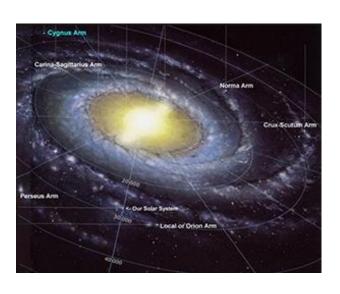




ViaLactea



The Milky Way as a Star Formation Engine







ViaLactea @ INAF







No	Name			Short name	Country		
1	ISTITUTO NAZIONAL	E DI ASTROFISICA		INAF	Italy		
2	UNIVERSITY OF LEE	DS		UNIVLEEDS	United Kingdom		
3	MAX PLANCK GESEL WISSENSCHAFTEN B	LSCHAFT ZUR FOER E.V.	DERUNG DER	MPG Germany			
4	MAGYAR TUDOMAN' AUTOMATIZALASI KU	YOS AKADEMIA SZAN JTATO INTEZET	SZTAKI	Hungary			
5	CARDIFF UNIVERSIT	Υ		CU	United Kingdom		
6	UNIVERSITE D'AIX M	ARSEILLE		AMU	France		
7	CENTRE NATIONAL I	DE LA RECHERCHE S	CIENTIFIQUE	CNRS	France		
8	NAGOYA UNIVERSIT	Υ		UON	Japan		
9	THE UNIVERSITY OF	EXETER		UNEXE	United Kingdom		

OACT

OACN

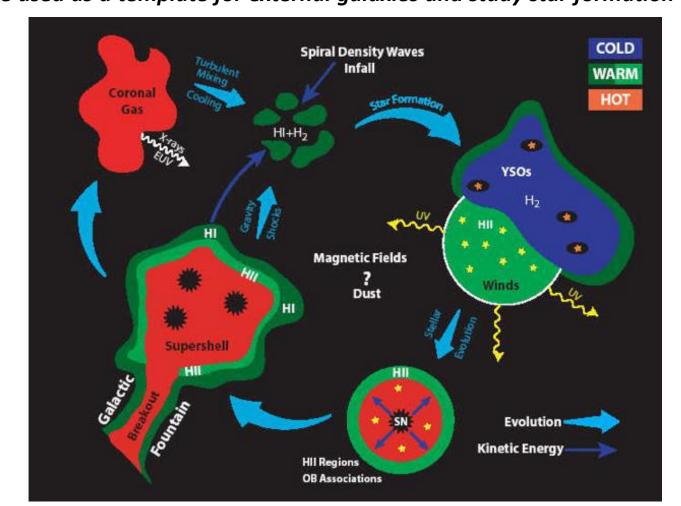


ViaLactea Mission



The goal is to exploit the combination of all the new-generation Infrared \rightarrow Radio surveys of the Galactic Plane from space missions and ground-based facilities, using a novel data and science analysis paradigm based on 3D visual analytics and data mining framework, to build and deliver a quantitative 3D model of our VIALACTEA Galaxy as a star formation engine that will be used as a template for external galaxies and study star formation across

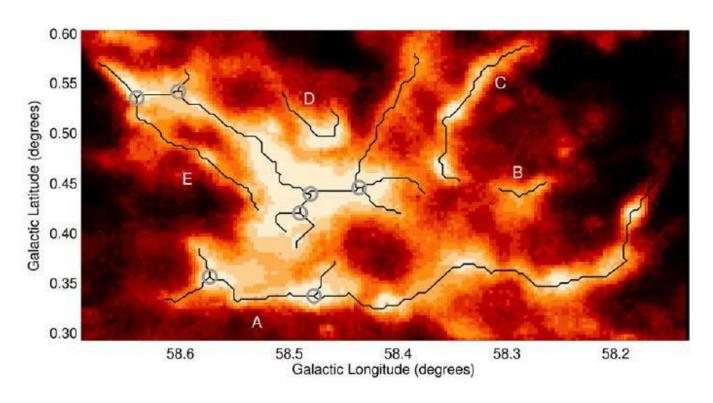
the cosmic time





ViaLactea – Goals





Identify the critical parameters that make star formation different:

- Spontaneous/triggered
- Filaments or no filaments ?
- Depending on the Galaxy position
- w.r.t. spiral arms
- etc.

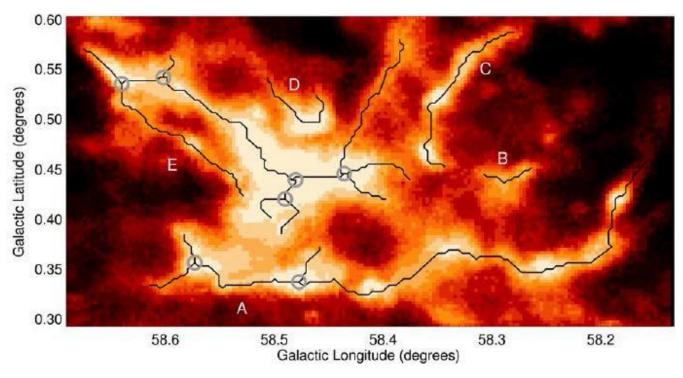


understand if & how the mix of the ingredients conspire to determine a global SF law



ViaLactea – Tasks



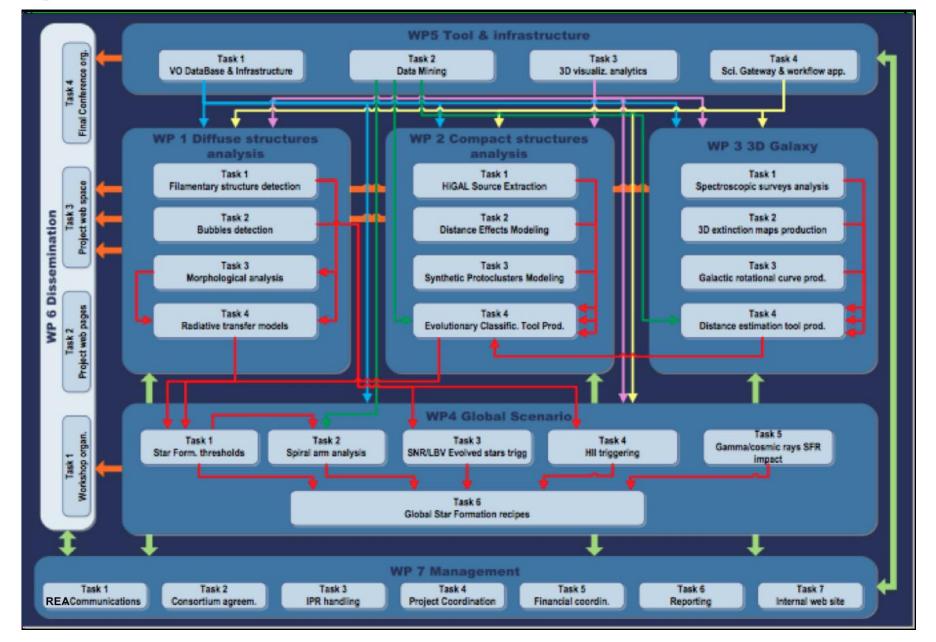


- ☐ Measure the star formation rate and history Galaxy-wide;
- ☐ Formation and fragmentation of Filamentary Molecular Clouds;
- ☐ Determining the relative importance of global vs local, spontaneous vs triggering, agents that give rise to star formation;
- ☐ Understanding star formation laws and the nature of thresholds as a function of ISM properties across a full range of galacto-centric radii metallicity and environmental conditions;
- ☐ Build bottom-up recipes and prescriptions useful for Xgal science.



ViaLactea – Project Flow

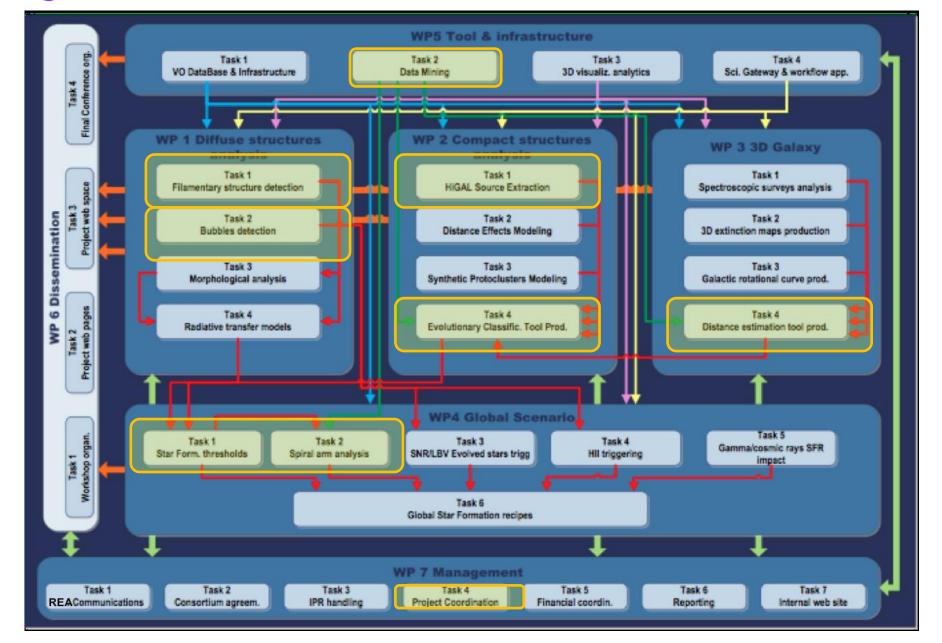






OACN - Data Mining







ViaLactea - Data



NIR Mid-IR

UKIDSS: J, H, K GLIMPSE: 3.6, 4.5, 5.8, 8.0 $[\mu]$

VISTA: K WISE: $[3 - 25] [\mu]$

MIPSGAL: 24 $[\mu]$

Far-IR

Hi-GAL: 70, 160, 250, 350, 500 [μ]

Sub-mm continuum

ATLASGAL: 870 $[\mu]$

JCMT: 870 [μ]

Near-IR, mid-IR and far-IR from Herschel, Spitzer and WISE

γ-ray imaging survey by AGILE and FermiLAT

Ground facilities: VISTA, JCMT, UKIRT and APEX, FCRAO, NANTEN2,

VLA, Parkes, Effelsberg

Molecular and atomic line surveys

GRS: 13CO at 21 cm

IGPS

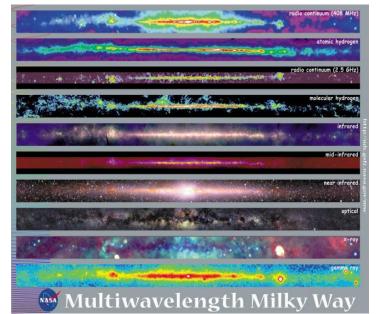
Radio continuum

CORNISH: 5 [GHz] (completed by Spitzer in the mid-IR and focusing on GLIMPSE region)

MAGPIS: [0.325 – 5] [GHz] (overlaps with MIPSGAL, GLIMPSE, ATLASGAL etc...)

Molecular Masers

Methanol Multi-Beam Survey (MMB): 5 cm methanol maser emission

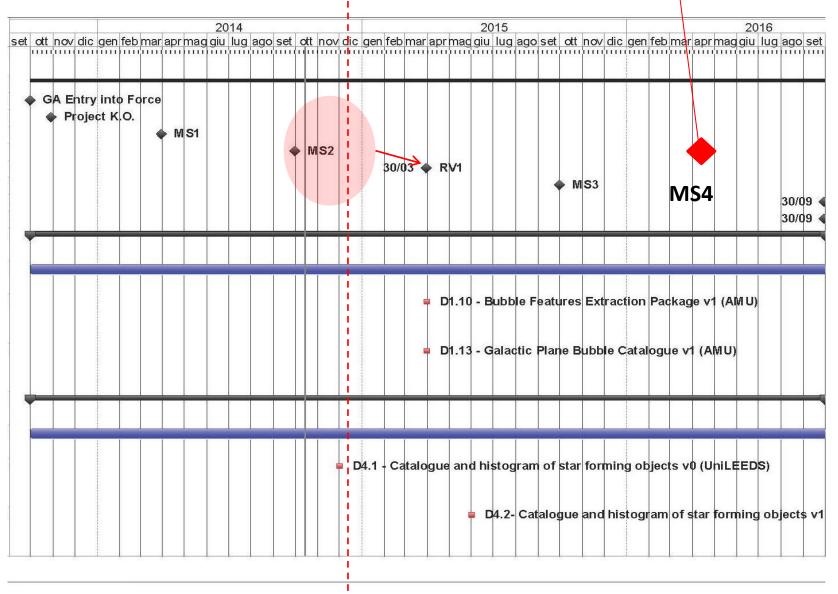




ViaLactea - Gantt

Hosted by OACN







ViaLactea – Science vs Technology



Most of the first FTE has been spent to find a common language among members...

How astronomers see astroinformaticians



How astroinformaticians see astronomers



...with doubtful but promising results



OACN – R&D Activity report



Compact source identification through band-merging

- ♦ V Q-FULLTREE method debug and preliminary test done;
- ♦ Q-FULLTREE scientific validation in progress;
- → Interaction with OACT+OATS infrastructures for integration and test in progress;

Filamentary structure Edge Detection

- ★ ✓ FilExSeC method design and development done;
- ★ ✓ FilExSeC method debug and preliminary test done;
- ★ ➤ FilExSeC scientific validation with IAPS in progress;
- → □ Started design & development of other methods;

Compact source distance estimation

- design and preliminary study in progress;
- → □ Started preliminary interaction with IAPS for Knowledge Base definition;

Star Forming source evolutionary classification

- Started design and preliminary study;
- → □ Started interaction with IAPS for Knowledge Base definition;

Bubble structures classification

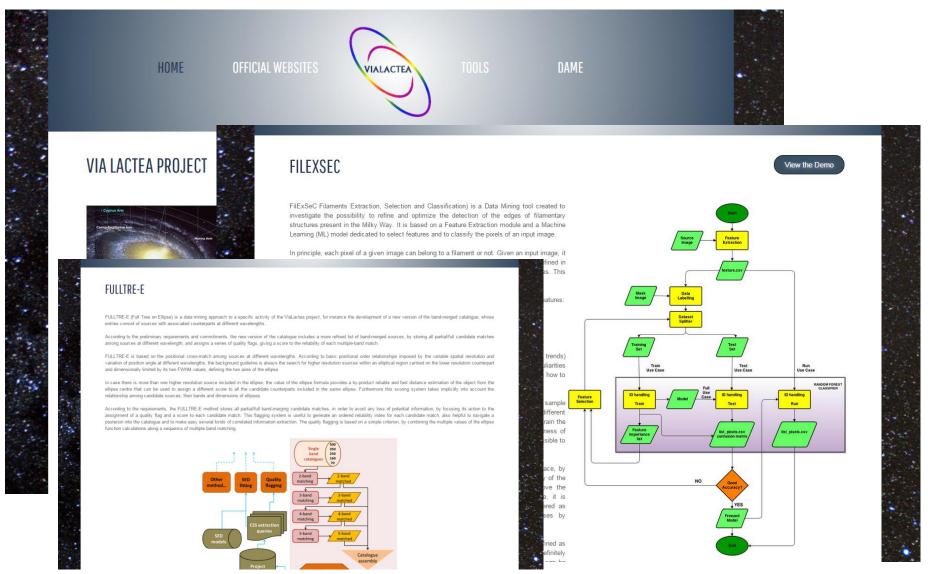
*
Started interaction with OACT for Knowledge Base definition;



ViaLactea – OACN web resource



http://pc169.na.astro.it:8080





ViaLactea official Science Gateway



http://muoni-server-01.oact.inaf.it:8081





VIALACTEA Project Science Gateway

Welcome Statistics Publications Help

VIALACTEA Project Science Gateway Welcome

Welcome to VIALACTEA Project Science Gateway!

The aim of the VIALACTEA project is to exploit the combination of all the new-generation Infared -> Radio surveys of the Galactic Plane from space missions and ground-based facilities, using a novel data and science analysis paradigm based on 3D visual analytics and data mining framework, to build and deliver a quantitative 3D model of our VIALACTEA Galaxy as a star formation engine that will be used as a template for external galaxies and study star formation across the cosmic time.

The main objectives of the project are:

- To boost the scientific exploitation of ESA missions space data by developing new and carefully tailored image processing tools to carry out detection and
 extraction of compact sources and filamentary structures, as well as more complex shape-finding to identify bubble-like features from large scale Galactic Plane
 imaging surveys both in the infrared continuum and in gas molecular lines.
- To combine in a VO-compatible and interoperable way the new-generation Galactic Plane surveys from space-borne missions and ground-based observatories, most
 of which are object of considerable European investments, to obtain a sub-arcminute resolution complete and homogeneous data coverage over the entire Galactic
 Plane from the infrared to the radio, that will extend the usage of already available space data and will multiply by many times the exploitation potential of the
 individual datasets.
- . To build and visualize a new 3D representation of the Milky Way Galaxy.
- To determine, map and visualize in 3D Galaxy-wide the quantitative relationship between the physical mechanisms locally responsible for the onset, triggering and regulation of star formation, and the end products of star formation measured by the Star Formation Rate and Star Formation Efficiency.
- To bring to a common forum the scientific astronomical expertise and the e-Science technological know-how of european-leading research groups to develop the
 next generation data analysis tools that will enable time-effective steradiant-scale science in Galactic star formation.

Participants of the project are:

ISTITUTO NAZIONALE DI ASTROFISICA (INAF), UNIVERSITY OF LEEDS, MAX PLANCK GESELLSCHAFT ZUR FOERDERUNG DER WISSENSCHAFTEN E.V., MAGYAR TUDOMANYOS AKADEMIA SZAMITASTECHIKAI ES AUTOMATIZALASI KUTATO INTEZET, CARDIFF UNIVERSITY, UNIVERSITÈ D'AIX MARSEILLE, CENTRE NATIONAL DE LA RECHERCHE SCIENTIFIQUE, NAGOYA UNIVERSITY, THE UNIVERSITY OF EXETER







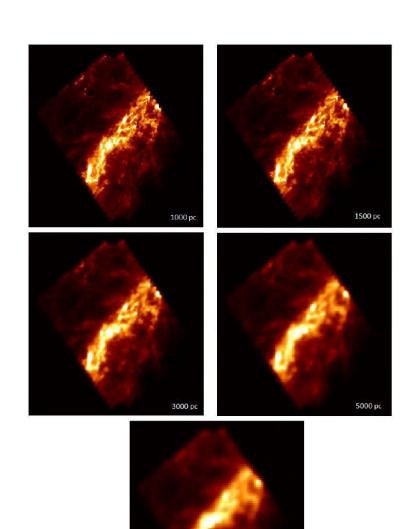


OACN – WP5 Activities



WP2 – Task 1
Band-merging
Q-FULLTREE

Design



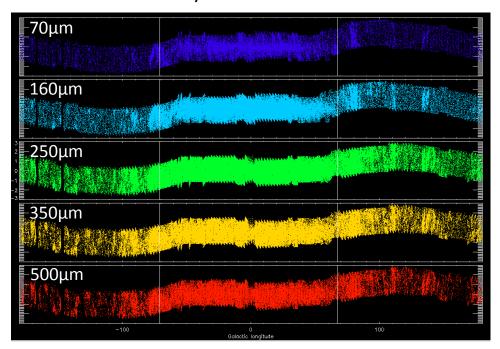


WP2 – task 1 - Bandmerging



Task 1: Compact Source Extraction and band-merging

- ☐ Hi-GAL Source extraction and photometry
- Band-merging with ancillary information (from near-IR to radio)



A first result from OACN of a band-merged catalogue using a data-mining approach has been implemented for the Herschel bands



The **source extraction** with CuTEx (*Molinari et al.,* **2010a**) has been run over the entire Galactic plane.

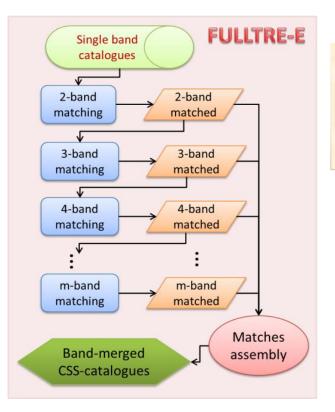
The -71° < *l* < 67.5° portion of the HERSHEL/Hi-GAL photometry lists should be band-merged, filtered and complemented with distances and ancillary photometry : MIPSGAL, UKIDSS, WISE, MSX; ATLASGAL, BGPS

- Captures and maintains multiple counterpart associations;
- Topological quality flagging;
- Ingested into a VO-like database so that complex queries are possible;
- Interfaced with Visualization tools;
- Massively based on multi-threading parallelization.



WP2 – task 1 - Bandmerging





The data mining approach, named FULLTRE-E (Full Tree on Ellipse), is based on the positional cross-match among sources at different wavelengths, by always respecting the order relationship imposed by spatial resolution.

the FULLTRE-E method stores all partial/full bandmerging candidate matches, in order to avoid any loss of potential information, by focusing its action to the assignment of quality flags and a score to each candidate match.

$$\frac{x^2}{a^2} + \frac{y^2}{b^2} \le 1 \rightarrow Ell(x_{bi}, x_{bj}) \le 1$$
 candidate match found

a and b are the two semi-axes of the ellipse (calculated upon the two given values of FWHM of the source, centre of the ellipse)

x and y are the coordinates of the higher resolution counterpart (opportunely corrected by the position angle variation)

 (x_{bi}, x_{bi}) a match between two-band sources



Bandmerging (FULLTRE-E Summary)



$$\frac{x^2}{a^2} + \frac{y^2}{b^2} \le 1 \rightarrow Ell(x_{bi}, x_{bj}) \le 1$$
 candidate match found

CSS =
$$\{x_{b1}, x_{b2}, ..., x_{bM}\} \equiv (x_{b1}, x_{b2}), (x_{b2}, x_{b3}), (x_{b1}, x_{b3}), ..., (x_{bM-1}, x_{bM}).$$

Confidence Level
$$\rightarrow CL(x_{bi}, x_{bj}) = 1 - Ell(x_{bi}, x_{bj})$$



scoring



number of elliptical matches NE Theoretical NE
$$\rightarrow$$
 TNE= $\binom{M}{2} = \frac{M!}{(2!)[(M-2)!]}$ CL terms

Merit Score
$$\rightarrow MS(CSS) = MS\{x_{b1}, x_{b2}, ..., x_{bM}\} = \frac{NE}{TNE} \sum_{i=1}^{NE} CL_i$$





QR = 3 to the min(MS1, MS2, MS3)

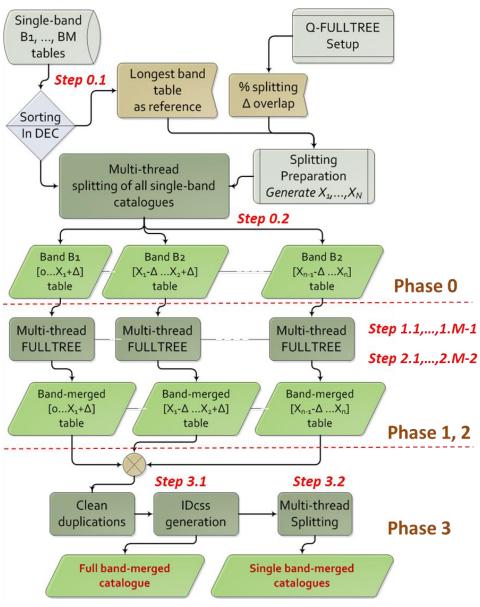
Quality Fitness
$$QF(CSS) = \frac{MS(CSS)}{\sum_{k=1}^{N} MS(CSS_k)}$$

$$\begin{cases} if \ exists \ a \ CSS_j(x_{bi}) \ such \ that \ QF(CSS_j) > 0.5, then \ AQF(CSS_j) = 1 \\ else \ AQF(CSS_j) = 0 \ \forall \ j = 1 ... \ N \end{cases}$$



Enhancing performance (Q-FULLTREE)





The wrapping system built around the original FULLTREE module is designed to improve the computing performance of the entire band-merging tool. Based on:

- ✓ Split %: the percentage of splitting used to generate sub-tables of the input singleband catalogues;
- ✓ △ overlap: the quantity of overlap (in arcmin) around which to replicate catalogue entries in the sub-tables;
- ✓ Pivot band: the reference band related to the longest single-band catalogue.

Worst gain in speedup: 200x (mostly higher)

5 bands:

on a bi-CPU 1.6GHz, 16 cores:

27 days \rightarrow 3.3 hours

On a quad-CPU 2.4GHz, 32 cores:

23 days → 1.3 hours

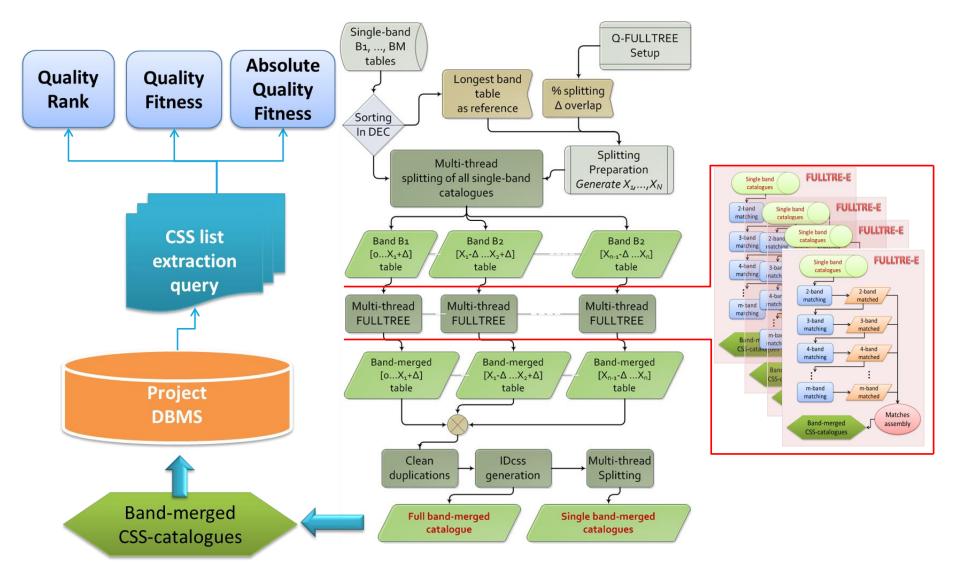
On CT cluster (1 CPU 2.4 GHz, 12 cores):

29 days \rightarrow 3,15 hours



Q-FULLTREE architecture

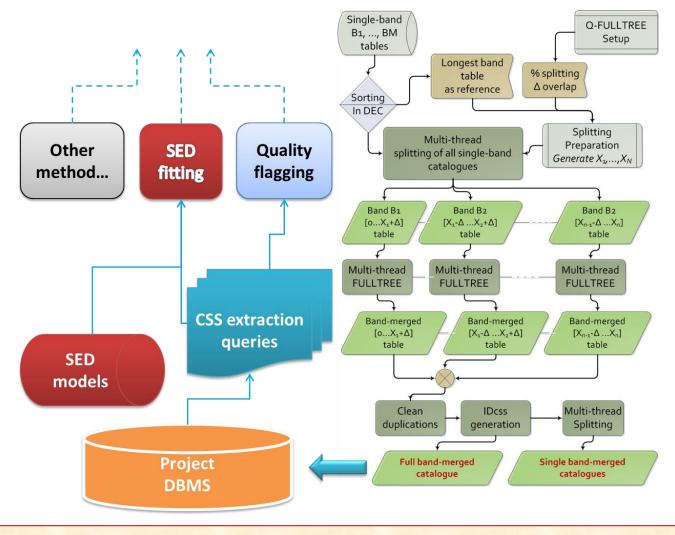






Bandmerging - connections



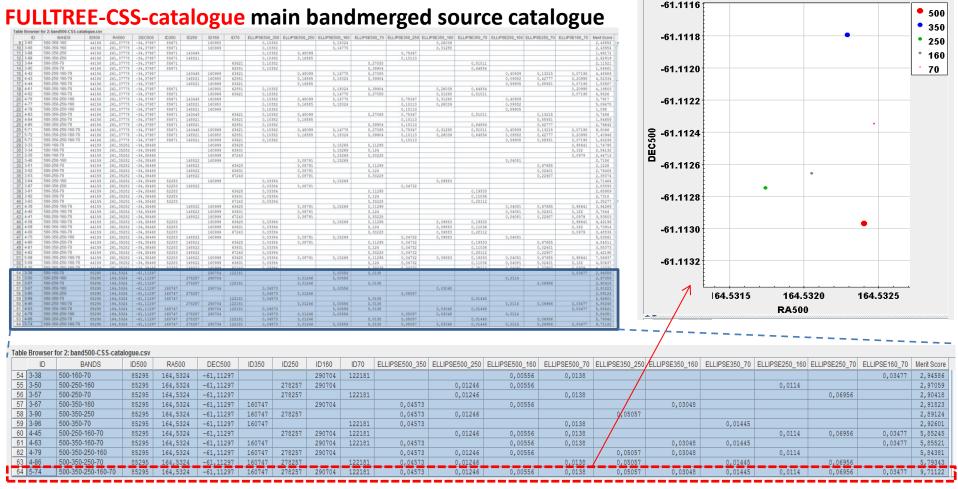


The information provided by the Q-FULLTREE output as a starting point of a workflow in which information coming from different analysis modules could be correlated to improve the overall knowledge



Bandmerging – ex. 1 (1 CSS)





Example of a full match (5 Hi-GAL bands, from 500μ to 70μ)

In the light-blue sub-table all the intermediate CSS found are shown



Bandmerging – ex. 1 (1 CSS)

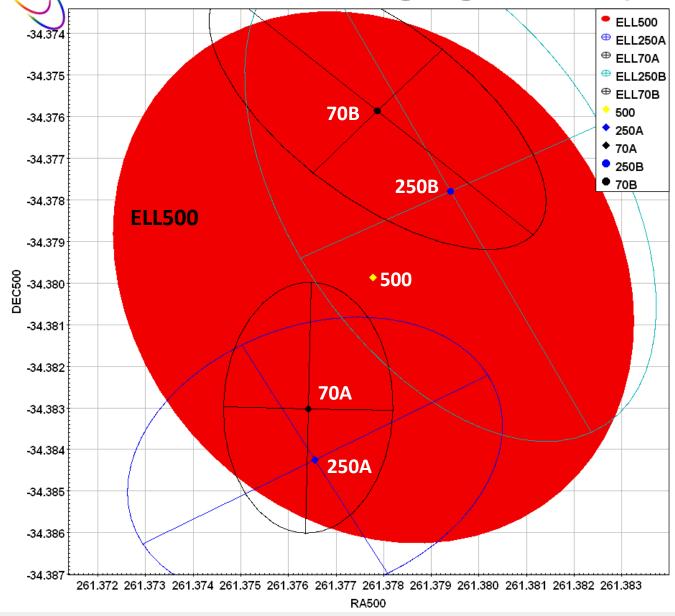


\odot		
ble Browser for 6: band500-ex2.csv		
ID BANDS ID500 ID350 ID250 5-74 500-350-250-160-70 85295 160747 278257	ID160 ID70 ELLIPSE500_350 ELLIPSE500_250 ELLIPSE500_160 ELLIPSE500_70 ELLIPSE300_250 ELLIPSE300_250 ELLIPSE300_170 ELLIPSE300_250 ELLIPSE300_250 ELLIPSE300_170 ELLIPSE	ELLIPSE350_160
	-61.107 -61.108	
MS = 9.71	-61.109	
-61.1116 • 500 • 350 • 250 • 160 -61.1120 • 70	-61.110	
-61.1122	-61.111	
8 -61.1124 -61.1126	160 250	
-61.1128	²⁵⁰ -61.113	© ELL500 ⊕ ELL350
-61.1130 -61.1132	-61.114	⊕ ELL250
64.5315 164.5320 164.5325 RA500	-61.115	⊕ ELL160 ⊕ ELL70
	-61.116 ELL500	→ 500
	-61.117	◆ 350 ◆ 250
	-61.118	160
		◆ 70

164.527 164.528 164.529 164.530 164.531 164.532 164.533 164.534 164.535 164.536 164.537 164.538 RA500

Bandmerging – ex.2 (3 CSS)





Example of 3 CSS with 3-band matches

 $CSS_1 = 500-250A-70A$

$$MS_1 = 2.12$$

 $QR_1 = 1$
 $QF_1 = 0.44$

 $CSS_2 = 500-250B-70A$

$$MS_2 = 0,65$$

 $QR_2 = 3$
 $QF_2 = 0.14$

250B ≠ **70A**

 $CSS_3 = 500-250B-70B$

$$MS_3 = 2,01$$

 $QR_3 = 2$
 $QF_3 = 0.42$

261,37778 -34.37987 143445 261.37656 -34.38425 209,2... 63621 261.37642 -34.38302 21,72 -91,2.. 44156 261,37778 148821 261,37941 -34,3778 116,9... 63621 261,37642 -34,38302 0,27085 0,64989 -34.37987 -34,37987 47,48 37,59 148821 261,37941 -34,3778 116,9... 62851 261.37787 -34,37587

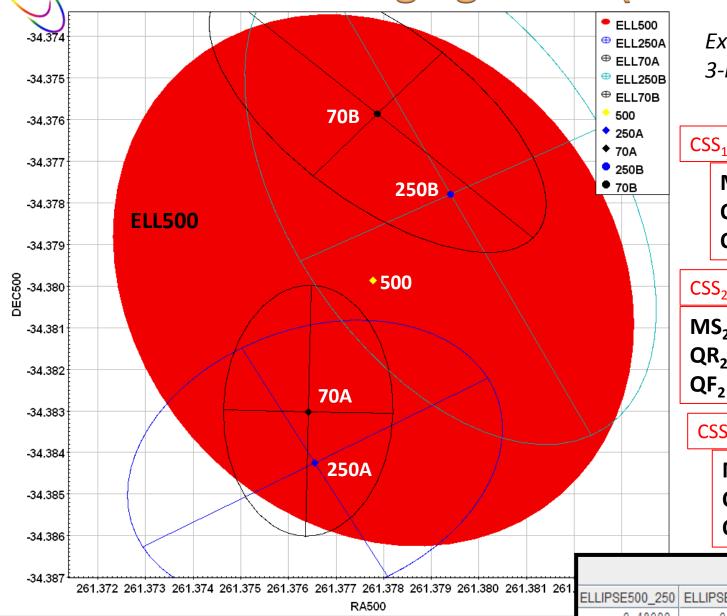
Bandmerging – ex.2 (3 CSS)

209,2... 63621

116,9... 63621

116,9... 62851





143445

148821

148821

261.37656

261,37941

-34,38425

-34,3778

261,37941 -34,3778 46,88

44156

44156

261.37778

261,37778

-34.37987

-34,37987 47,48

-34,37987 47,48

37,59

Example of 3 CSS with 3-band matches

 $CSS_1 = 500-250A-70A$

 $MS_1 = 2.12$ $QR_1 = 1$ $QF_1 = 0.44$

 $CSS_2 = 500-250B-70A$

 $MS_2 = 0.65$ $QR_2 = 3$

 $QF_2 = 0.14$

= 0.14

250B ≠ 70A

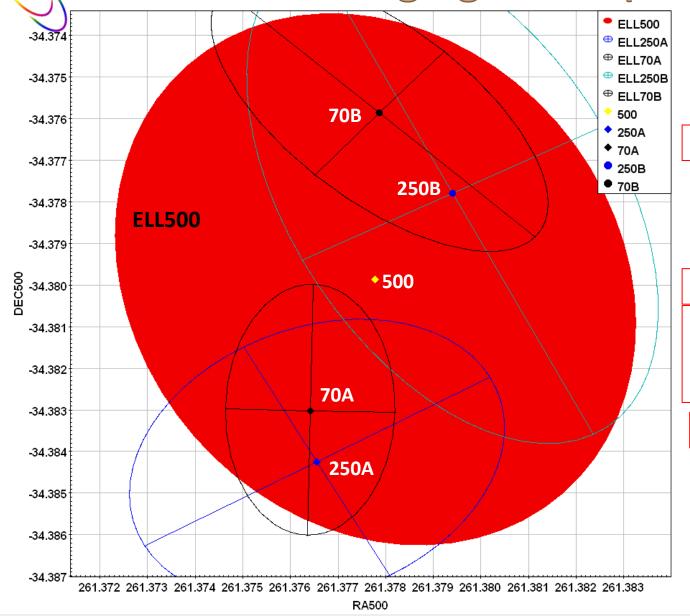
 $CSS_3 = 500-250B-70B$

 $MS_3 = 2,01$ $QR_3 = 2$ $QF_3 = 0.42$

ELLIPSE500_2	250	ELLIPSE500_70	ELLIPSE250_70	Merit Score
0,480	99	0,27085	0,13215	2,11601
0,165	85	0,27085		0,64989
0,165	85	0,39904	0,42777	2,00734
6				

Bandmerging – ex.2 (3 CSS)





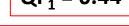
Example of 3 CSS with 3-band matches

 $CSS_1 = 500-250A-70A$

$$MS_1 = 2.12$$

 $QR_1 = 1$

$$QF_1 = 0.44$$



CSS₂ 50 250B-70A

 $MS_2 = 65$ $QR_2 = 3$

OE - 0 1/

 $QF_2 = 0.14$

250B ≠ **70A**

 $CSS_3 = 500-250B-70B$

 $MS_3 = 2,01$

 $QR_3 = 2$

 $QF_3 = 0.42$

AQF = 0

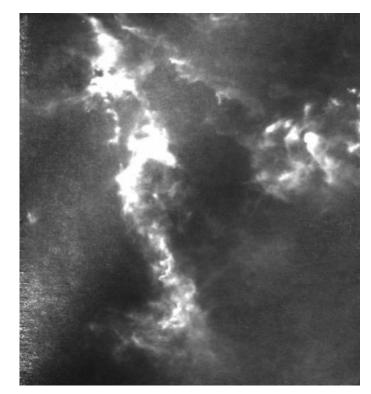
| SANDES | 10 |





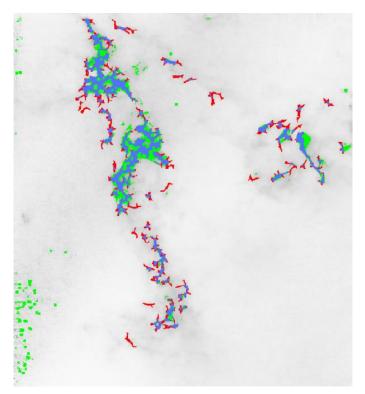
OACN – WP5 Activities

WP1 – Task 1 Filament Detection Optimization



FilExSeC

Design





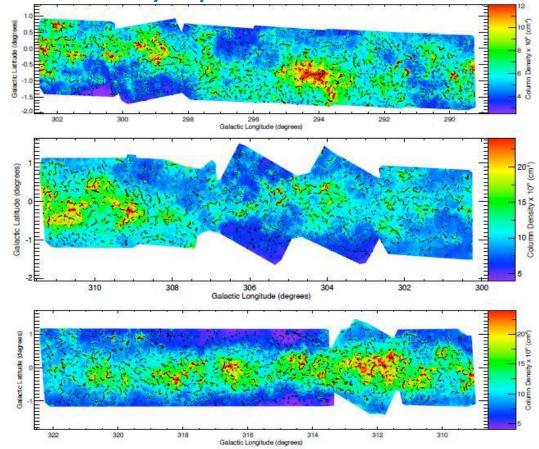
WP1 – task 1 – FilExSeC



Task 1: Filamentary structure detection

- ☐ Production of column density maps of entire galactic plane
- ☐ Automated filament extraction workflow for Hi-GAL survey

Column density maps with the identified filaments

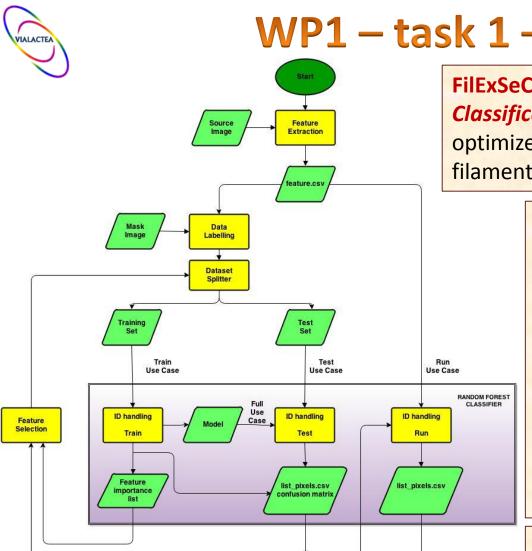


The filament extraction code was run on the column density maps covering the region between Galactic longitude 290° -- 320°, with different threshold levels equal to 2.5, 3. and 3.5 times the local standard deviation of the minimum eigenvalue (*Schisano et al.*, 2014)

OACN Data Mining goal:

- To refine the edges;
- ❖ To extend filament regions.

The right calculation of physical quantities related to filaments strongly depends on their dimensions, so the correct definition of edges is crucial.



Good Accuracy'

> Freezed Model

YES



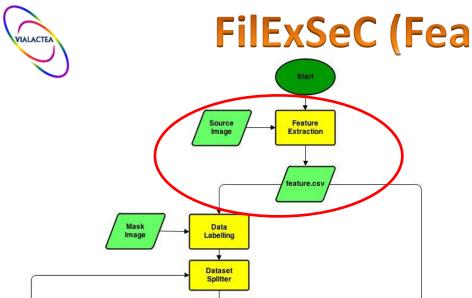


FilExSeC (Filaments Extraction, Selection and *Classification*), a data mining tool to refine and optimize the detection of the edges of filamentary structures.

The method consists in two main phases:

- **Feature Extraction**: a set of features depending by its neighbors associated to each pixel of the input image
- **Classification**: image pixels are classified as filamentary or background, by using a supervised Machine Learning method, trained by these features

A further phase, *Feature Selection*, finds the most relevant features. By reducing the initial data parameter space, it is possible indeed to improve the execution efficiency of the model, without affecting its performances.



ID handling

list pixels.csv

Good

Accuracy'

Model

YES

Case

ID handling

mportance

Feature

FilExSeC (Feature Extraction)

Run

ID handling

Run

list pixels.csv

RANDOM FOREST



Given an input image, it is possible to characterize each pixel by means of a set of features.

3 types of features are extracted for each pixel:

- Haralick features (Haralick, 1979);
- Haar-like features (Viola & Jones, 2001);
- Statistical features.

For most of the extracted features it is expected to have peculiar correlated values (or trends) for the pixels belonging to a filament, although hidden by background noise.

These peculiarities can be indeed used by a ML algorithm in order to learn how to discriminate the hidden correlation among filament pixels



FilExSeC (Features)



Haralick Features (1979)

- Based on co-occurrence matrix (GLCM)
- Element C_{i,j} represents, for a fixed distance and direction, the probability to have two pixels in the image at that distance, with grey level Z_i and Z_i, respectively

								1	2	3	4	5	6	7	8
1 (1	1	5	6	8	GLCM	1	1	2	0	0	1	0	0	0
	2	3	5	7	1		2	0	1 0	1	0	1	0	0	0
	4	5	7 (1	2		3	8	0	0	0	1	0	0	0
	8	5 (1	2	<u>)</u> 5		4	0	0	0	0	1	0	0	0
							5	_1	0	0	0	0	1	2	0
							6	0	0	0	0	0	0	0	1
							7	2	0	0	0	0	0	0	0
							8	0	0	0	0	1	0	0	0

Haralick features extracted from $C_{i,j}$ (number of pairs)

Contrast	$m = \sum_{i} \sum_{j} (i - j)^2 C_{i,j}$
Energy	$\sum_{i}\sum_{j}C_{i,j}^{2}$
Entropy	$-\sum_{i}\sum_{j}C_{i,j}\log C_{i,j}$
Correlation	$\frac{\sum_{i} \sum_{j} (i - \mu_{i})(j - \mu_{j}) C_{i,j}}{\sigma_{i} \sigma_{j}}$



Robert Haralick





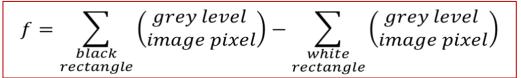
FilExSeC (Features)

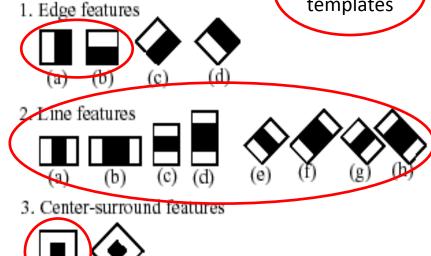


Paul Viola

<u>Haar-like Features (2001)</u>

- Each Haar-like variable involves 2 or 3 interconnected black and white rectangles (masks or templates)
- Values of each feature are obtained by sliding masks on the image and calculating:



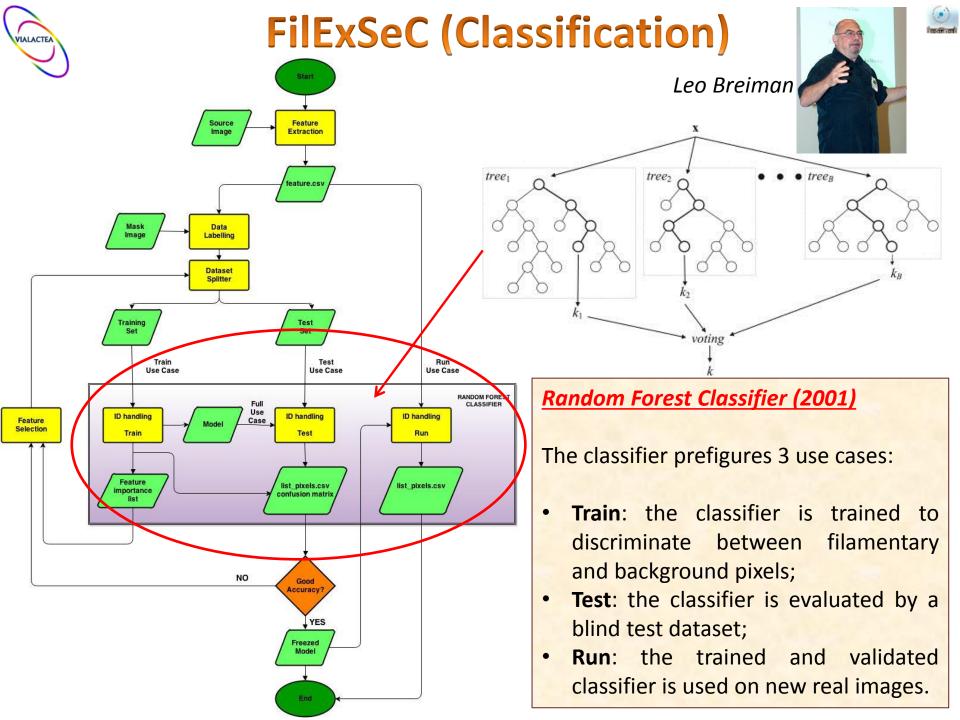


Statistical Features

For each pixel, the following features are calculated in a centered window:

- gradients (horizontal, vertical, main diagonal, secondary diagonal)
- Mean, standard deviation, skewness, kurtosis, entropy, range

Moreover, the pixel value is considered as a Statistical Feature too





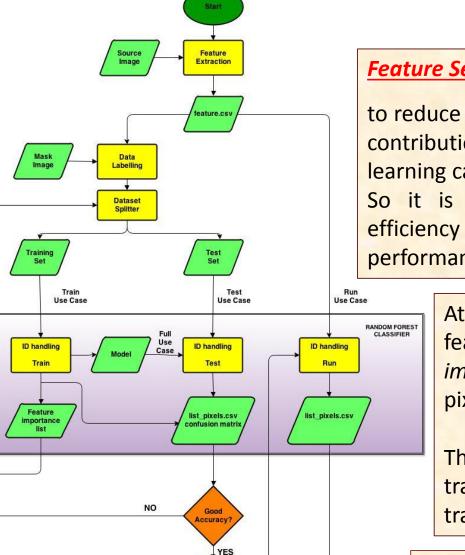
Feature

FilExSeC (Feature Selection)



Mark A. Hall





Model

Feature Selection (Backward Elimination 1999)

to reduce the parameter space, by weighting the contribution carried by each feature to the learning capability of the classifier.

So it is possible to improve the execution efficiency of the model, without affecting its performances.

At the end of this phase, a subset of features having higher weight (defined as *importance*) in recognizing filament pixels is considered.

This subset is then used to definitely train and test the classifier with new training and testing subsets.

Tests revealed that Haralick features are useless



FilExSeC (products)



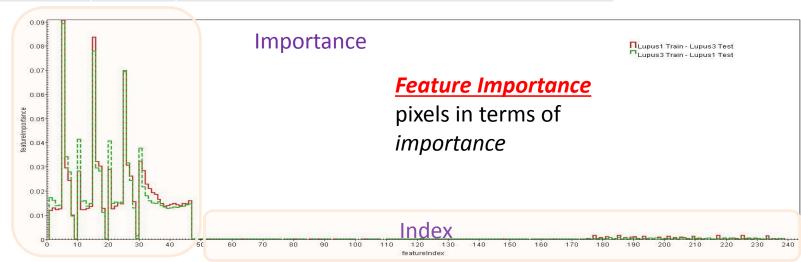
Main Output: A CSV table

Pixel	Statistical	Haar-like	Haralick	Pixel	Output
identification	features	features	features	Value	1 filament
id, row, column					0 background

Confirmed Filament Pixels	CFP	pixels correctly recognized as belonging to a filament
Undetected Filament Pixels	UFP	pixels of filaments classified as background
New Filament Pixels	NFP	background pixels classified as belonging to a filament
Confirmed Background Pixels	СВР	background pixels correctly recognized

Confusion Matrix:

Pixels grouped in 4 categories





FilExSeC – Testing

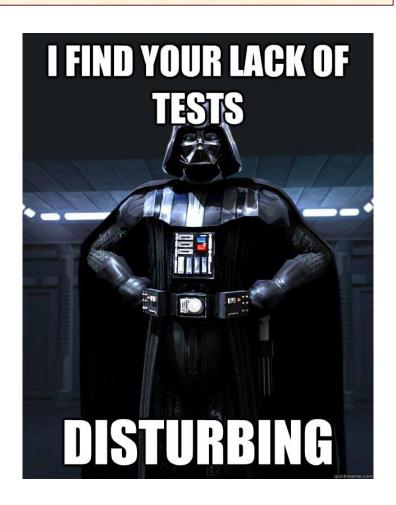


In order to evaluate FilExSeC, several tests have been performed. They have been useful to:

- evaluate the robustness of the method;
- evaluate the performances of the algorithm;
- optimize the parameters settings.

Test have been performed by:

- Varying the number and type of features:
 - Haralick yes/no
 - Pixel value yes/no
 - Stats yes/no
 - Different Haar-like templates
 - Different settings of Haralick, Haar-like and Statistical windows
- Varying the configuration of the classifier
 - Different number of trees of the Random Forest (1000 or 10000)
- Using different datasets:
 - Lupus region;
 - Hi-GAL strips;
- Using different ratios for Train and Test set





FilExSeC – Tests Summary



EXP	DATA FEATURE EXTRACTION						RANDOM FOREST				FI	LAMENT STATISTIC	CS	BACKGROUND STATISTICS				
	TRAIN		TEST		HARALICK	HAA	AR-LIKE	STAT		SE.	TUP		PURITY	COMPLETENESS	DICE	PURITY	COMPLETENESS	DICE
ID	name	%	name	%	windows	template	windows	windows	trees	max depth	min split	min leaf	%	%	%	%	%	%
T1a	lupusIII	15	lupusIII	85	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	84	62	72	99	99	99
T1b	lupusIII	30	lupusIII	70	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	88	65	75	99	100	99
T1c	lupusIII	60	lupusIII	40	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	92	70	79	99	100	99
T2	lupusIII	100	lupusi	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	61	67	64	99	98	99
ТЗа	lupusi	100	lupusIII	100	5x5, 7x7, 9x9		9x9 to 24x24		1000	none	2	1	48	29	36	98	99	98
тзь	lupusIII	100	lupusi	100	5x5, 7x7, 9x9		9x9 to 24x24		1000	none	2	1	52	24	33	97	99	98
T4	lupusIII	100	lupusi	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	10000	none	2	1	60	67	64	99	98	99
T5	lupusi	100	lupusIII	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	77	50	61	98	100	99
T6a	Hi-GAL 1+2	100	Hi-GAL 3+4	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	83,24	21,93	34,72	98,32	99,90	99,11
T6b	Hi-GAL 1+3	100	Hi-GAL 2+4	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	82,85	20,71	33,14	98,11	99,90	99,00
T6c	Hi-GAL 2+4	100	Hi-GAL 1+3	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	59,00	38,53	46,62	98,18	99,20	98,69
T6d	Hi-GAL 3+4	100	Hi-GAL 1+2	100	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	59,24	36,87	45,45	97,98	99,18	98,58
T7a	lupusi	100	lupusIII	100			9x9 to 24x24	3x3, 5x5, 7x7, 9x9	1000	none	2	1	81,32	54,66	65,37	98,54	99,59	99,07
T7b	lupusi	100	lupusIII	100			2x2 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	82,23	55,64	66,37	98,58	99,61	99,09
T7c	lupusi	100	lupusIII	100			2x2 to 24x24	3x3, 5x5, 7x7, 9x9	1000	none	2	1	82,73	56,94	67,45	98,62	99,61	99,11
T8a	Hi-GAL	60	Hi-GAL	40	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	88	38	53	98	100	99
T8b	Hi-GAL	80	Hi-GAL	20	5x5, 7x7, 9x9		9x9 to 24x24	5x5, 7x7, 9x9	1000	none	2	1	90	39	55	98	100	99
T9a	Hi-GAL	80	Hi-GAL	20			2x2 to 24x24	3x3, 5x5, 7x7, 9x9	1000	none	2	1	88,41	43,00	57,86	98,49	99,85	99,16
T9b	Hi-GAL	80	Hi-GAL	20			1x1 to 24x24	3x3, 5x5, 7x7, 9x9	1000	none	2	1	88,52	43,62	58,45	98,49	99,85	99,16
Т9с	Hi-GAL	80	Hi-GAL	20			2x2 to 24x24	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	89,59	44,98	59,89	98,52	99,86	99,19
T9d	Hi-GAL	80	Hi-GAL	20			1x1 to 24x24	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	89,39	44,88	54,80	98,54	99,86	99,19
T10a	Hi-GAL 1+2	100	Hi-GAL 3+4	100		all	*	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	84,88	43,75	57,74	98,78	99,83	99,30
T10b	Hi-GAL 1+3	100	Hi-GAL 2+4	100		all	*	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	85,12	42,63	56,81	98,63	99,82	99,22
T10c	Hi-GAL 2+4	100	Hi-GAL 1+3	100		all	*	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	73,67	51,82	60,85	98,57	99,45	99,01
T10d	Hi-GAL 3+4	100	Hi-GAL 1+2	100		all	*	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	72,62	49,64	58,97	98,39	99,40	98,89
T11	lupusi	100	lupusIII	100		all	*	1x1, 3x3, 5x5, 7x7, 9x9	1000	none	2	1	85,41	63,88	73,09	98,83	99,65	99,24
T12a	HI-GAL	100	HI-GAL289	100		all	*	1x1, 3x3, 5x5, 7x7, 9x10	1000	none	2	1	71,08	54,14	61,47	96,80	98,44	97,61
T12b	HI-GAL	100	HI-GAL300	100		all	*	1x1, 3x3, 5x5, 7x7, 9x10	1000	none	2	1	52,06	71,14	60,12	98,06	95,7	96,86
T12c	HI-GAL	100	HI-GAL310	100		all	*	1x1, 3x3, 5x5, 7x7, 9x10	1000	none	2	1	52,21	72,70	60,78	98,05	95,38	96,69



FilExSeC – Tests parameters



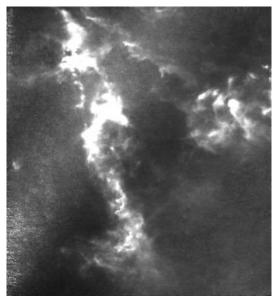
Туре	Parameters			Features	Max. Num.	
	Name	Template	Values			
	Black rectangle		from 2x2 to			
	dimensions		24x24			
	Black rectangle		from 2x4 to			
	dimensions		12x24	Difference between		
Haar-like	Number of			"black" and "white"	158	
	black		1-2	rectangles		
	rectangles					
	Black rectangle					
	dimensions		from 1 to 24			
	$ \vec{d} $ =1,2,3,4			Contract Energy	192	
Haralick	directions = 0°, 45°, 90°, 135°			Contrast, Energy, Entropy, Correlation		
	windows = 5x5 – 7x7 – 9x9			Entropy, correlation		
Statistical	windows = 3x3 – 5x5 – 7x7 – 9x9			Gradients (vertical,		
				horizontal, diagonal)	41	
				Mean – Std		
Statistical				Skewness - Kurtosis		
				Entropy – Range		
	windows = 1x1			Pixel Value		



FilExSeC – Tests dataset (1)



Lupus I

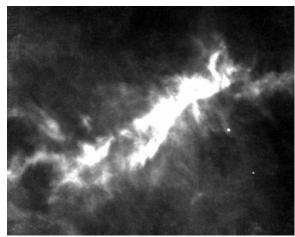


Lupus I Mask

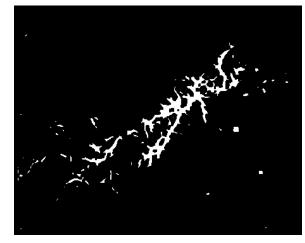


Lupus region
Sub-region I & III

Lupus III



Lupus III Mask



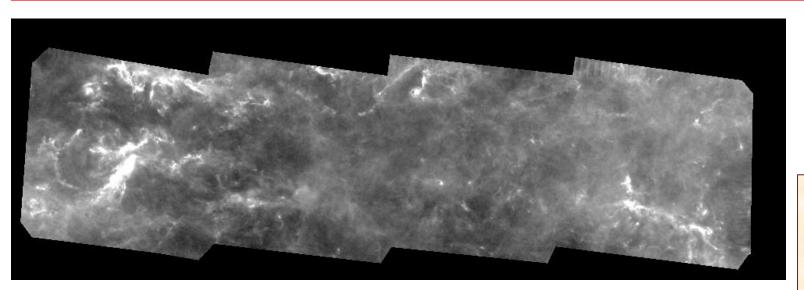
<u>Tests</u>



FilExSeC – Tests dataset (2)

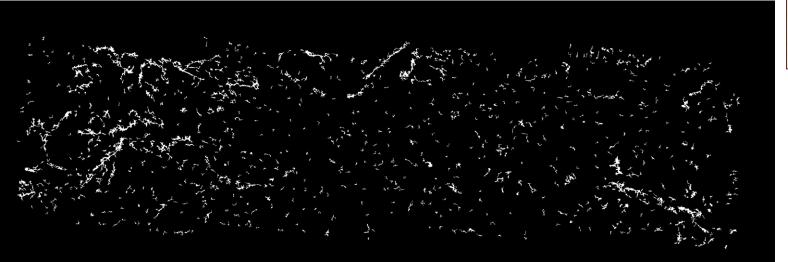


Column density map (250μ) of Hi-GAL 224-300 deg region: 2973x1001 pixels



Tests T6 - T8 - T9 T10

T12 (as Train set)

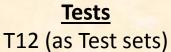


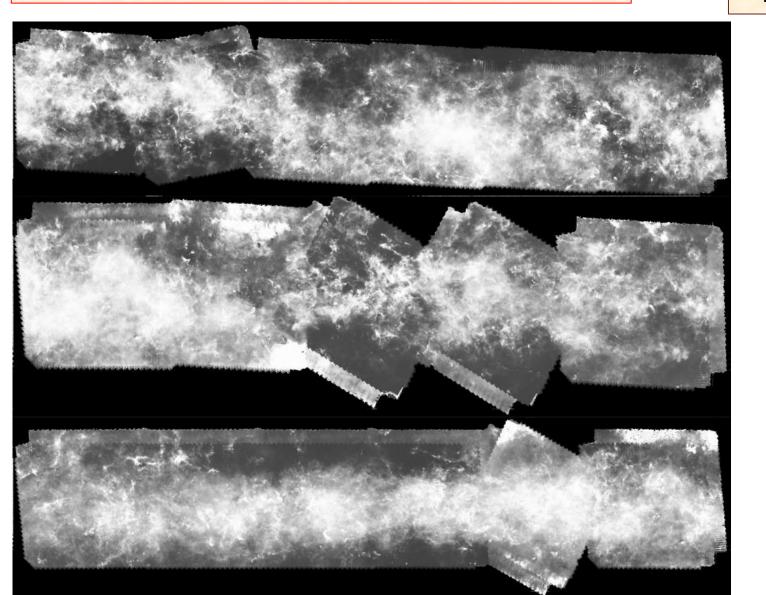


FilExSeC – Tests dataset (3)



Column density map (250µ) of Hi-GAL 289-320 deg region





Hi-GAL 289-300

Hi-GAL 300-310

Hi-GAL 310-320

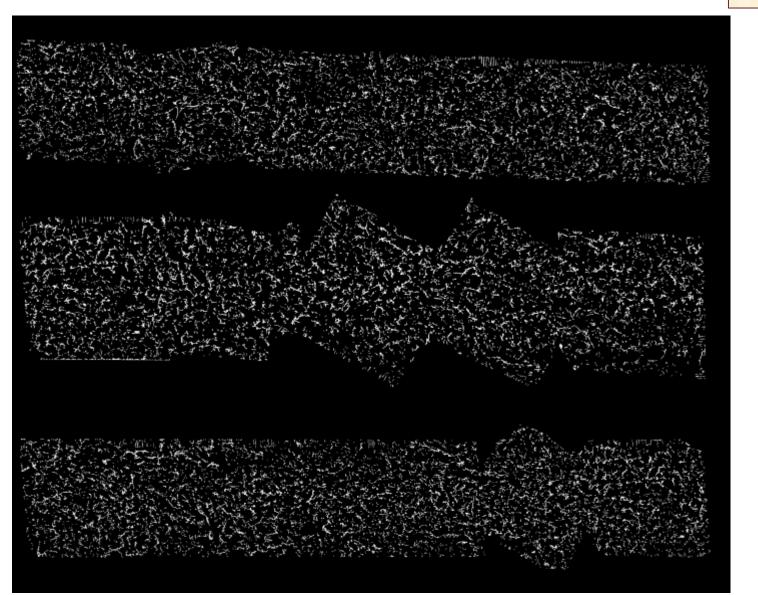


FilExSeC – Tests dataset (3)



Column density map (250µ) of Hi-GAL 289-320 deg region

<u>Tests</u> T12 (as Test sets)



<u>Hi-GAL 289-300</u> <u>Mask</u>

<u>Hi-GAL 300-310</u> <u>Mask</u>

<u>Hi-GAL 310-320</u> <u>Mask</u>

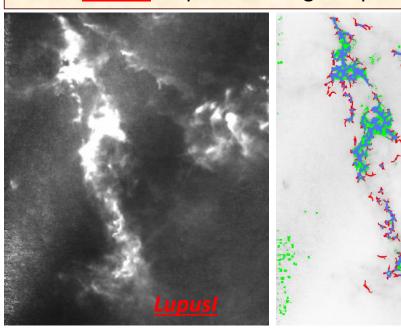


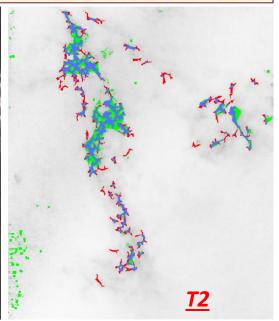
FilExSeC – Tests with Lupus (1)



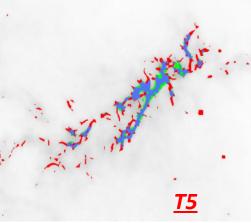
Test T2: Lupus III Training – Lupus I Test

Test T5: Lupus I Training – Lupus III Test









Same Features
Train/Test dataset inverted

T2/T5	TEST	FIL	BG
m wida .	T2	61%	99%
purity	T5	77%	98%
completences	T2	67%	98%
completeness	T5	50%	100%
DICE	T2	64%	99%
DICE	T5	61%	99%

Performances depend on the train image quality.

The results confirm the capability to extend the filament pixel regions

$$purity = \frac{CFP}{CFP + NFP}$$

$$completeness = \frac{CFP}{CFP + UFP}$$

$$DICE = \frac{2 * purity * complet.}{purity + complet.}$$

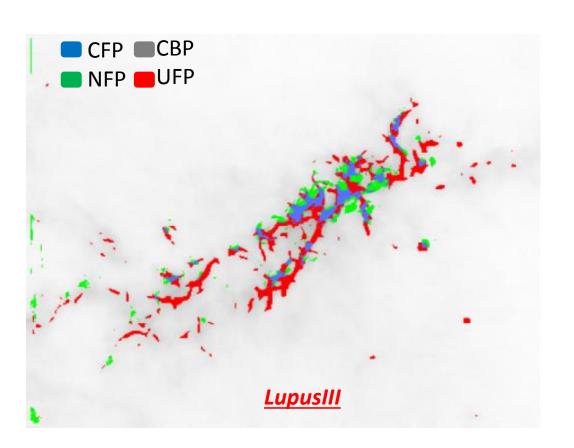


FilExSeC – Tests with Lupus (2)



Test T3a: Lupus I Training – Lupus III Test

Statistical Features Excluded



T3a	FIL	BG
Purity	48%	98%
Completeness	29%	99%
DICE	36%	98%

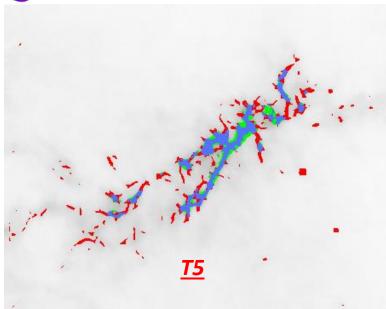
Without statistical features the performances strongly decrease.

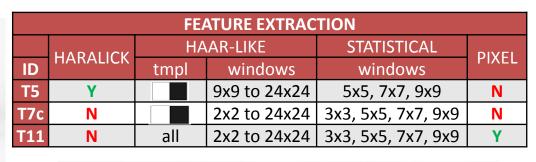
Haar-like + Haralick give a lower contribute

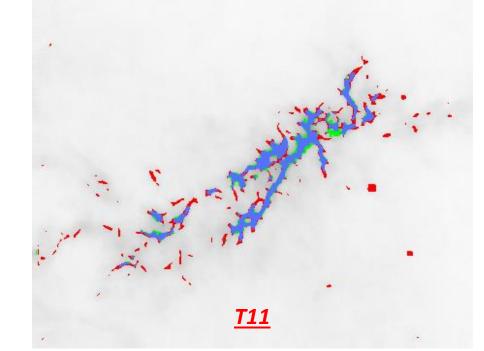


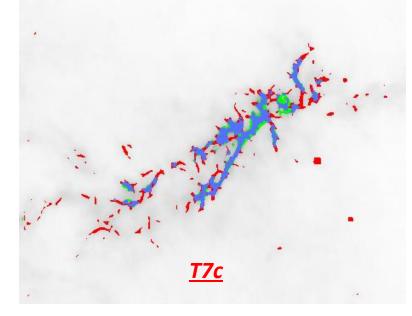
FilExSeC – Tests with Lupus (3)







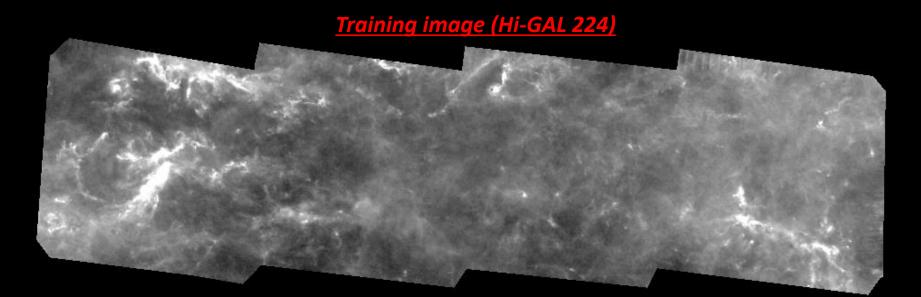




	T5		T7c		T11	
	FIL (%)	BG (%)	FIL (%)	BG (%)	FIL (%)	BG (%)
Purity	77.38	98.41	82.73	98.62	85,41	98,83
Complet.	50.42	99.52	56.94	99.61	63,88	99,65
DICE	61.06	98.96	67.45	99.11	73,09	99,24



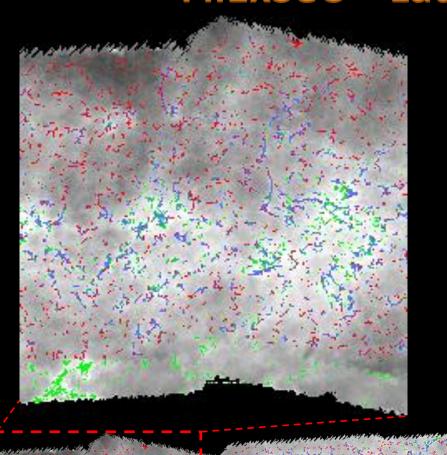




Test image: T12a (Hi-GAL 289-300)





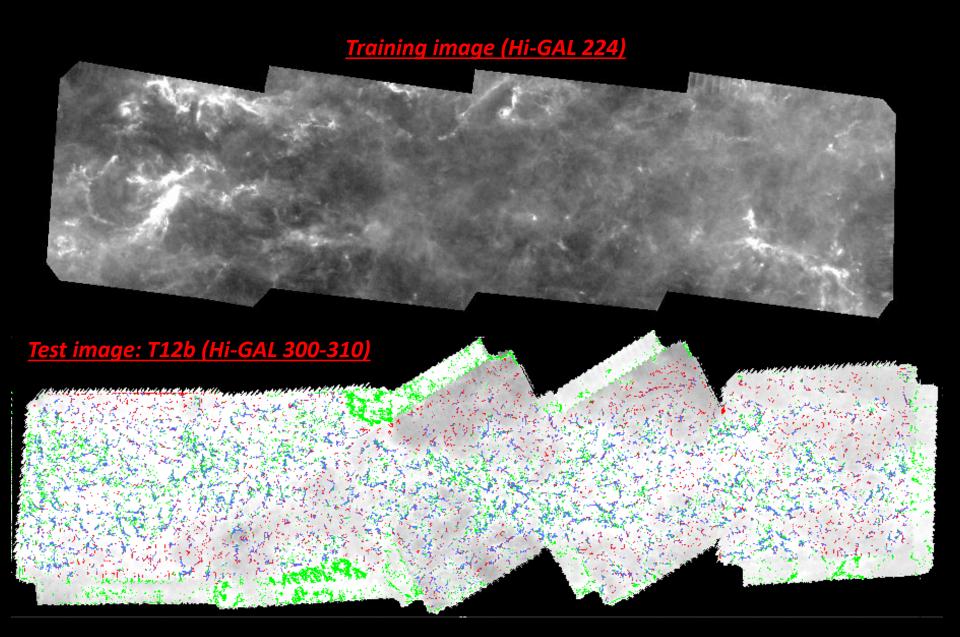


<u>Test image: T12a (Hi-GAL 289-300)</u>

T12 a	FIL	BG
Purity	71.08%	96.80%
Completeness	54.14%	98.44%
DICE	61.47%	97.61%





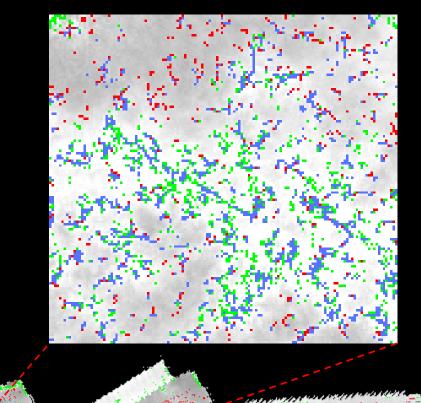


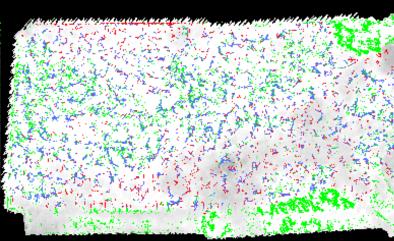


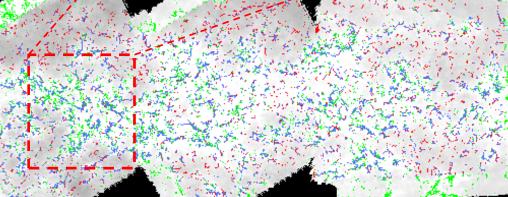




T12b	FIL	BG
Purity	52.06%	98.06%
Completeness	71.14%	95.70%
DICE	60.12%	96.86%

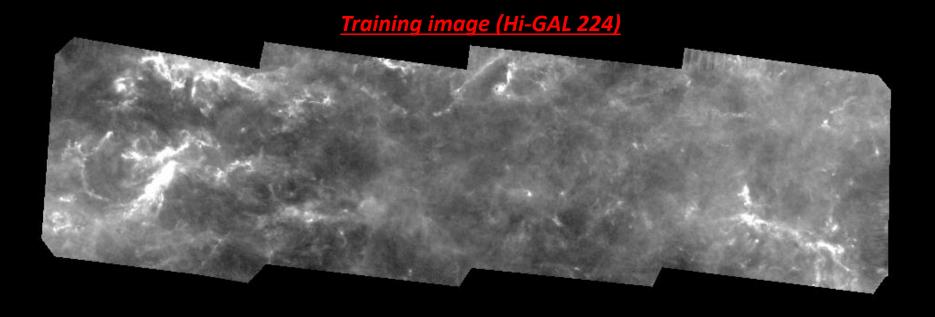


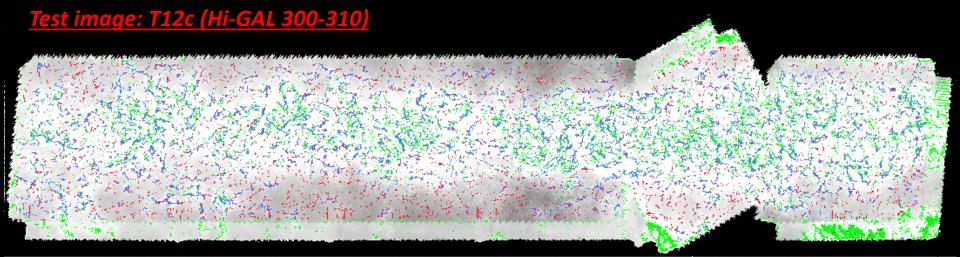






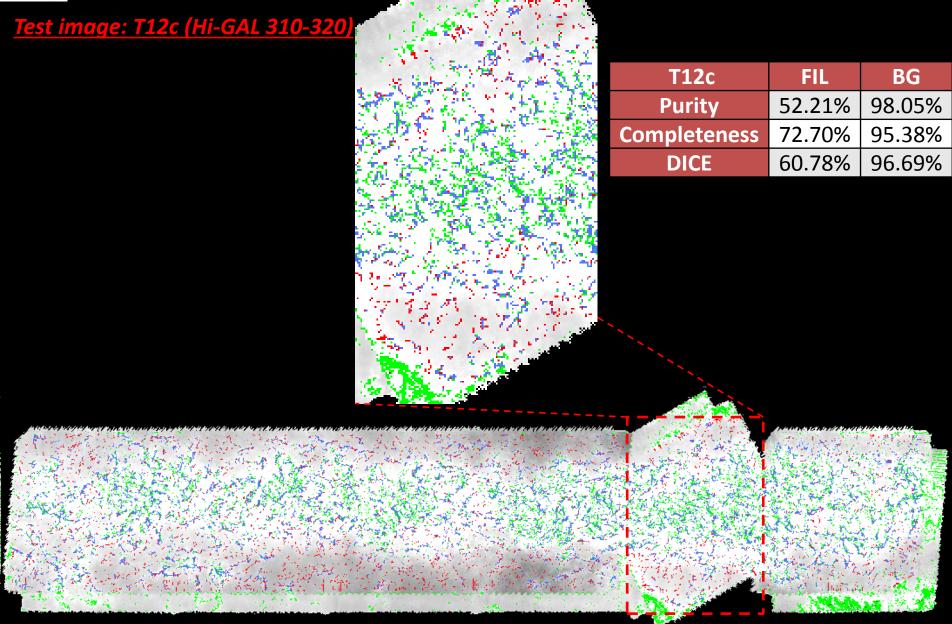














FilExSeC – Tests Results



Robustness of the method

- The features mostly contributing to classification are always the same, regardless of the datasets;
- Comparing tests on the same region by slightly varying training set, the results show extremely low % of background detection difference (always < 0.5%) and of filament (max 6.75%)

High Performances

- Global efficiency grows up by choosing a more realistic training set;
- High capability to identify additional filament pixels with respect to the traditional method.
 However, a certain number of very little filamentary structures was not recognized, by confusing them as background

Best Configuration

- Haralick features are useless
- All Haar-like templates needed with rectangles size up to 24x24
- Windows for Statistical features 3x3 5x5 7x7 9x9
- Pixel value is very important
- A low number of trees is sufficient for the best classification (1000 trees)

Next step is to verify, together with IAPS astronomers, the results obtained by FilExSeC from a physical point of view, by calculating, for example, the variation of filament's mass function by adding the new filament pixels found



FilExSeC – Alternative edge detectors



The main limitation of FilExSec is that it works with masks obtained by traditional methods. This causes a bias on our performances.



It is necessary to find a method directly working on original images without any priors



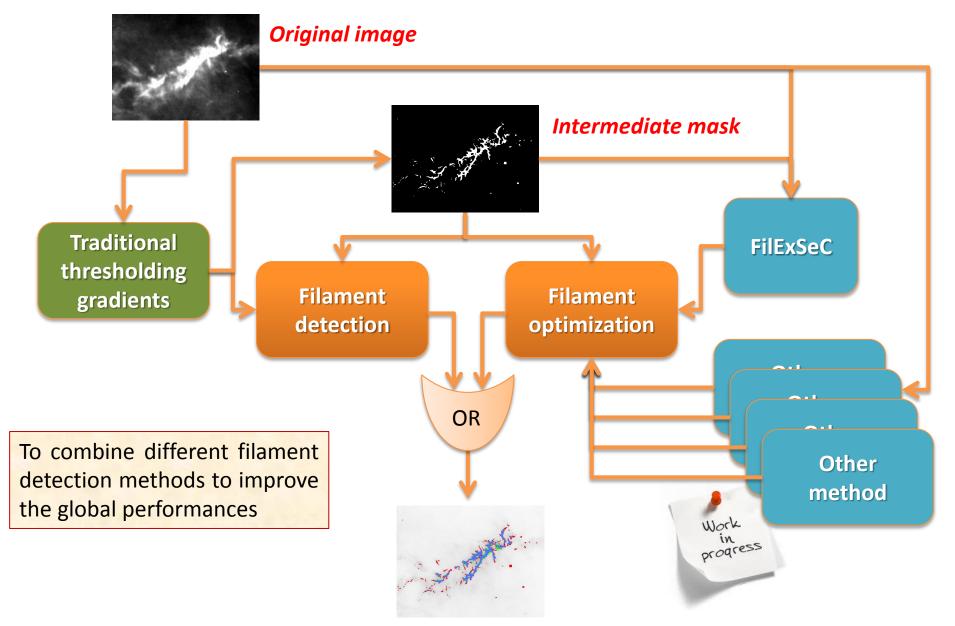
At this time, we are under investigation on new edge detectors:

- Boosted Edge Learning (Dollar et al. 2006)
- gPb (global Probability of boundary) (Arbelaez et al. 2011)
- Beam-curve Pyramid based edge detector (Alpert et al. 2010)
- Curvelets and Wavelets (Starck et al. 2002 and Mallat 1998)
- Fuzzy Logic Edge Detectors (Becerikli et al. 2005)
- Canny and Sobel filters enhancement (Canny 1986 and Sobel 2014)



FilExSeC (connections)



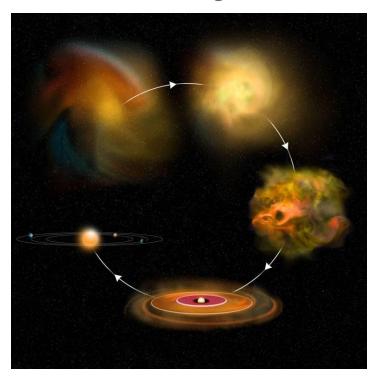




OACN – WP5 Activities



WP2 – Task 4 Star Forming Sources Evolutionary Classification



Design



OACN – WP5 Activities



WP2 – Task 4
Star Forming Sources
Evolutionary Classification



Design



WP2 – task 4 Evolutionary classification



An evolutionary classification tool for ViaLactea, will catalogue "clumps" in terms of the evolutionary stage and mass regime of the ongoing star formation. There are two components that need to be developed at the foundation of the classification tool:

- 1. an evolutionary classification toolbox
- 2. a set of star-forming clumps in known stages of evolution to be used as a training/test-set for machine-learning algorithms... ...and adopt some kind of evolutionary scheme

Data-mining approaches to source classification

Weak Gated Classification

We know nothing about the sources evolutionary stage;

Identify over-densities in the given parameter space (e.g., built on the evolutionary toolbox, plus any other available evidence);

Data are then grouped into clusters: groups of data entries sharing common but *a priori* unknown correlations among parameter space features.

Supervised Classification

For a subsample of points, its category/class is well known;

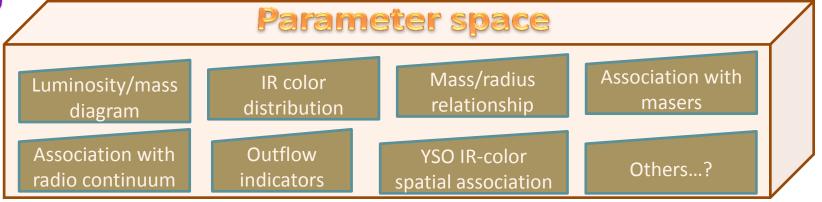
Need order of 10³ objects to be used as a training set;

Balanced population of classes in the training set.



SF-SEC (1st approach)

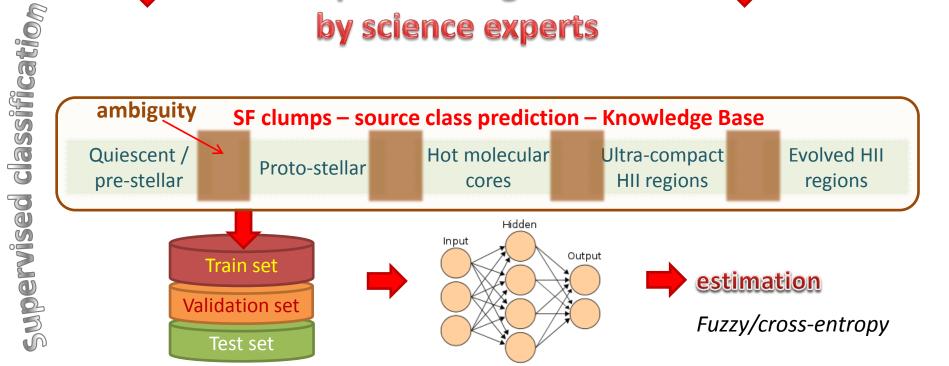






class partitioning inferred by science experts





VIALACTEA

classification

gated

SF-SEC (2nd approach)



