

Cross-identification au CDS

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Sommaire

1 Simple x-match

2 Multi-catalogue χ -match & x-id

X-Match within the VO

Using VO tools

- No dedicated VO protocol
- Topcat
 - ▶ Use the VO Registry and download a VOTable
 - ▶ Perform one Cone Search by row
 - ★ good for VizieR's stats...
 - ★ ...but saturates the server when multi-threaded!
- Topcat / Aladin
 - ▶ Use the VO Registry and download VOTables
 - ▶ Perform the x-match locally
 - ★ Tycho2 vs 2MASS
 - ★ ⇒ Download a lot of data
 - ★ ~~ `java.lang.OutOfMemoryError`
 - ★ ⇒ User must slice the sky, ...
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X-Match within the VO

TAP for small/medium x-matches

- Full VO x-match: TAP is the only way
- TAP (Table Access Protocol)
 - ▶ ADQL: Astronomical Data Query Language
 - ▶ UWS: Universal Worker Service
- Example

```
SELECT TOP 1000 *
FROM twomass AS a
JOIN tycho2 AS b
ON 1=CONTAINS(POINT('ICRS', a.ra, a.dec),
               CIRCLE('ICRS', b.ra, b.dec, 5./3600.))
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TAP for small/medium x-matches

- CDS TAP Service **benchmark** (Gilles Landais)
 - ▶ **No output (simple count(*))**
 - ▶ Before optimization (Q3C / H3C)
 - ★ Hipparcos Main (118 k) vs 2MASS (470 M): 14 min
 - ★ Tycho2 (2 M) vs 2MASS (470 M): 48 min
 - ★ 2 large catalogues: > 1 day
 - ▶ After optimization (H3C)
 - ★ Hipparcos Main (118 k) vs 2MASS (470 M): 6.5 min
 - ★ Tycho2 (2 M) vs 2MASS (470 M): 11.5 min
 - ★ GSC ACT (25 M) vs 2MASS (470 M): 1 h 30 min
- Pro: SQL flexibility
- Limit: all known TAP implementations on top of a SGBD
 - ▶ PostgreSQL (CDS, GAVO, ESAC): no multi-threading
 - ▶ Possible performances issues
 - ▶ SGBD fitted for large catalogues x-match?

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The CDS X-Match service

Web interface

- Released in 2011
- General purpose service (any table having positions) \Rightarrow no probabilities
- **UWS** (Grégory Mantelet's library)
- **Very efficient** basic xmatch designed for the biggest available catalogues
 - ▶ 2 catalogues at the same time
 - ▶ **Simple** (few options)
 - ▶ E.g. SDSS DR9 vs 2MASS at 2" done in 15 min
 - ★ xmatching: 5 min (50 M links)
 - ★ building result file: 10 min (14 GB)
 - ★ running on a **single** server
- >20 G links computed in 2015, e.g.
 - ▶ NOMAD (1.1 G) vs USNOA2 (0.5 G)
 - ▶ 750 M links, 117 GB, 1h28

CDS X-Match Service X-match Tables management Documentation

Choose tables to cross-match

SIMBAD  2MASS 

VizieR SIMBAD My store VizieR SIMBAD My store

SIMBAD astronomical database
7,144,748 objects with position 

2MASS All-Sky Catalog of Point Sources (Cutri+ 2003)
470,992,970 rows 

Hide options

Cross-match criteria

By position
Radius: 5 arcsec

By position including error
Sigma: 3.43935 (completeness: 99.73 %)
Max. distance: 5 arcsec

Cross-match area

All sky

Cone
Center: Position/Object name
Radius: 5 deg

Healpix cell (ICRS, NESTED scheme)
Nside: 4
Index: 0

Begin the X-Match

Visualize and manage your cross-match jobs

List of X-match jobs

Table 1	Table 2	Options	Begin	Stop
No job in list				

Navigation icons: back, forward, search, etc.

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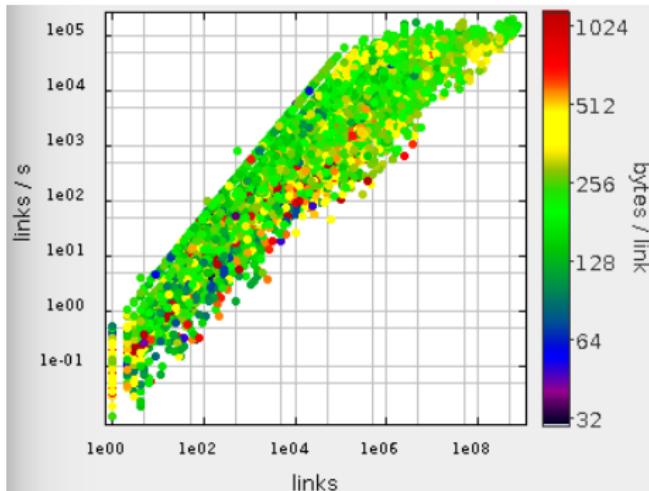
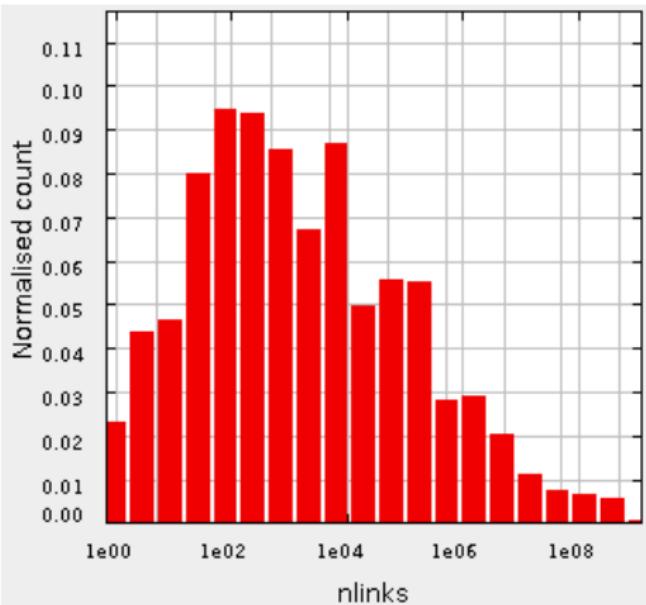
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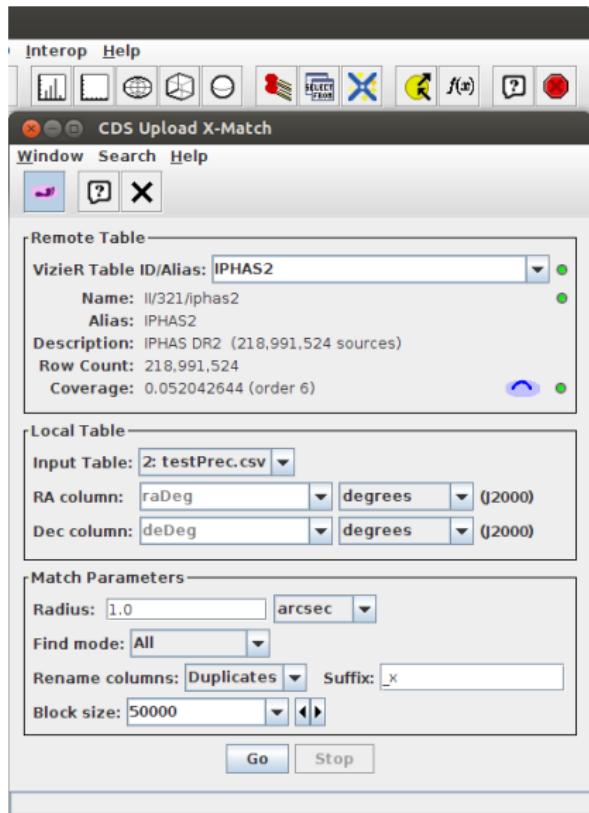
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The CDS X-Match service

HTTP API

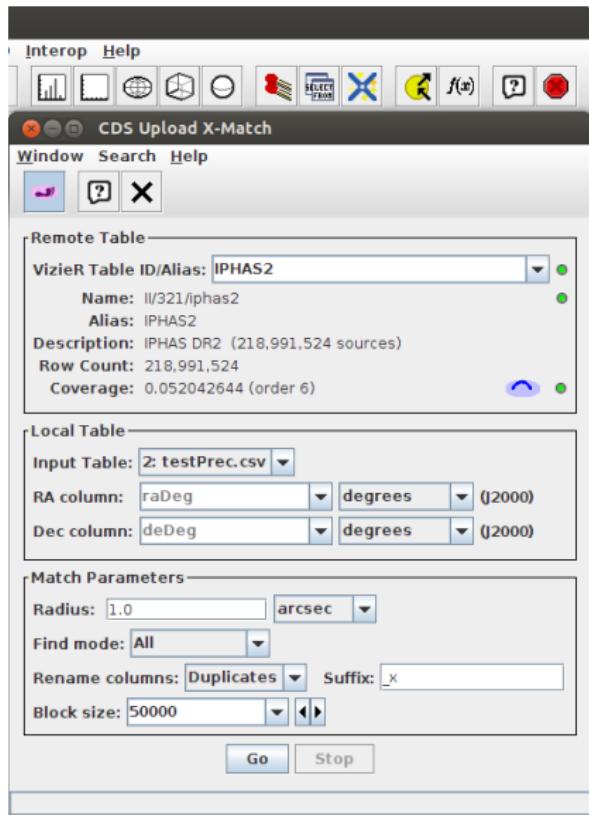
- Released in 2013
- Uses **DALI**
- Synchronous programmatic access
- Available from
 - ▶ Web Browser, wget, curl, ...
 - ▶ TOPCAT, Portal MAST, Astropy
- Output limited to 2 M rows
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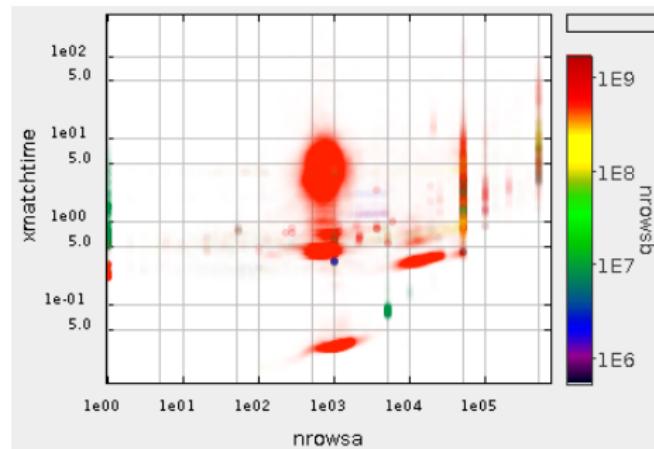
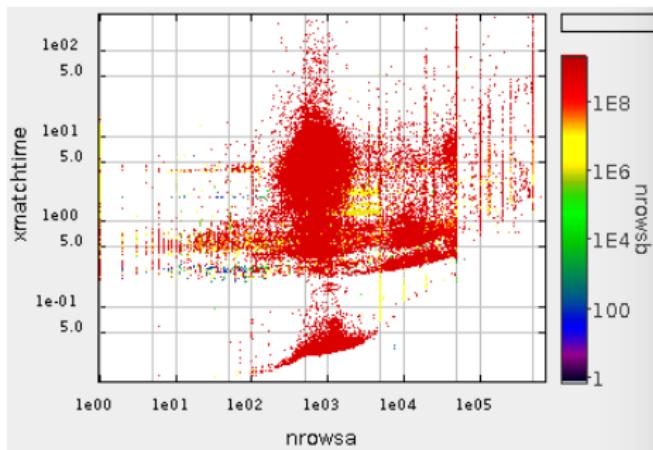
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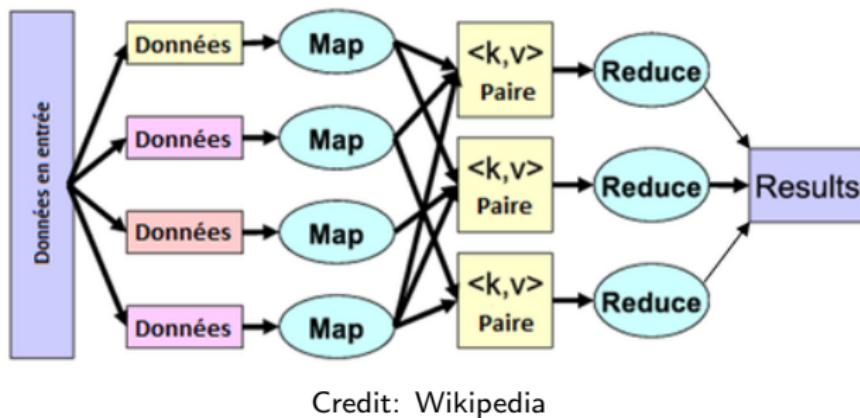
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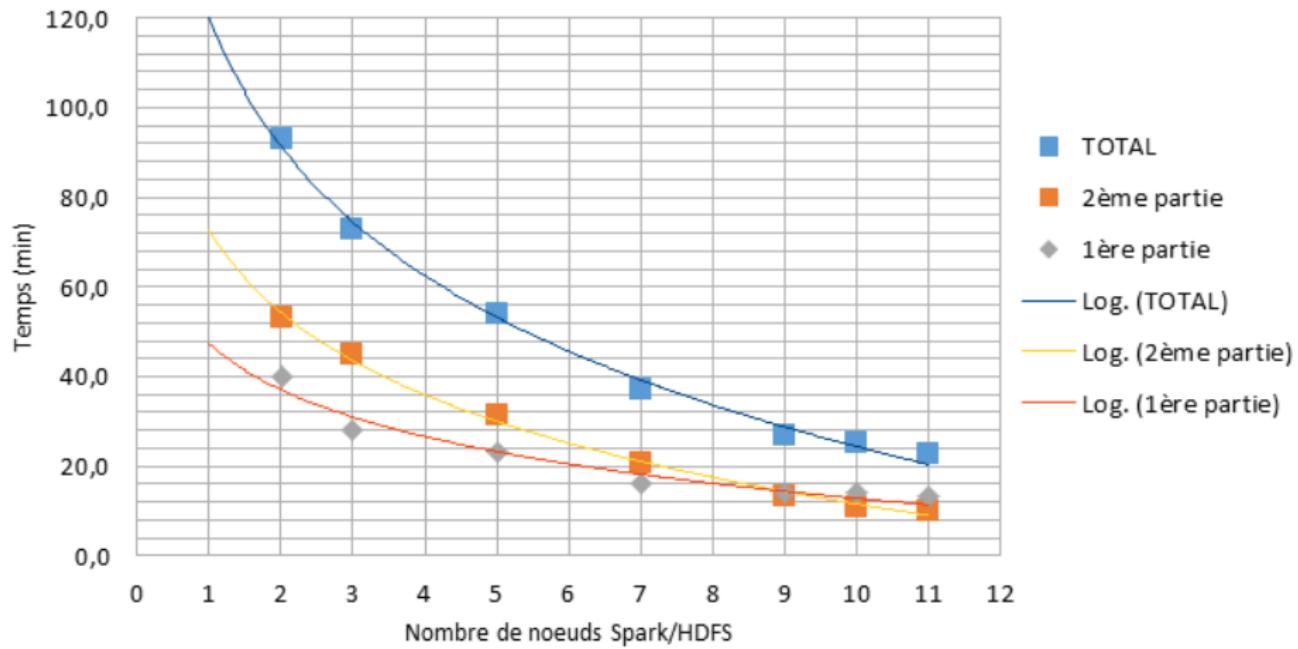
Test of Big Data techno

- N. Wali, A. Schaaff, F.-X. Pineau
- HADOOP / SPARK: MapReduce
- Test on SDSS DR9 / 2MASS
- Bottleneck: shuffle
 - ▶ Co-partitioning OK
 - ▶ But no way to co-locate data!!
(no "block affinity groups")



Test of Big Data techno

Temps de XMatch en fonction du nombre de noeuds



Sommaire

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From x-match to χ -match and x-id

- Existing tools: simple fixed radius x-match
- ARCHES (FP7): we would like identifications
 - ▶ very tricky / complex!
 - ▶ 2 (possibly) independant parts
 - ★ astrometric part (take into account positional errors)
 - ★ photometric part (not adressed here!)
 - ★ $proba_{id} \propto prior \times likelihood_{astrom} \times likelihood_{phot}$
 - ★ $proba_{id} \propto proba_{astrom} \times likelihood_{phot}$
 - ▶ SEDs \Rightarrow multi-catalogue x-match

Candidate selection

Method

- Steps to probabilistic positional x-match
 - ▶ Make simplifying assumptions
 - ▶ Select candidates: select and group together sources possibly being various detections of a same real source
 - ★ Need for a selection criterion
 - ▶ Make hypothesis: are the sources really from a same real sources or from different real sources?
 - ▶ For each hypothesis:
 - ★ derive the associated *likelihood*
 - ★ derive the associated *prior*
 - ▶ Compute astrometry based probabilities

Candidate selection

Simplifying assumptions

- Radical simplifying assumptions:
 - ▶ No proper motions
 - ▶ No blending
 - ▶ No clustering (density of sources = Poisson law)
 - ▶ No systematic offsets
 - ▶ You can trust positional uncertainties provided in catalogues

Candidate selection

Selection criterion

How to select a group of n sources from n distinct catalogues as possibly being various observations of a same actual source?

- *Statistical hypothesis testing*
 - ▶ H_0 (null hypothesis): all n sources are from the same real source
 - ▶ $H_1 = \bar{H}_0$ (alternative hypothesis): at least one source (out of n) is spurious
- User input: γ , the probability to accept H_0 while it is true
 - ▶ γ (I call it completeness) is called *true negative rate*
 - ▶ we usually fix $\gamma = 0.9973$ (99.73%, value of the 3σ rule in 1 dimensional pb)
 - ▶ \Leftrightarrow fixing the **type I error** = 0.027% = proba to reject null hypothesis while it is true
 - ▶ we (theoretically) miss 27/10 000 real association
- The criterion used is simply a χ^2 test of $2(n - 1)$ degrees of freedom

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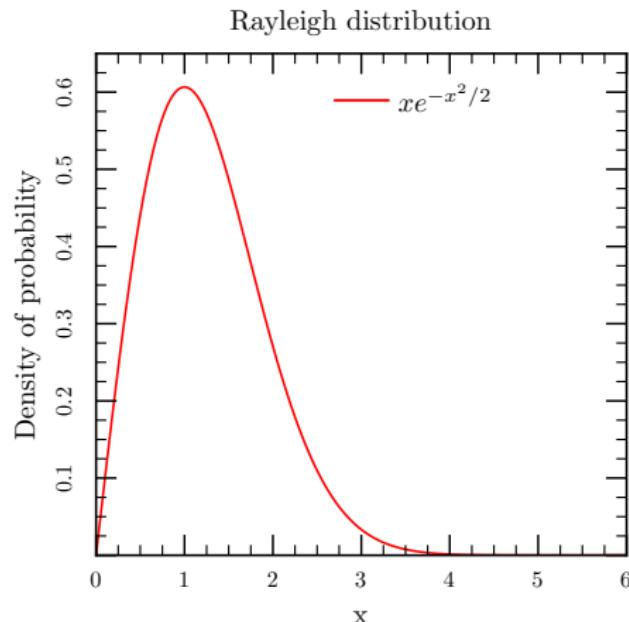
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Classical 2 catalogues case

- For real associations, i.e. when H_0 is true
 - ▶ The distribution of normalized distances is a Rayleigh distribution of scale $\sigma = 1$

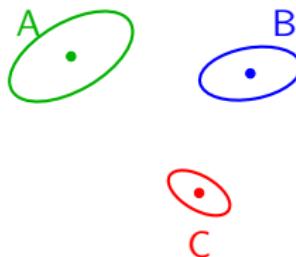
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Candidate selection

Iterative form

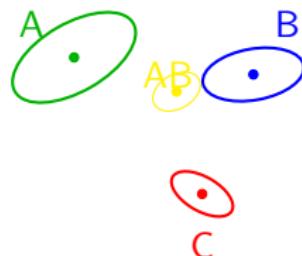
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- \Rightarrow we can perform $(n-1)$ successives and iteratives xmatches
- At each step, we use for next position and next error ellipse
 - ▶ the weighted mean position of the previous xmatch (MLE)
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- Result independant of the xmatch order (for INNER JOIN!)



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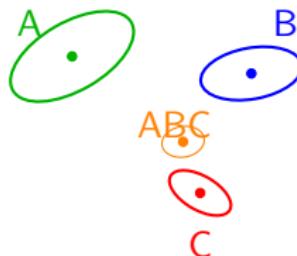
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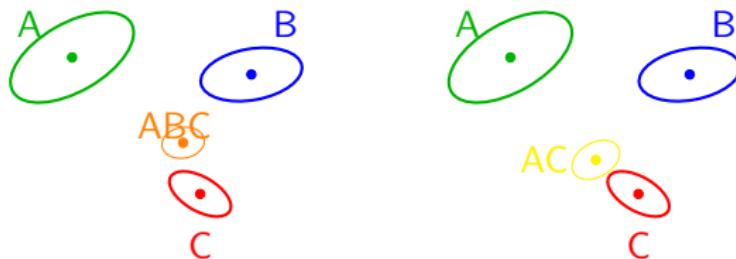
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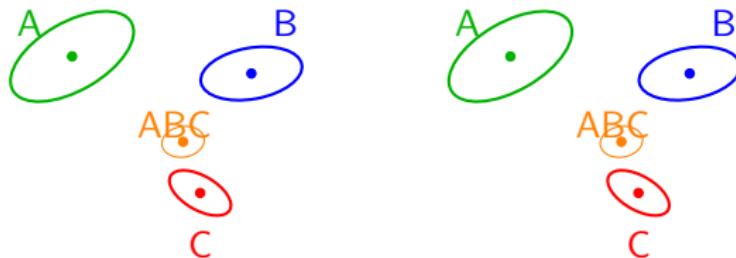
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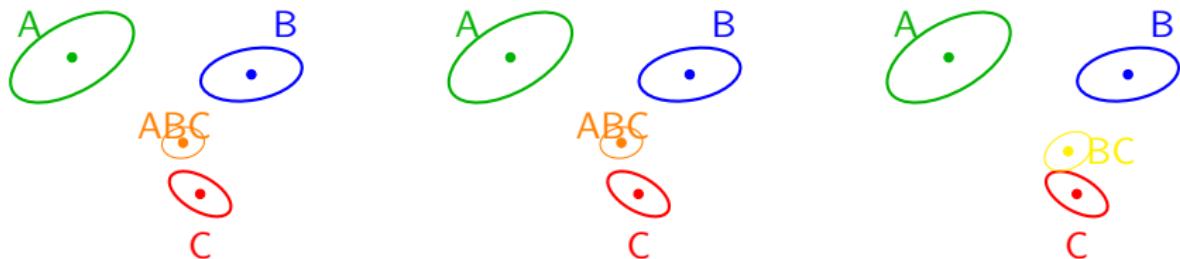
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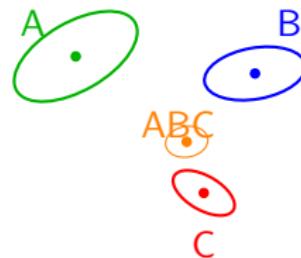
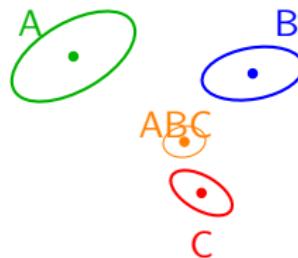
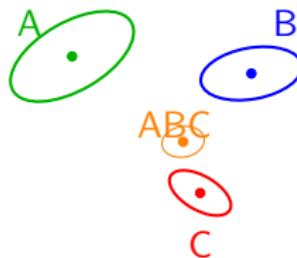
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Making hypotheses

For 2 and 3 catalogues

To compute Bayes probabilities, we MUST consider all possible hypothesis.

- Law of total probabilities:

$$\sum_{i=1}^n p(H_i) = 1$$

- For 2 catalogues
 - ▶ 2 hypothesis

- ★ AB (H_0)
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• For 3 catalogues

▶ 5 hypothesis

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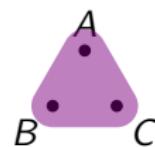
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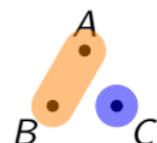
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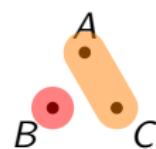
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●
B

- For 3 catalogues

- ▶ 5 hypothesis
 - ★ ABC (H_0)
 - ★ AB_C
 - ★ AC_B
 - ★ A_BC
 - ★ A_B_C

A
●
B
●
C

Making hypotheses

For 2 and 3 catalogues

To compute Bayes probabilities, we MUST consider all possible hypothesis.

- Law of total probabilities:

$$\sum_{i=1}^n p(H_i) = 1$$

- For 2 catalogues

- ▶ 2 hypothesis
 - ★ AB (H_0)
 - ★ A_B

A
●
B

- For 3 catalogues

- ▶ 5 hypothesis
 - ★ ABC (H_0)
 - ★ AB_C
 - ★ AC_B
 - ★ A_BC
 - ★ A_B_C

A
●
B
●
C

Making hypotheses

For n -catalogues

- We generalised for n catalogues
- The number of hypothesis to be tested is given by the BELL number

Table : Values of the seven first BELL numbers

n	2	3	4	5	6	7
B_n	2	5	15	52	203	877

- ▶ n number of catalogues
- ▶ $n=5$ catalogues \rightsquigarrow 52 probabilities to be computed
- \Rightarrow Combinatorial explosion!

Making hypotheses

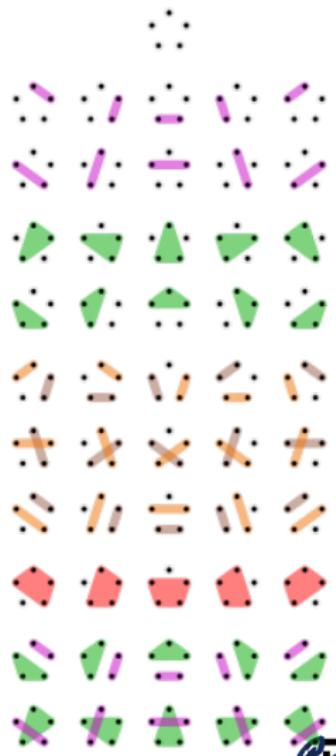
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Making hypotheses

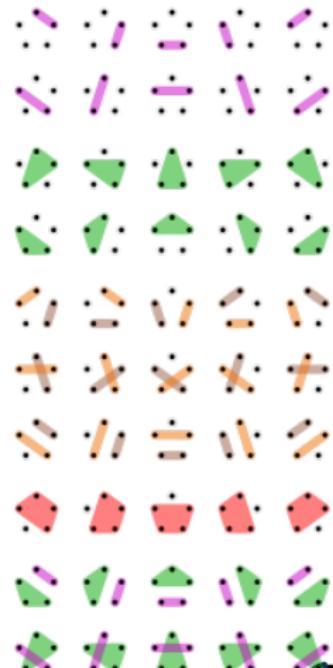
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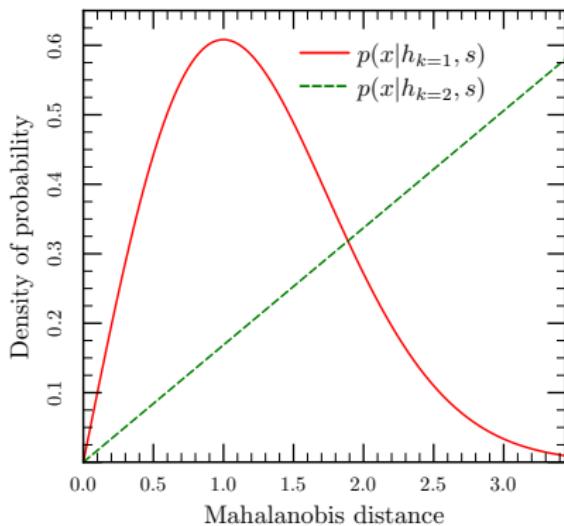


Likelihoods

For 2 and 3 catalogues

- Likelihood depends only on the nyumber of “actual” source

Likelihoods for $n = 2$ and $\gamma = 0.9973$

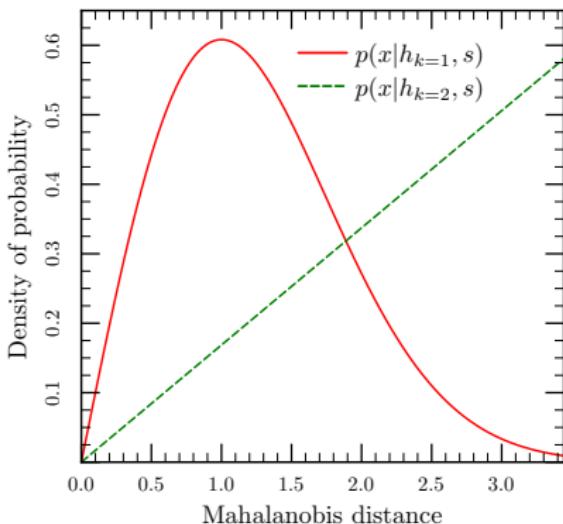


Likelihoods

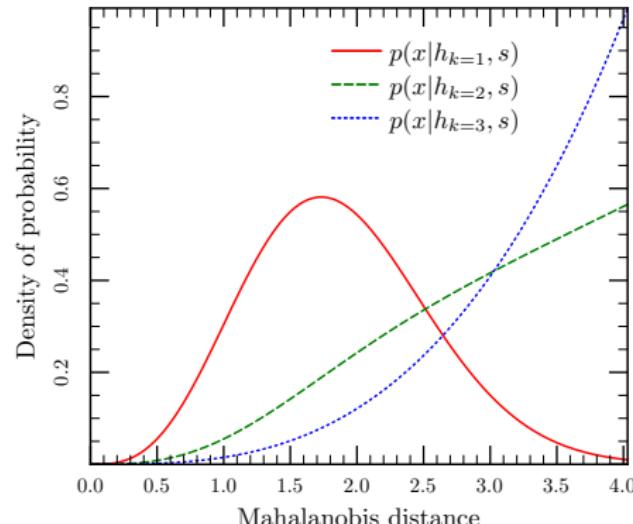
For 2 and 3 catalogues

- Likelihood depends only on the nyumber of “actual” source

Likelihoods for $n = 2$ and $\gamma = 0.9973$



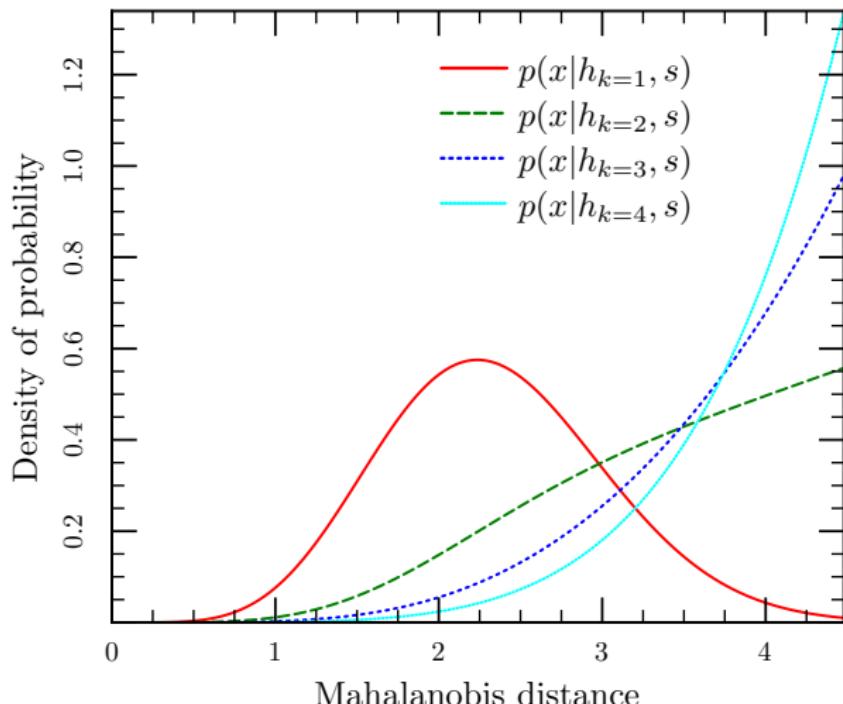
Likelihoods for $n = 3$ and $\gamma = 0.9973$



Likelihoods

For 4 catalogues

Likelihoods for $n = 4$ and $\gamma = 0.9973$

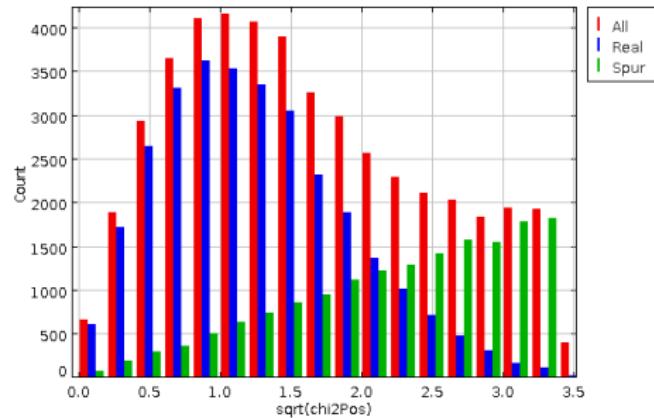


Bayesian probabilities

For 2 catalogues

- Blue curve: $n_{H_0} \times p(x|H_0)$
- Green curve: $n_{H_1} \times p(x|H_1)$
- Red curve = blue + green
- For an association of given x :

$$p(H_0|x) = \frac{\text{Blue curve}(x)}{\text{Red curve}(x)}$$



- Bayes formula:

$$p(H_0|x) = \frac{p(H_0)p(x|H_0)}{p(H_0)p(x|H_0) + p(H_1)p(x|H_1)}$$

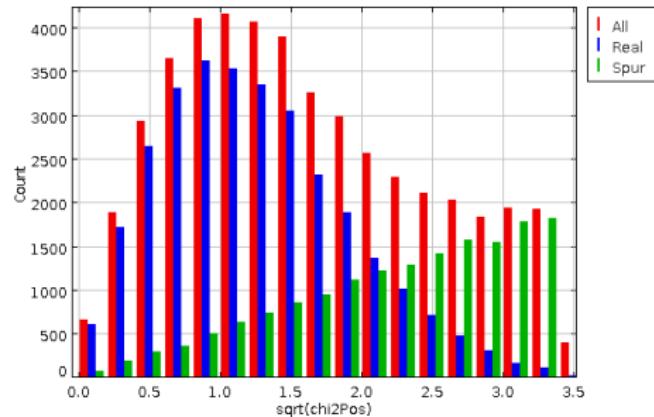
- Here priors $p(H_0) = n_{H_0}/n_{\text{tot}}$ and $p(H_1) = n_{H_1}/n_{\text{tot}}$

Bayesian probabilities

For 2 catalogues

- Blue curve: $n_{H_0} \times p(x|H_0)$
- Green curve: $n_{H_1} \times p(x|H_1)$
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- Here priors $p(H_0) = n_{H_0}/n_{\text{tot}}$ and $p(H_1) = n_{H_1}/n_{\text{tot}}$

Priors from geometrical considerations

- Common surface area of n catalogues: Ω
- Region of acceptance of the χ^2 test:
 - ▶ 2 catalogues case: ellipse
 - ▶ n catalogues case: $2(n - 1)$ -dimensional ellipsoïd
- $\hat{n}_{spur} \propto$ mean volume of $2(n - 1)$ -dimensional ellipsoïd / Ω^{n-1}
- ⇒ a way to define priors

Test on synthetic catalogues

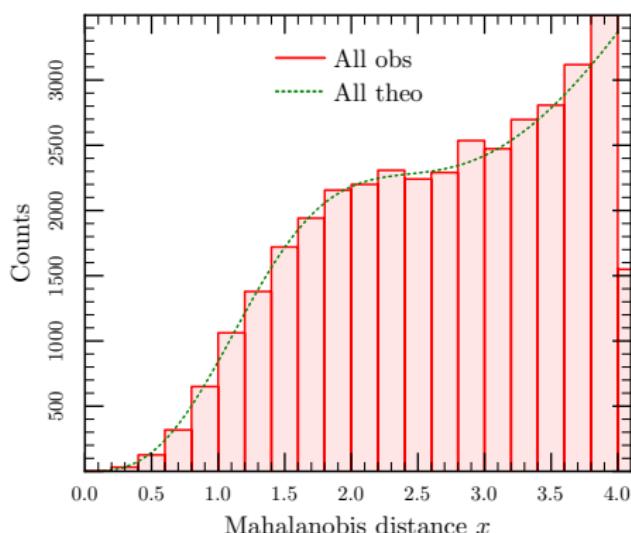
$n_A = 40\,000$ $n_B = 20\,000$ $n_C = 35\,000$

$n_{AB} = 6\,000$ $n_{AC} = 12\,000$ $n_{BC} = 18\,000$

$n_{ABC} = 10\,000$ Cone radius= 0.42°

Err A: cte 0.4"; Err B: $\mathcal{N}(0.75'', 0.1'')$; Err C: linear from 0.8 to 1.2"

Theoretical and Observed all associations



Mahalanobis distance x

Green curve computed from geometrical considerations only!

Test on synthetic catalogues

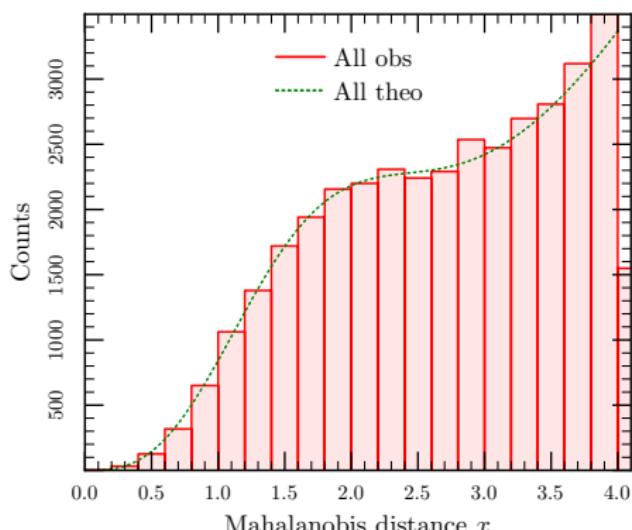
$$n_A = 40\,000 \quad n_B = 20\,000 \quad n_C = 35\,000$$

$$n_{AB} = 6\,000 \quad n_{AC} = 12\,000 \quad n_{BC} = 18\,000$$

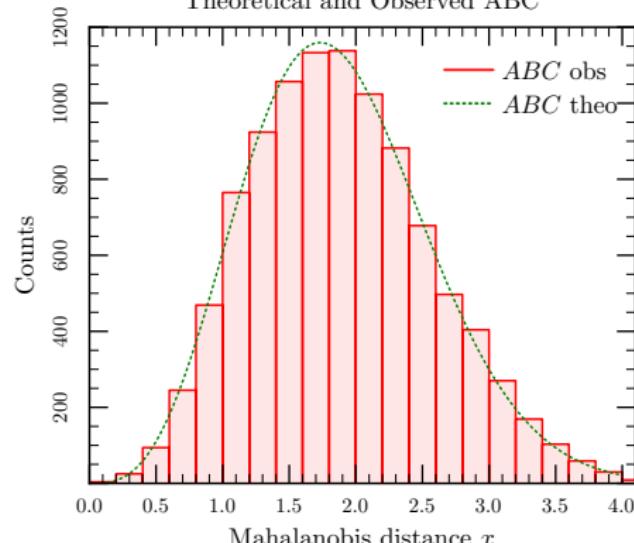
$$n_{ABC} = 10\,000 \quad \text{Cone radius} = 0.42^\circ$$

Err A: cte 0.4"; Err B: $\mathcal{N}(0.75", 0.1")$; Err C: linear from 0.8 to 1.2"

Theoretical and Observed all associations

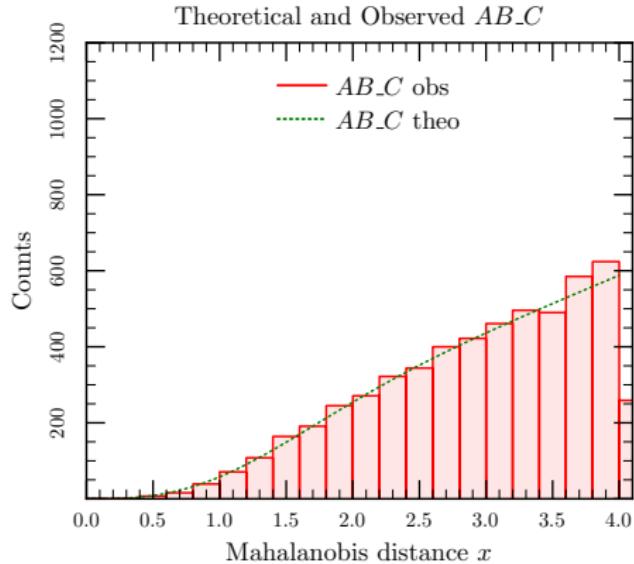
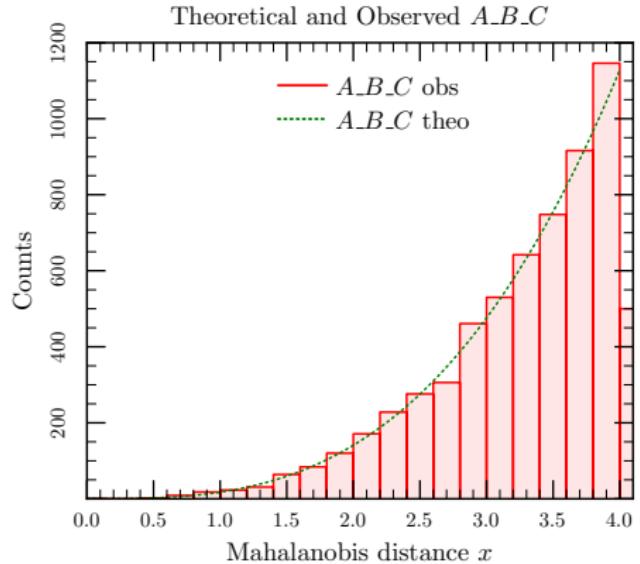


Theoretical and Observed ABC

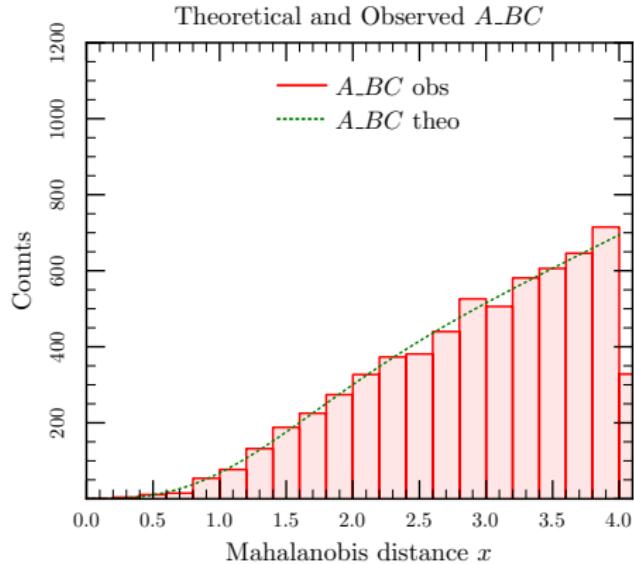
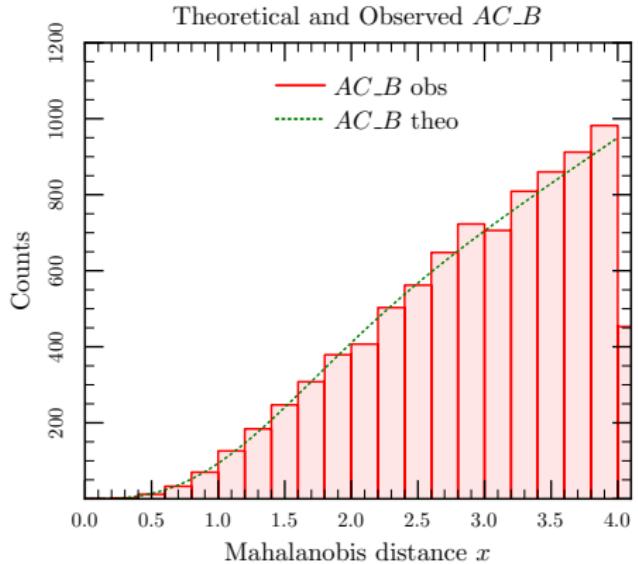


Green curve computed from geometrical considerations only!

Test on synthetic catalogues



Test on synthetic catalogues



Features: various xmatch algorithms

Algorithm	param	#tbl	prop.mot.	index struct.
chi2 (χ^2)	proba	2	l^1, r^2, b^3	M/TM-tree

XMatches are chainable: 1 χ^2 xmatch of 4 tables = 3 χ^2 xmatches of 2 tables!

4 to 1 joins (left-right) & (right-left) are supported according to the algorithm.

- ¹ l: left table contains extended objects or proper motions;
- ² r: right table contains extended objects or proper motions;
- ³ b: both left and right tables contain extended objects or proper motion.
- ⁴ mec: Minimum Enclosing Cone

Features: various xmatch algorithms

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chi2 (χ^2)	proba	2	I ¹ , r ² , b ³	M/TM-tree
proba2_vx	proba	2	no (?)	M-tree
proba3_vx	proba	3	no (?)	M-tree
proba4_vx	proba	4	no (?)	M-tree
probaN_vx	proba	n	no (?)	M-tree

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proba4_vx	proba	4	no (?)	M-tree
probaN_vx	proba	n	no (?)	M-tree
knn	k+dist	2	r, b	kd/M/TM-tree
cone	dist	2	I, r, b	kd/M/TM-tree

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4 to 11 joins ($LIRFLIR'L'I'R'F'$) are supported according to the algorithm.

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cone	dist	2	l, r, b	kd/M/TM-tree
mec ¹	dist	n	no (?)	kd/M-tree

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cone	dist	2	l, r, b	kd/M/TM-tree
mec ¹	dist	n	no (?)	kd/M-tree
ext_l ¹	r	2	no	M-tree
ext_r ²	r	2	no	M-tree
ext_b ³	r	2	no	M-tree

XMatches are chainable: 1 χ^2 xmatch of 4 tables = 3 χ^2 xmatches of 2 tables!

4 to 11 joins ($LIRFL\bar{I}\bar{R}L'I'R'F'$) are supported according to the algorithm.

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probaN_vx	proba	n	no (?)	M-tree
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cone	dist	2	l, r, b	kd/M/TM-tree
mec ¹	dist	n	no (?)	kd/M-tree
ext_l ¹	r	2	no	M-tree
ext_r ²	r	2	no	M-tree
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...

XMatches are chainable: 1 χ^2 xmatch of 4 tables = 3 χ^2 xmatches of 2 tables!
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- 3 b: both left and right tables contain extended objects or proper motion.
- 1 mec: Minimum Enclosing Cone

ARCHESS X-match tool Web interface

ARCHESS X-MATCH TOOL Anonymous Web form

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Observatoire astronomique de Strasbourg



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astronomiques de Strasbourg

Remote directory

Upload a file:

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Aucun fichier

File list

3xmme_uniquesources
2mass.174.10491_7.223
sdss9.174.10491_7.223
galex5als.174.10491_7.223

X-match script

Script examples

[Chi-square xmatch of sdss/2mass in a cone, simple version](#)

Type, modify or copy/paste here the xmatch script to be executed:

```
1 #####  
2 # Name: chi2xmatch.xms  
3 # Description: Load SDSS and 2MASS data from VizieR and perform a chi-square  
4 # xmatch (full join) of the two loaded tables.  
5 # Remarks:  
6 # - look at how we add a 0.1 arcsec systematic on SDSS positional errors  
7 # - look also at how we add a new computed column to 2MASS data.  
8 # Input files: none  
9 # Output files:  
10 # - sdss_2mass.vot: result of the xmatch  
11 #####  
12  
13 # Load SDSS DR9  
14 get VizieRLoader tabname=V/139/sdss9 mode=cone center="174.10491 +7.22343" radius=12.3arcmin allcolumns  
15 set pos ra=RAJ2000 dec=DEJ2000  
16 set poserr type=RCD_DEC_ELLIPSE param1=e_RAJ2000 param2=e_DEJ2000  
17 set cols *  
18  
19 # Load 2MASS data  
20 get VizieRLoader tabname=II/246/out mode=cone center="174.10491 +7.22343" radius=12.3arcmin allcolumns  
21 set pos ra=RAJ2000 dec=DEJ2000  
22 set poserr type=ELLIPSE param1=errMaj param2=errMin param3=errPA  
23 set cols *  
24  
25 # Perform the xmatch  
26 xmatch.chi2.nStep=1 nMax=1 completeness=0.0073 join=full  
27 ...
```

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Result log

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<http://serendib.unistra.fr/ARCHESSWebService/index.html>

Final words

- We plan to integrate the ARCHES engine in the CDS XMatch Service
- Multi-catalogue χ -matches based on INNER JOINS can be performed iteratively
- No more true for e.g. LEFT JOINS (result depends on x-match order)
- Multi-catalogue x-match feasible in ADQL?
- A (evolving) standard describing how to compute probabilities (possibly several ways)??