteria
 - Apply growth criterion out to merging distance to fold in initial merging pass
- Use a local measure of noise
- Flag probable side-lobes
 - Sort detections according to intensity
 - Search through detections and for all objects within a moving, adaptive window (within intensity range of side-lobes), if within location of side-lobes flag it

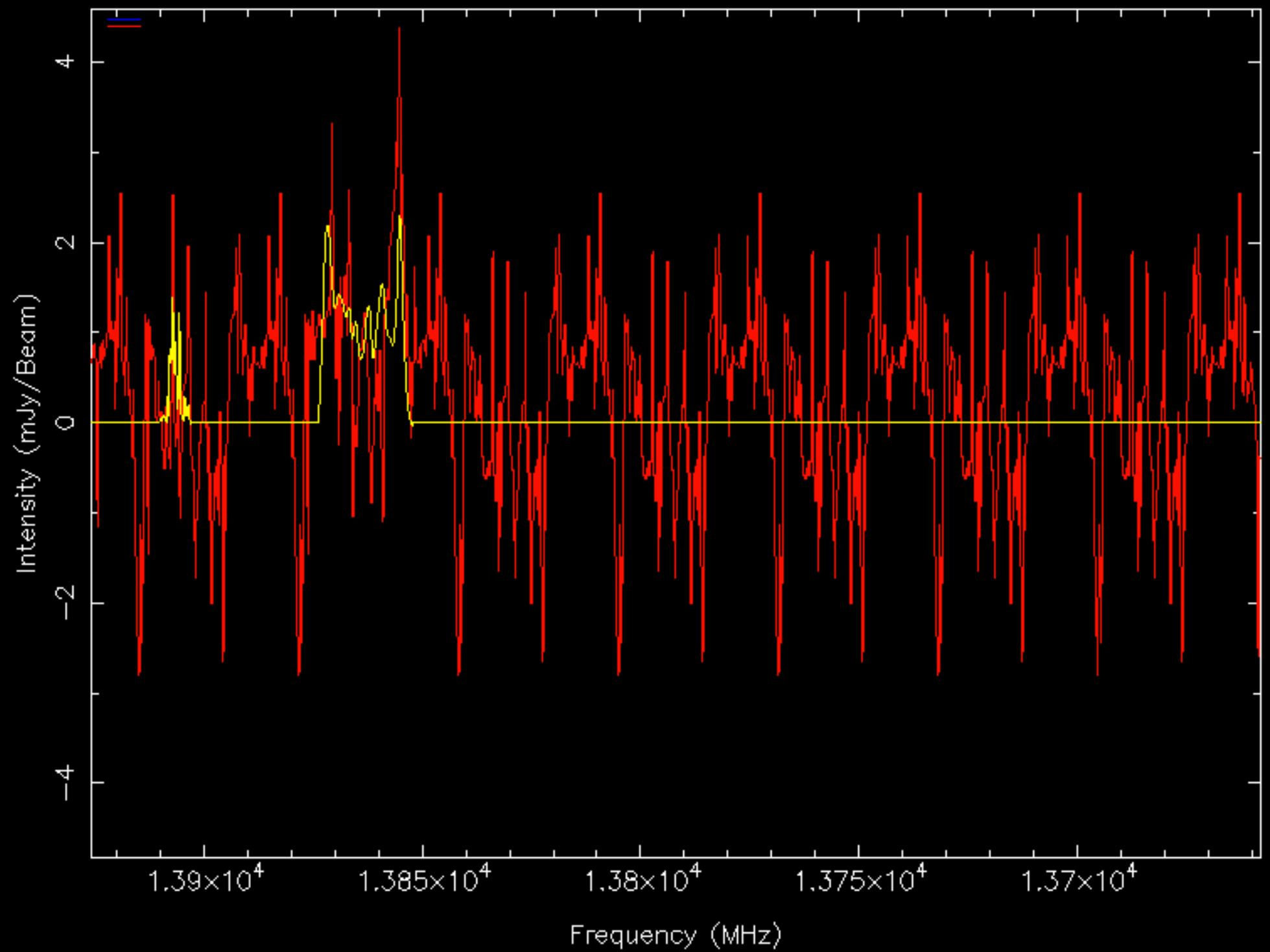


A purpose built HI source finder algorithm

Key differences

- Treat datacube as a set of spectra rather than a collection of voxels
- Use shape information rather than a detection threshold
 - Can potentially detect faint objects that a detection threshold would miss
 - Recover 'true' extent of source compared to using growth threshold
- Implicit is the assumption that every detection has a discernible shape
- Assume that we have a well defined psf





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Specific HI source finding algorithm

- Divide data cube amongst GPUs
- Sub-sample the data cube
- For a given spectrum
 - Pre-condition using iterative median smoothing
 - Use wavelet analysis to construct the noise spectrum + baselines and remove
 - Detect objects using shape information
 - Cross-correlation?
 - Wavelet analysis?
 - Gamma test? (Even if just for measure of noise in spectrum)

Specific HI source finding algorithm

- For each detection, scan neighbouring positions in spiral pattern to determine the volume containing the detection
 - Have a frequency range to process for neighbours
 - Well-known (**and SOLVED**) mouse navigating a maze problem
 - The solution provides a 'shrink-wrapped' volume
- Merge detections
- Merge GPU results
- Apply size criterion
 - May have been incorporated earlier
- Flag probable side-lobes

Preliminary work

- Prototyping iterative median smoothing as a pre-conditioner
- Using the WSRT simulated datacube
- Comparing to performance of Hanning filtering
- Results
 - Quantitatively, residuals cf. input spectrum are reduced by ~20-40%
 - Comparable to Hanning filtering, but doesn't add/remove structure in the cases where Hanning filtering does

Conclusion

- Duchamp is a great general purpose source finder
- The efficiency of Duchamp could be improved
- Side lobes are ignored by Duchamp
- Proposing to treat datacube as a set of spectra and use shape information to find HI sources
- Development underway