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Technical Report: CSV M Ecosystem

Using CSV M format in various scientific fields.

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Abstract

The CSV M format is derived from CSV format and allows the storage of tabular like data with a limited but extensible amount of metadata. This approach could help computer scientists because all information needed to use subsequently the data is included in the CSV M file and is particularly well suited for handling RAW data in a lot of scientific fields and to be used as a canonical format. The use of CSV M has shown that it greatly facilitates: the data management independently of using databases; the data exchange; the integration of RAW data in dataflows or calculation pipes; the search for best practices in RAW data management. The efficiency of this format is closely related to its plasticity: a generic frame is given for all kind of data and the CSV M parsers don't make any interpretation of data types. This task is done by the application layer, so it is possible to use same format and same parser codes for a lot of purposes. In this document some implementation of CSV M format for ten years and in different laboratories are presented. Some programming examples are also shown: a Python toolkit for using the format, manipulating and querying is available. A first specification of this format (CSV M-1) is now defined, as well as some derivatives such as CSV M dictionaries used for data interchange. CSV M is an Open Format and could be used as a support for Open Data and long term conservation of RAW or unpublished data.

Keywords

RAW Data, Tabular Data, CSV, Metadata, Data exchange, Dataflow, Canonical format, Python language, web-publishing framework, Chemistry, Bioinformatics, Enzyme science, Structural Biology, Environmental Sciences, Open Format, Open Data.

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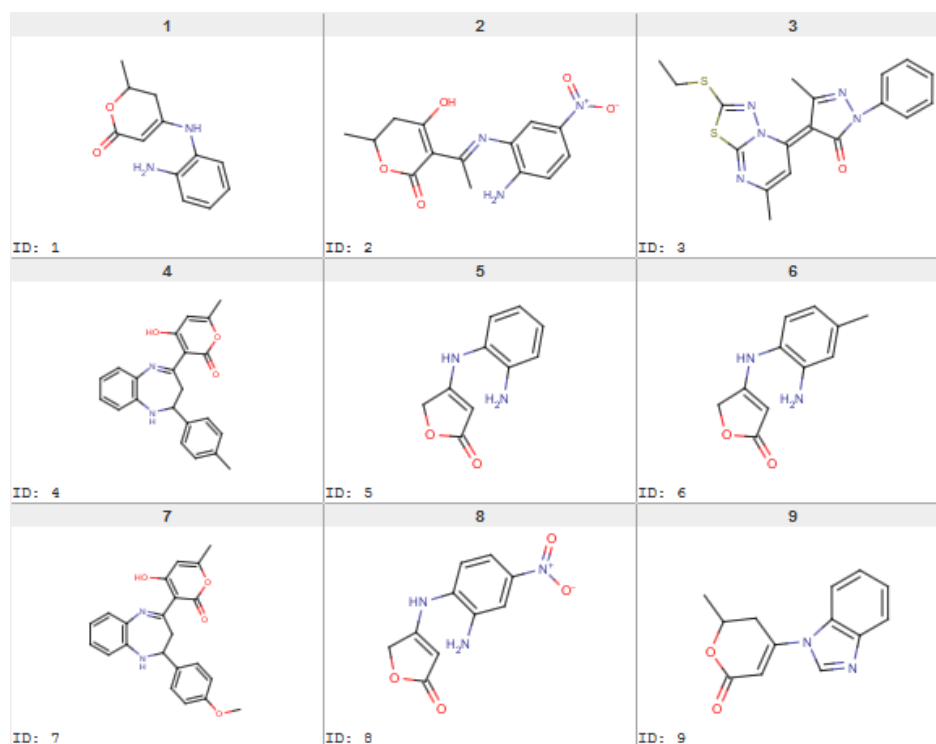
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1. The basics: table, file, annotations and a programming sample

This CSV sample is relative to the field of medicinal chemistry, *courtesy from* Pascal Hoffmann and coll¹². This is the case of a collection of 80 molecular files (af01.mdl, af02.mdl, ..., af80.mdl) in a directory. Each file stores the molecular 2D coordinates (formula) for a given compound, using a given format (here MDL Molfile format³).

Figure 1. –A view⁴ of the molecule's collection.



We wanted to store a table of these molecules, including the molecule's number (*ID*), an identifier (*ident*), the chemist related to the compound synthesis (*lab*), an amount (*vrac*) available of the compound, another compound identifier (*rprod*), the related laboratory notebook identifier (*rlab*), the molecular file itself (*molfile*) including a relative and the compound formula (*smi*) written in another format (here SMILES⁵ chemical format) and included in a cell of the table:

Figure 2. –A view of the collection written in a table, only rows [1..10] and [78..80] are shown.

ID	ident	lab	vrac	rprod	rlab	molfile	smi
1	af01	hoffmann	114	AF01	MAT1	mols\af01.mdl	C1C(OC(=O)C=C1Nc1ccccc1N)C
2	af02	hoffmann	85	AF02	MAT2	mols\af02.mdl	c1(c(ccc1)[N+](=O)[O-])N)/N=C(/C1=C(O)CC(OC1=O)C)C
3	af03	hoffmann	60	AF03	FM401	mols\af03.mdl	n12nc(sc1nc(C)/c/2=C/1\C(=O)N(N=C1C)c1ccccc1)SCC
4	af04	hoffmann	50	AF04	FM5b	mols\af04.mdl	c1(ccc(cc1)C)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2
5	af05	hoffmann	100	AF05	MAT5	mols\af05.mdl	C1(=CC(=O)OC1)Nc1ccccc1N
6	af06	hoffmann	71	AF06	MAT6	mols\af06.mdl	C1(=CC(=O)OC1)Nc1ccc(cc1N)C
7	af07	hoffmann	60	AF07	FM500	mols\af07.mdl	c1(ccc(cc1)OC)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2
8	af08	hoffmann	50	AF08	MAT8	mols\af08.mdl	C1(=CC(=O)OC1)Nc1ccc(cc1N)[N+](=O)[O-]
9	af09	hoffmann	60	AF09	MAT38	mols\af09.mdl	c12c(ccc2)n(cn1)C1=CC(=O)OC(C1)C
10	af10	hoffmann	45	AF10	MAT39	mols\af10.mdl	c12c(ccc(c2)C)n(cn1)C1=CC(=O)OC(C1)C
...							

¹ L. Hammal, S. Bouzroua, C. André, B. Nedjar-Kolli, P. Hoffmann (2007) Versatile Cyclization of Aminoanilino Lactones: Access to Benzotriazoles and Condensed Benzodiazepin-2-thione. *Synthetic Commun.*, 37:3, 501-511.

² M. Fodili, M. Amari, B. Nedjar-Kolli, B. Garrigues, C. Lherbet, P. Hoffmann (2009) Synthesis of Imidazoles from Ketimines Using Tosylmethyl Isocyanide (TosMIC) Catalyzed by Bismuth Triflate. *Lett. Org. Chem.*, 6, 354-358.

³ Chemical table file [Wikipedia] - http://en.wikipedia.org/wiki/Chemical_table_file

⁴ ChemAxon MarvinView - <http://www.chemaxon.com/products/marvin/marvinview/>

⁵ Chemical file format [Wikipedia] - http://en.wikipedia.org/wiki/Chemical_file_format

78	af78	hoffmann	35	AF78	FM406	mol\af78.mdl	c1(ccccc1)N1C(=O)/C(=c\2/n3c(nc(c2)C)sc2c3ccc(c2)OC)/C(=N1)C
79	af79	hoffmann	43	AF79	FM267	mol\af79.mdl	c1(ccccc1)n1c(c(c(n1)C)C(=O)/C=C(\Nc1sc2c(n1)c(ccc2)C)/C)O
80	af80	hoffmann	45	AF80	FM257	mol\af80.mdl	s1cccc1Cn1c(c(nc1)C)c1c(O)cc(C)oc1=O

Then we wanted to write the table: *i*) as a CSV derived file for data and *ii*) with a block for metadata:

Figure 3. –The corresponding CSV file (rows [1..10] and [78..80]).

```

01 → af01 → hoffmann → 114 → AF01 → MAT1 → mol\af01.mdl → c1c(oc(=O)c=c1nc1cccc1N)c
02 → af02 → hoffmann → 85 → AF02 → MAT2 → mol\af02.mdl → c1(c(ccc(c1)[N+](=O)[O-])N)/N=C(/c1=C(O)CC(OC1=O)C)\c
03 → af03 → hoffmann → 60 → AF03 → FM401 → mol\af03.mdl → n12nc(sc1nc(C)c/c/2=C/1\C(=O)N(N=C1C)1cccc1)SCC
04 → af04 → hoffmann → 50 → AF04 → FM5b → mol\af04.mdl → c1(ccc(cc1)C)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2
05 → af05 → hoffmann → 100 → AF05 → MAT5 → mol\af05.mdl → c1(=CC(=O)OC1)Nc1cccc1N
06 → af06 → hoffmann → 71 → AF06 → MAT6 → mol\af06.mdl → c1(=CC(=O)OC1)Nc1ccc(cc1N)C
07 → af07 → hoffmann → 60 → AF07 → FM500 → mol\af07.mdl → c1(ccc(cc1)OC)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2
08 → af08 → hoffmann → 50 → AF08 → MAT8 → mol\af08.mdl → c1(=CC(=O)OC1)Nc1ccc(cc1N)[N+](=O)[O-]
09 → af09 → hoffmann → 60 → AF09 → MAT38 → mol\af09.mdl → c12c(cccc2)n(cn1)c1=CC(=O)OC(C1)C
10 → af10 → hoffmann → 45 → AF10 → MAT39 → mol\af10.mdl → c12c(ccc(c2)C)n(cn1)c1=CC(=O)OC(C1)C
...
78 → af78 → hoffmann → 35 → AF78 → FM406 → mol\af78.mdl → c1(ccccc1)N1C(=O)/C(=c\2/n3c(nc(c2)C)sc2c3ccc(c2)OC)/C(=N1)C
79 → af79 → hoffmann → 43 → AF79 → FM267 → mol\af79.mdl → c1(ccccc1)n1c(c(c(n1)C)C(=O)/C=C(\Nc1sc2c(n1)c(ccc2)C)/C)O
80 → af80 → hoffmann → 45 → AF80 → FM257 → mol\af80.mdl → s1cccc1Cn1c(c(nc1)C)c1c(O)cc(C)oc1=O

#TITLE →
#HEADER → ID → ident → lab → vrac → rprod → rlab → molfile → smi
#TYPE → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT
#WIDTH → 10 → 10 → 10 → 10 → 10 → 10 → 10 → 10
#META →

```

This is what we call a CSV file (CSV with Metadata). In this case the field separator is a TAB (red arrows) but the CSV specification allows any character. The data block is the same than shown in [Figure 2](#) but it is followed by rows beginning with # character and defining a metadata block.

Using flat CSV files (CSV-1 specification) the # rows are considered as metadata rows if the # is immediately followed by a known keyword (TITLE, HEADER, TYPE, WIDTH, META). If the keyword is not recognized by the CSV parser, the row is not processed.

This is also possible to use this property in order to extinct some data rows in the CSV without deleting them; and to annotate the file adding rows for remarks. The following CSV illustrates these two cases outside the metadata block (rows 5, 6, 7). The blank lines are not taken in account by the CSV parser, so they could be used to make the table more readable for humans. As a result, in [Figure 4](#) the compound with ID=3 will not be included in data structure stored in computer's memory.

Figure 4. –A tagged (remarks) CSV file.

```

1 01 → af01 → hoffmann → 114 → AF01 → MAT1 → mol\af01.mdl → c1c(oc(=O)c=c1nc1cccc1N)c
2 02 → af02 → hoffmann → 85 → AF02 → MAT2 → mol\af02.mdl → c1(c(ccc(c1)[N+](=O)[O-])N)/N=C(/c1=C(O)CC(OC1=O)C)\c
3
4
5 # =====
6 # this record is not verified to date
7 #03 → af03 → hoffmann → 60 → AF03 → FM401 → mol\af03.mdl → n12nc(sc1nc(C)c/c/2=C/1\C(=O)N(N=C1C)1cccc1)SCC
8
9 04 → af04 → hoffmann → 50 → AF04 → FM5b → mol\af04.mdl → c1(ccc(cc1)C)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2
10 05 → af05 → hoffmann → 100 → AF05 → MAT5 → mol\af05.mdl → c1(=CC(=O)OC1)Nc1cccc1N
11 06 → af06 → hoffmann → 71 → AF06 → MAT6 → mol\af06.mdl → c1(=CC(=O)OC1)Nc1ccc(cc1N)C
12 07 → af07 → hoffmann → 60 → AF07 → FM500 → mol\af07.mdl → c1(ccc(cc1)OC)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2
13 08 → af08 → hoffmann → 50 → AF08 → MAT8 → mol\af08.mdl → c1(=CC(=O)OC1)Nc1ccc(cc1N)[N+](=O)[O-]
14 09 → af09 → hoffmann → 60 → AF09 → MAT38 → mol\af09.mdl → c12c(cccc2)n(cn1)c1=CC(=O)OC(C1)C
15 10 → af10 → hoffmann → 45 → AF10 → MAT39 → mol\af10.mdl → c12c(ccc(c2)C)n(cn1)c1=CC(=O)OC(C1)C
16
17 ...
18 78 → af78 → hoffmann → 35 → AF78 → FM406 → mol\af78.mdl → c1(ccccc1)N1C(=O)/C(=c\2/n3c(nc(c2)C)sc2c3ccc(c2)OC)/C(=N1)C
19 79 → af79 → hoffmann → 43 → AF79 → FM267 → mol\af79.mdl → c1(ccccc1)n1c(c(c(n1)C)C(=O)/C=C(\Nc1sc2c(n1)c(ccc2)C)/C)O
20 80 → af80 → hoffmann → 45 → AF80 → FM257 → mol\af80.mdl → s1cccc1Cn1c(c(nc1)C)c1c(O)cc(C)oc1=O
21
22 #TITLE →
23 #HEADER → ID → ident → lab → vrac → rprod → rlab → molfile → smi
24 #TYPE → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT
25 #WIDTH → 10 → 10 → 10 → 10 → 10 → 10 → 10 → 10
26 #META →

```

The metadata block is used to embed the minimal canonical information about the table:

- The #TITLE row is for information (one ASCII line) about the table.
- The #HEADER row stores for column titles.
- The #TYPE row stores for the data types for columns. CSV lets you define your data types, only fuzzy and very minimal data types are provided: TEXT (textual data), NUMERIC (columns of integer or float values), BOOLEAN (0/1, Y/N, Yes/No you can define your own code).
- The #WIDTH row is used to store a value significant of column's width.
- The #META row is used to store other/not defined information.

The CSV parser doesn't use the information stored in metadata block: all the information is read as strings and returned as a data structure depending on the programming language used.

Metadata is not interpreted at the parser level and it is a choice. Using this approach makes it possible to add all data types for #TYPE keywords, this is the reason of the reduced number of predefined types in CSV specification. The data type recognition is done only when the data structure is available in memory: this step is uncoupled from the parser step. A side result is that the same parser can be used for all data types.

In the case of Python language, a Class (a `csvm_ptr` object) is returned with some methods (i.e. `csvm_ptr_dump` for display the data structure on standard output). The following code (Code 1) shows the contents of the corresponding data structure. These lines are extracted from a Python toolkit (Pybuild) that support CSV files.

Code 1. – Beginning of Python CSV Class code.

```
class csvm_ptr:
    """
    Follows CSV specs (v:1.x) for contents of data structure. Standard column
    types are NUMERIC,TEXT,DATE,BOOLEAN. Some of us, use also INTEGER, FLOAT
    for numeric types. Some of us, use also NODE, LINK, IMAGE for web data
    embedded in CSV files. WIDTHs (10,50 if not set) are for Javascript tables
    and can be omitted.
    *** 1.01/080304/fred
    """
    def __init__(self):
        self.SOURCE = ""          # path/file name of readed CSV file
        self.CSV = ""            # CSV or CSV depending of file contents
        self.TITLE_N = 0         #Titles of CSV file (let for future, only one string used today)
        self.TITLE = ""          # Title of CSV file
        self.HEADER_N = 0        # Number of data columns titles
        self.HEADER = []         # List of data column titles
        self.TYPE_N = 0          # Number of data columns types (= self.HEADER_N)
        self.TYPE = []           # List of data column types
        self.WIDTH_N = 0         # Number of data columns widths (= self.HEADER_N)
        self.WIDTH = []          # List of data column widths
        self.DATA_R = 0          # Number of data rows
        self.DATA_C = 0          # Number of data columns (= self.HEADER_N)
        self.DATA = []           # String matrix containing data
        self.META = ""           # Meta string
```

The following Python code (Code 2) shows how 1) to load a CSV file (using the TAB "\t" as delimiter) with the function `csvm_ptr_read_extended_csvm`, 2) to display the Class and 3) to free the memory.

Code 2. – Loading and printing a CSV file using Pybuild toolkit.

```
print "> A new blank CSV object"
c = csvm_ptr()
print "> Populates it with a CSV file ... "
c = csvm_ptr_read_extended_csvm(c, file_cleanpath("test/hoffmann.csv"), "\t")
print "> data dump ... "
c.csvm_ptr_dump(0,0)
print "> Clear CSV object"
c.csvm_ptr_clear()
print
```

The corresponding output is shown in [Figure 5](#), the result of `csvm_ptr_dump` method shows the metadata block (pink text, the order of metadata rows is #WIDTH, #TEXT, #HEADER) followed by data rows (all cells are surrounded by brackets). Please notice that the previous annotated CSV file is used: the row with [03] value in first column {number} is missing.

Figure 5. – Console output corresponding to Code 1 example, only 11 first data rows and 5 last data rows are displayed.

```

=> A new blank CSVM object
=> Populates it with a CSVM file ...
=> data dump ...

DUMP: CSVM info {
SOURCE test\hoffmann.csvm
CSV CSVM
META []
TITLE_N 1
TITLE
HEADER_N 15
TYPE_N 15
WIDTH_N 15
0 10 TEXT {number}
1 10 TEXT {name}
2 10 TEXT {plate}
3 10 TEXT {chemist}
4 10 TEXT {amount}
5 10 TEXT {ref_product}
6 10 TEXT {ref_labbook}
7 10 TEXT {id_lab}
8 10 TEXT {id_team}
9 10 TEXT {id_box}
10 10 TEXT {rights}
11 10 TEXT {chr_row_box}
12 10 TEXT {num_col_box}
13 10 TEXT {OpenBabel Symmetry Classes}
14 10 TEXT {smi}
DATA_R 79
DATA_C 15
79 15
0 [01][af01][cob.1][hoffmann][114][AF01][MAT1][CCC][03][01][L][A][02][6 12 7 13 4 10 15 16 9 5 3 14 8 11 2 1][C1C(OC(=O)C=C1Nc1cccc1N)C]
1 [02][af02][cob.1][hoffmann][85][AF02][MAT2][CCC][03][01][L][B][02][19 20 10 12 21 13 16 15 18 11 2 17 3 4 8 14 9 5 1 22 6 7][c1(c(ccc(c1)[N+](=O)[O-])N)/N=C/C1=C(O)CC(OC1=O)C)\C]
2 [04][af04][cob.1][hoffmann][50][AF04][FM5b][CCC][03][01][L][D][02][19 8 7 16 7 8 22 25 15 17 12 24 21 20 14 10 23 18 13 11 5 6 9 4 3 2 1][c1(ccc(cc1)C)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2]
3 [05][af05][cob.1][hoffmann][100][AF05][MAT5][CCC][03][01][L][E][02][12 5 6 11 7 3 9 13 14 8 4 2 10 1][C1(=CC(=O)OC1)Nc1cccc1N]
4 [06][af06][cob.1][hoffmann][71][AF06][MAT6][CCC][03][01][L][F][02][12 4 5 11 6 7 9 14 15 8 13 3 10 2 1][C1(=CC(=O)OC1)Nc1ccc(cc1N)C]
5 [07][af07][cob.1][hoffmann][60][AF07][FM500][CCC][03][01][L][G][02][19 9 8 17 8 9 26 25 16 18 15 23 22 21 14 11 24 20 13 12 5 6 10 4 3 2 7 1][c1(ccc(cc1)OC)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2]
6 [08][af08][cob.1][hoffmann][50][AF08][MAT8][CCC][03][01][L][H][02][13 5 6 12 7 10 8 14 15 11 16 2 9 1 17 3 4][C1(=CC(=O)OC1)Nc1ccc(cc1N)[N+](=O)[O-]]
7 [09][af09][cob.1][hoffmann][60][AF09][MAT38][CCC][03][01][L][A][03][16 15 7 3 11 6 4 8 17 9 12 5 13 14 10 2 1][c12c(cccc2)n(cn1)C1=CC(=O)OC(C1)C]
8 [10][af10][cob.1][hoffmann][45][AF10][MAT39][CCC][03][01][L][B][03][17 16 6 5 11 7 14 8 18 9 12 4 13 15 10 2 1 3][c12c(ccc(c2)C)n(cn1)C1=CC(=O)OC(C1)C]
9 [11][af11][cob.1][hoffmann][43][AF11][MAT36][CCC][03][01][L][C][03][17 16 12 15 9 10 11 8 18 3 7 5 14 13 6 4 1 2][c12c(cc(cc2)C)n(cn1)C1=CC(=O)OC(C1)C]
10 [12][af12][cob.1][hoffmann][45][AF12][FM5d][CCC][03][01][L][D][03][26 24 13 6 8 10 27 21 17 18 16 20 19 25 14 12 22 23 15 9 7 5 11 4 3 1 2][c1(c(cccc1)O)C1Nc2c(N=C(c3c(=O)oc(cc3O)C)C1)cccc2]
...
74 [76][af76][cob.1][hoffmann][50][AF76][FM414][CCC][03][01][L][D][11][22 9 14 11 19 13 10 1 20 12 21 18 15 8 17 7 6 16 7 5 6 4 3 2][n12nc(sc1nc(C)c/c2=C/1\C(=O)N(N=C1C)c1cccc1S]
75 [77][af77][cob.1][hoffmann][45][AF77][FM410][CCC][03][01][L][E][11][18 8 6 5 6 8 24 20 17 9 19 26 21 11 15 13 22 23 25 12 7 16 10 14 4 2 3 1][c1(ccccc1)N1C(=O)/C(=c\2/n3c(nc(c2)C)sc2c3ccc(c2)C)/C(=N1)C]
76 [78][af78][cob.1][hoffmann][35][AF78][FM406][CCC][03][01][L][F][11][19 9 6 5 6 9 25 21 18 10 20 27 22 12 16 14 23 24 26 13 8 17 11 15 4 2 3 7 1][c1(ccccc1)N1C(=O)/C(=c\2/n3c(nc(c2)C)sc2c3ccc(c2)OC)/C(=N1)C]
77 [79][af79][cob.1][hoffmann][43][AF79][FM267][CCC][03][01][L][G][11][21 16 7 6 7 16 26 22 7 11 20 9 17 10 19 23 25 14 18 15 24 12 8 13 2 5 1 4 3][c1(ccccc1)n1c(c(c(n1)C)C(=O)/C=C\Wc1sc2c(n1)c(ccc2)C)/C)O]
78 [80][af80][cob.1][hoffmann][45][AF80][FM257][CCC][03][01][L][H][11][8 6 5 7 15 16 3 12 14 1 19 11 2 21 13 17 18 4 9 20 10][s1cccc1Cn1c(c(nc1)C)c1c(O)cc(O)c1=O]
}
done
=> Clear CSVM object

```

2. Advanced: combining with databases

CSVM is obviously not a substitute for databases, but it is possible to use a CSVM table for storing simple database schema and database tables. The following sample shows this kind of approach in the case of a chemical inventory. These files were generated after analysis (SQL requests) of a relational database system (Hughes Technologies mQSL) and prior to be converted and injected in another RDBMS (MySQL or PostgreSQL).

2.1. Database tables and schema - Chemical inventory

A simple chemical database of two tables and a relation between the column of indice=3 (from 0 to n-1) in table 'assoc' to column of indice=0 in table 'user'. Note that the schema contains heterogeneous data in columns (data from keyword TABLES, or FOREIGN) so all columns titles are set to the same word 'DATA', *courtesy from Anne Laure Leomant⁶ and Nathalie Gouardères.*

Figure 6. – Example of database schema encoded in a CSVM file.

```
DB      192.168.10.17  anonymous      -      Chemb
TABLE   user           chemb\user.csv  0      -
TABLE   assoc       chemb\assoc.csv 0      -
FOREIGN assoc       3              user    0

#TITLE   Database import schema
#HEADER  KEYWORD  DATA  DATA  DATA  DATA
#TYPE    TEXT     TEXT  TEXT   TEXT   TEXT
#WIDTH   10      10    10     10     10
#META    april/25/2008 15:42:45
```

The database tables are a particular case of CSVM file. Database systems have specific data types (i.e. REAL, INTEGER, SMALLINT ...) depending on the records used in tables that are more precise than the generic CSVM type (NUMERIC in this case). But the CSVM approach allows the use of database types in metadata #TYPE row. Also in row #WIDTH it is possible to store the dimension of database fields when it is needed by database type. The following example shows the last 3 lines of CSVM file `assoc.csvm` corresponding to the 'assoc' table included in of the previous schema:

Figure 7. – Example of a database table encoded in a CSVM file.

```
...
7374  124-63-0  acros    88  12564  CH3C102S  1  14-Apr-2006  16:02:07  methane sulfonyl chloride  0.00  100ml
7375  7758-99-8  billault 90  40-3588  CuO4S    1  28-Mar-2006  17:15:13  cuivre sulfate             0.00  500g
7376  142-96-1  acros    48  14969  C8H180   1  28-Mar-2006  17:20:00  di n-butyl ether           0.00  11

#TITLE   assoc
#HEADER  clef  cas    four  uid  ref  formule  quantite  date  heure  nom  prix  cond
#TYPE    INT  NCHAR  CHAR  INT  CHAR  CHAR     INT      CHAR  TIME  CHAR  CHAR  CHAR
#WIDTH   5   16    16    3   16   18       2       11    8    48   7    16
#META    april/23/2008 10:30:45
```

Notice that other information on databases (i.e. data types for descriptions and conversions) could be also encoded in CSVM files.

So CSVM, even if it is primarily conceived for handling RAW data is also usable to get a 'flat' view of simple database systems. It could be used as a canonical format to do a lot of transformation between databases, without any modification of CSVM parsers.

⁶ A.L. Leomant. Evaluation d'un SGBDR pour une chimiothèque. Rapport L3 SID, Université Paul Sabatier Toulouse III (2008).

3. Indexes, collection of files, documents ...

Initially CSVSM was used primarily for indexing purposes: file's catalog (filenames/paths and some metadata about the files) rather than as a substitute for spreadsheets tables. The following figure shows this kind of use: the result for a recursive directory list.

Figure 8. – Example of using a CSVSM table for a file catalog (this output is a dump of a CSVSM Python class).

```
DUMP: CSVSM info {
SOURCE
CSV CSVSM
META []
TITLE_N 1
TITLE
HEADER_N 3
TYPE_N 3
WIDTH_N 3
0 50 TEXT {DIR}
1 50 TEXT {FILE}
2 50 TEXT {-}
DATA_R 118
DATA_C 3
118 3
0 [c:\Python26\Lib\site-packages\build][args.py][18:jun:2009]
1 [c:\Python26\Lib\site-packages\build][date.py][19:oct:2006]
2 [c:\Python26\Lib\site-packages\build][dir.py][06:may:2010]
...
115 [c:\Python26\Lib\site-packages\build\depot\numerix\_dist][__init__.py][16:apr:2008]
116 [c:\Python26\Lib\site-packages\build\_struct][strpdb.py][26:mar:2007]
117 [c:\Python26\Lib\site-packages\build\_struct][__init__.py][31:oct:2006]
}
done
```

Notice that the last column is filled by the date of creation of files, but the column name is pending (the #HEADER value is a blank) before saving data in a CSVSM file. *This is a very interesting feature of CSVSM approach related to RAW data: in some circumstances the name of a column is not yet defined but the data must be stored anyway.*

3.1. Software pipes - Molecular calculations

CSVSM can be useful for launching calculations, you can store in the CSVSM the operations to do, including launching external programs, store intermediate results (or index filenames) and complete or add columns from step to steps. Eventually a software component using a CSVSM parser can generate from these tables, information (files, parameters) needed by job schedulers/batch processors.

Very often calculations are applied to data files, so this is a particular case of indexing. The interest of CSVSM space is that all steps (defining files, parameters, results) could be made easily and in the same data format (ready for use by spreadsheets or scientific data analysis and graphic software).

In the next section, a complete example will be shown (case of Enzyme kinetics). Here a simpler sample issued from [structural biology/rational drug design] science is given. The data shown is a CSVSM file, used by a software component added to UCSF Chimera^{7 8} software and used to launch X-Score^{9 10} calculations inside the interface, *courtesy from Mansi Trivedi*¹¹.

Each calculation uses a ligand and a protein. For a given ligand, one or two torsion angles were selected using the Chimera GUI and different files (`PARAM` column) corresponding to different angular values

⁷ UCSF Chimera - <http://www.cgl.ucsf.edu/chimera/>

⁸ UCSF Chimera--a visualization system for exploratory research and analysis. E.F. Pettersen, T.D. Goddard, C.C. Huang, G.S. Couch, D.M. Greenblatt, E.C. Meng, T.E. Ferrin. *J Comput. Chem.* (2004) 25:13, 1605-1612.

⁹ Renxiao Wang – <http://sw16.im.med.umich.edu/software/xtool/>

¹⁰ R. Wang, Y. Lu, S. Wang. Comparative Evaluation of 11 Scoring Functions for Molecular Docking. *J. Med. Chem.* (2003) 46:12, 2287-2303.

¹¹ M. Trivedi – *In silico* approach to study protein-ligand binding affinity. Rapport M2P Bioinformatique, Université Paul Sabatier (2009).

('ANG1' and 'ANG2' columns) were generated. For calculation the corresponding log files were added ('LOG' column) in the CSV table.

Figure 9. – Calculations driven by a CSV table.

```

0 5 4 - C:/Local/Chimera/calc/ligand_5_4.mol2 C:/Local/Chimera/calc/ligand_5_4_xscore.log -
1 5 24 - C:/Local/Chimera/calc/ligand_5_24.mol2 C:/Local/Chimera/calc/ligand_5_24_xscore.log -
2 5 44 - C:/Local/Chimera/calc/ligand_5_44.mol2 C:/Local/Chimera/calc/ligand_5_44_xscore.log -
3 5 64 - C:/Local/Chimera/calc/ligand_5_64.mol2 C:/Local/Chimera/calc/ligand_5_64_xscore.log -
4 40 4 - C:/Local/Chimera/calc/ligand_40_4.mol2 C:/Local/Chimera/calc/ligand_40_4_xscore.log -
5 40 24 - C:/Local/Chimera/calc/ligand_40_24.mol2 C:/Local/Chimera/calc/ligand_40_24_xscore.log -
6 40 44 - C:/Local/Chimera/calc/ligand_40_44.mol2 C:/Local/Chimera/calc/ligand_40_44_xscore.log -
7 40 64 - C:/Local/Chimera/calc/ligand_40_64.mol2 C:/Local/Chimera/calc/ligand_40_64_xscore.log -
8 75 4 - C:/Local/Chimera/calc/ligand_75_4.mol2 C:/Local/Chimera/calc/ligand_75_4_xscore.log -
9 75 24 - C:/Local/Chimera/calc/ligand_75_24.mol2 C:/Local/Chimera/calc/ligand_75_24_xscore.log -
10 75 44 - C:/Local/Chimera/calc/ligand_75_44.mol2 C:/Local/Chimera/calc/ligand_75_44_xscore.log -
11 75 64 - C:/Local/Chimera/calc/ligand_75_64.mol2 C:/Local/Chimera/calc/ligand_75_64_xscore.log -
12 110 4 - C:/Local/Chimera/calc/ligand_110_4.mol2 C:/Local/Chimera/calc/ligand_110_4_xscore.log -
13 110 24 - C:/Local/Chimera/calc/ligand_110_24.mol2 C:/Local/Chimera/calc/ligand_110_24_xscore.log -
14 110 44 - C:/Local/Chimera/calc/ligand_110_44.mol2 C:/Local/Chimera/calc/ligand_110_44_xscore.log -
15 110 64 - C:/Local/Chimera/calc/ligand_110_64.mol2 C:/Local/Chimera/calc/ligand_110_64_xscore.log -
16 145 4 - C:/Local/Chimera/calc/ligand_145_4.mol2 C:/Local/Chimera/calc/ligand_145_4_xscore.log -
17 145 24 - C:/Local/Chimera/calc/ligand_145_24.mol2 C:/Local/Chimera/calc/ligand_145_24_xscore.log -
18 145 44 - C:/Local/Chimera/calc/ligand_145_44.mol2 C:/Local/Chimera/calc/ligand_145_44_xscore.log -
19 145 64 - C:/Local/Chimera/calc/ligand_145_64.mol2 C:/Local/Chimera/calc/ligand_145_64_xscore.log -

#TITLE XSCORE For 2 TORSION
#HEADER IND ANGL1 ANGL2 PARAM PDB LIG LOG COF
#TYPE STRING STRING STRING STRING STRING STRING STRING STRING
#WIDTH 300 300 300 300 300 300 300 300

```

After calculation, the log files generated by X-SCORE were parsed, converted in CSV tables, and the tables were indexed (extraction of score values). Then, the score values were aggregated with the upper table (addition of columns). All the process is driven by the table of Figure 9 that contains all information needed for calculation and collection of results. In this case the first column ('IND') is used for store indices of rows but it can be used in *key/values* scheme, as it was done in the following sample:

Figure 10. – Calculation and parameters driven by a CSV table, columns PHI and PSI are used for angle values.

```

1
2 PDB --> dock_test2.pdb -->
3 FIXPDB --> e:\test\xscoretest\data>e:\test\xscoretest\input -->
4 FIXMOL2 --> e:\test\xscoretest\data>e:\test\xscoretest\input -->
5 SCOREPATHS --> e:\test\xscoretest\input --> e:\test\xscoretest\output -->
6 XSCOREIN -->
7 ANG>0 --> mol507_0_xscore.mol2 --> mol507_0_xscore.log -->
8 ANG>20 --> mol507_20_xscore.mol2 --> mol507_20_xscore.log -->
9 ANG>40 --> mol507_40_xscore.mol2 --> mol507_40_xscore.log -->
10 ANG>60 --> mol507_60_xscore.mol2 --> mol507_60_xscore.log -->
11 ANG>80 --> mol507_80_xscore.mol2 --> mol507_80_xscore.log -->
12 ANG>100 --> mol507_100_xscore.mol2 --> mol507_100_xscore.log -->
13 ANG>120 --> mol507_120_xscore.mol2 --> mol507_120_xscore.log -->
14 ANG>140 --> mol507_140_xscore.mol2 --> mol507_140_xscore.log -->
15 ANG>160 --> mol507_160_xscore.mol2 --> mol507_160_xscore.log -->
16 ANG>180 --> mol507_180_xscore.mol2 --> mol507_180_xscore.log -->
17 ANG>200 --> mol507_200_xscore.mol2 --> mol507_200_xscore.log -->
18 ANG>220 --> mol507_220_xscore.mol2 --> mol507_220_xscore.log -->
19 ANG>240 --> mol507_240_xscore.mol2 --> mol507_240_xscore.log -->
20 ANG>260 --> mol507_260_xscore.mol2 --> mol507_260_xscore.log -->
21 ANG>280 --> mol507_280_xscore.mol2 --> mol507_280_xscore.log -->
22 ANG>300 --> mol507_300_xscore.mol2 --> mol507_300_xscore.log -->
23 ANG>320 --> mol507_320_xscore.mol2 --> mol507_320_xscore.log -->
24 ANG>340 --> mol507_340_xscore.mol2 --> mol507_340_xscore.log -->
25 ANG> --> mol507_xscore.mol2 --> mol507_xscore.log -->
26
27 #TITLE --> X-Score index file
28 #HEADER --> KW --> PHI --> PSI --> MOL2 --> LOG --> SCORE
29 #TYPE --> STRING --> NUMERIC --> NUMERIC --> STRING --> NUMERIC
30 #WIDTH --> 50 --> 50 --> 50 --> 50 --> 50
31 #META --> Test file not definitive format
32

```

The first column labeled 'KW' stores keyword that drives contents of next columns, and the last column ('SCORE') is present and ready to store score values. The CSV file is used to store parameters/results of calculations (rows 7-25) as well as configuration options of X-SCORE (rows 2-6). This case shows that in the same format space we can store configurations, parameters, results ...

4. From indexes to collections: files, information and dataflows

The concept was then used for indexes and for tabular data in the same or different collections. For a same catalog, the two next samples show differences (files *vs.* tables) on indexed contents: *i*) the case of a publishing chain (for standard files) and *ii*) the case of a collection of sparse environmental data (for CSV files). These two examples are issued from scientific works in the field of environmental sciences, but making collections of RAW/processed objects is a general problem in all sciences.

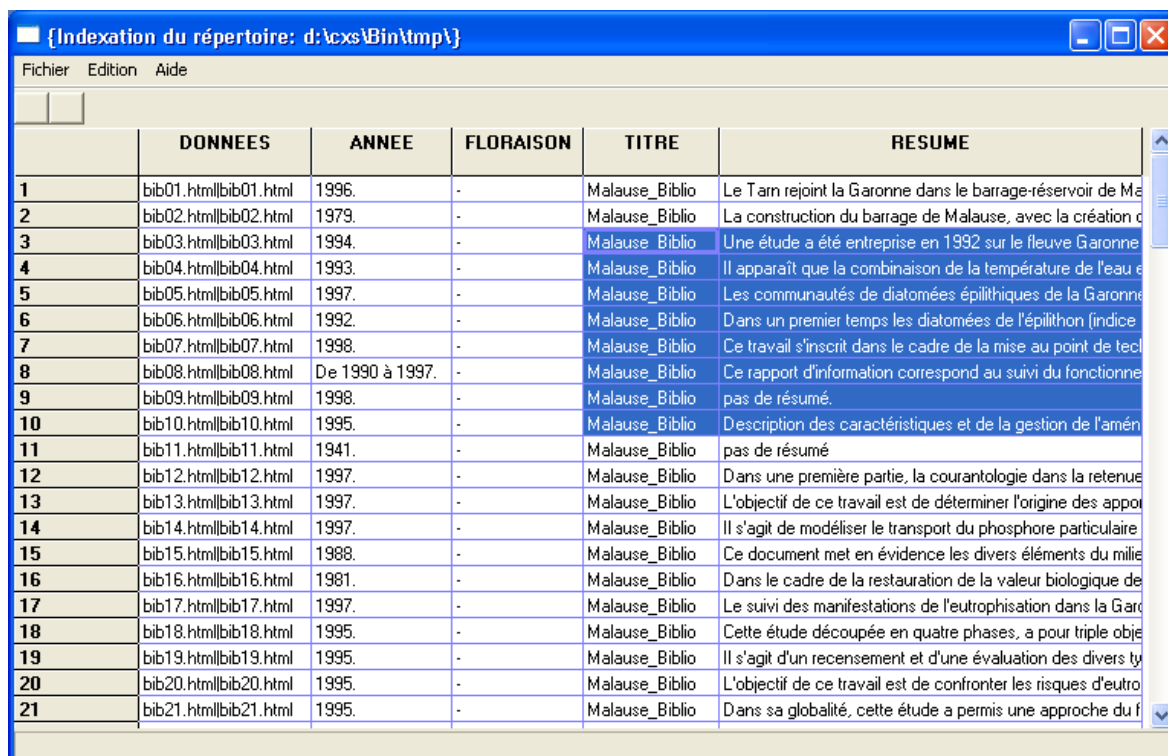
4.1. A publishing chain

The process starts from a set of documents (books, articles, reports, various documents) about a particular place (Malause) known as a sedimentation zone in a river (Garonne, France). *Courtesy from Philippe Vervier, Sarah Gimet*¹² and Gerome Beyries¹³.

This grey information (because all documents are not indexed in bibliographic systems) was digitized (PDF to ASCII text) and contents were extracted (text filters), normalized and saved in Open Office documents (using a common template). This new collection of documents was then used as a primary data source by software components: converted in XML (XML DocBook¹⁴), HTML¹⁵ and analyzed using a custom XML parser. Pertinent fields of information found in documents (year, title, abstract ...) were extracted and saved in a CSV file. *Courtesy from Jean-Olivier Butron*¹⁶.

The following figure shows a Perl CSV parser that displays CSV data in a grid widget. The column 'DONNEES' contains links to the corresponding HTML documents (URL|Name of URL), 'ANNEE' stands for year, 'TITRE' for title and 'RESUME' for abstract.

Figure 11. –A simple viewer (wxGrid widget¹⁷) parsing a CSV file (a collection of documents).



	DONNEES	ANNEE	FLORAISON	TITRE	RESUME
1	bib01.html bib01.html	1996.	-	Malause_Biblio	Le Tam rejoint la Garonne dans le barrage-réservoir de Mala
2	bib02.html bib02.html	1979.	-	Malause_Biblio	La construction du barrage de Malause, avec la création d
3	bib03.html bib03.html	1994.	-	Malause_Biblio	Une étude a été entreprise en 1992 sur le fleuve Garonne
4	bib04.html bib04.html	1993.	-	Malause_Biblio	Il apparaît que la combinaison de la température de l'eau e
5	bib05.html bib05.html	1997.	-	Malause_Biblio	Les communautés de diatomées épilithiques de la Garonne
6	bib06.html bib06.html	1992.	-	Malause_Biblio	Dans un premier temps les diatomées de l'épilithon (indicie
7	bib07.html bib07.html	1998.	-	Malause_Biblio	Ce travail s'inscrit dans le cadre de la mise au point de tecl
8	bib08.html bib08.html	De 1990 à 1997.	-	Malause_Biblio	Ce rapport d'information correspond au suivi du fonctionne
9	bib09.html bib09.html	1998.	-	Malause_Biblio	pas de résumé.
10	bib10.html bib10.html	1995.	-	Malause_Biblio	Description des caractéristiques et de la gestion de l'amén
11	bib11.html bib11.html	1941.	-	Malause_Biblio	pas de résumé
12	bib12.html bib12.html	1997.	-	Malause_Biblio	Dans une première partie, la courantologie dans la retenue
13	bib13.html bib13.html	1997.	-	Malause_Biblio	L'objectif de ce travail est de déterminer l'origine des appor
14	bib14.html bib14.html	1997.	-	Malause_Biblio	Il s'agit de modéliser le transport du phosphore particulaire
15	bib15.html bib15.html	1988.	-	Malause_Biblio	Ce document met en évidence les divers éléments du milie
16	bib16.html bib16.html	1981.	-	Malause_Biblio	Dans le cadre de la restauration de la valeur biologique de
17	bib17.html bib17.html	1997.	-	Malause_Biblio	Le suivi des manifestations de l'eutrophisation dans la Garc
18	bib18.html bib18.html	1995.	-	Malause_Biblio	Cette étude découpée en quatre phases, a pour triple obje
19	bib19.html bib19.html	1995.	-	Malause_Biblio	Il s'agit d'un recensement et d'une évaluation des divers ty
20	bib20.html bib20.html	1995.	-	Malause_Biblio	L'objectif de ce travail est de confronter les risques d'euro
21	bib21.html bib21.html	1995.	-	Malause_Biblio	Dans sa globalité, cette étude a permis une approche du f

¹² M. Gerino & coll. Bilan et dynamique de la matière organique et des contaminants au sein d'une discontinuité ex de la retenue de Malause. Restitution des travaux scientifiques du projet de recherche ECOBAG P2 (2006).

¹³ G. Beyries. Composants logiciels génériques pour les collections de données. Mémoire ingénieur Ecole des Techniques du Génie Logiciel (2004).

¹⁴ XML DocBook - <http://www.docbook.org/>

¹⁵ Norman Walsh style sheets - <http://nwalsh.com/>

¹⁶ F. Rodriguez, G. Beyries, J.O. Butron, C. Blonski. Maîtriser l'information chimique au quotidien: méthodes et composants logiciels à l'interface chimiebiologie. JECB21, Toulouse (2005).

¹⁷ wxWidgets Cross-Platform GUI Library - <http://www.wxwidgets.org/>

In this case the CSV file corresponds to the index of this collection and embeds information extracted from documents. At the CSV level: this is data; at indexed files (XML files converted in HTML files) level: this is metadata. All the different tasks were processed automatically without data input ‘at hand’ by an operator. A lot of conversion between files (Raw text, HTML, XML, XML-Docbook, HTML) were also configured and driven by CSV files. The following figure shows a CSV file used to drive changes in the CSV index itself (each row is a filter applied to the index).

Figure 12. – Some rows of a CSV file filled with rules for a text filter.

```
STR → {HTML : .} → - - - - -
STR → XML : index : → Indexation du répertoire : → - - - - -
STR → dbk.html → .html → - - - - -
STR → dbk.htm → .html → - - - - -
STR → _TABAbstract → _TABRESUME → - - - - -
STR → _TABabstract → _TABRESUME → - - - - -
STR → _TABTitle → _TABTITRE → - - - - -
STR → _TABtitle → _TABTITRE → - - - - -
STR → _TABfile → _TABDONNEES → - - - - -
STR → _TABFile → _TABDONNEES → - - - - -
STR → hashtablee2{ANNEE} → ANNEE → - - - - -
STR → hashtablee2{FLORAISON} → FLORAISON → - - - - -

#TITLE → Change some values in CSV indexes
#HEADER → CODE → TAG → TAG → TAG → TAG → TAG → TAG → TAG → TAG
#TYPE → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT → TEXT
#WIDTH → 50 → 50 → 50 → 50 → 50 → 50 → 50 → 50 → 50 → 50
```

The first column is used to store the type of filter using a keyword. The keyword STR is used to specify a string replacement of the value stored in second column by the value stored by the third column. More complex filters that ‘STR’ exist and need to use fully the 10 columns. Then the CSV index was integrated with a JavaScript framework in order to provide a view of the collection dedicated to end users. Buttons at the top of dynamic table are used to sort data of CSV index by a given column.

Figure 13. – Same data (as shown in Figure 10) parsed using a JavaScript list component and displayed as item list (bibliographic style) rather than columns. Data are sorted by year (ANNEE) and displayed in the order (ANNEE, DONNEES, FLORAISON, TITRE, RESUME).

When the publishing chain works, all to do is to write the CSV file and to process the chain. Edition of the CSV index can be done, manually (text editor), using a GUI, automatically by an indexer component). In this case the documents were stored in HTML (after processing) or native (hyperlink to the corresponding file or image) formats.

The following figure shows final results for this kind of publishing chain: *14-A*) the case of small organic molecules (ligands, inhibitors) interacting with some enzymes (GlcCerase, beta-glucosidase) and *14-B*) the case of enzymes itself (HAT).

Figure 14-A. – Chemical collection of ligands taken from the Protein Data Bank ¹⁸.

TYPE	IMG	NOM	SKETCH	MOL	VIEW
inh		IFM 5-hydroxymethyl- 3,4-dihydroxypiperidine isofagomine inhibiteur BglA, 1OIF	mol-22.sk.c	ifm.pdb	ifm-h
inh		NOJ 1-deoxynojirimycin inhibiteur BglA, 1OIM	mol-23.sk.c	noj.pdb	noj-h
inh		g2f 2,3,4,6-tetra-O-acetyl- beta-D-glucopyranose inhibiteur BglA, 1OIM	mol-24.sk.c	g2f.pdb	g2f-h
inh		ifl 1,2:3,5-di-O-isopropylidene- beta-D-fructofuranose inhibiteur BglA, 1OIM	mol-25.sk.c	ifl.pdb	ifl-h
inh		oxz oxazolone inhibiteur BglA, 1OIM	mol-26.sk.c	oxz.pdb	oxz-h

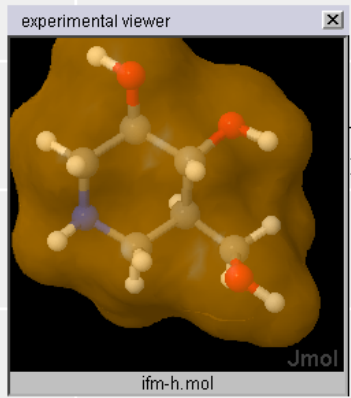


Figure 14-B. – Collection of macro-molecules (enzymes) structures taken from the Protein Data Bank.

Reports - Capture - Ligands - **Structures** - Manuals - Ligsum - Ligdata - PDBclean - Loopsum

Histone Acetyl Transferase

La table (v:1.01) regroupe les structures PDB en cours et offre une vue synthétique sur la collection. Les lignes incomplètes correspondent à des structures en cours d'annotation.

Links ^[1] available from this list.

FILE FLD LINK JMOL IMAGE ORGSCI MOL

FILE	FLD	IMAGE	PDB	JMOL	ORGSCI	MOLEC
1B6B	f		1B6B	1B6B	ovis aries	arylalky n-acetyl
1B87	f		1B87	1B87	enterococcus faecium	aminogly n6'-acet type 1
1B04	f		1B04	1B04	serratia marcescens	serratia aminogly
1B0B	f		1B0B	1B0B	saccharomyces cerevisiae	histone acetyltransferase
1CJW	f		1CJW	1CJW	ovis aries	serotonin n-acetyltransferase

1B87.pdb

Jmol

¹⁸ RCSB Protein Data Bank – <http://www.pdb.org/> a main source for crystallographic structures of proteins.

Courtesy from Marjorie Catala¹⁹; Casimir Blonski, Christian Lherbet and Cecile Baudoin-Dehoux^{20 21}.

The previous figure (14-A) shows the case of a chemical collection (small organic molecules). The collection mixes images ('IMG' column), some information (TYPE, NOM), 2D (SKETCH) and native 3D (MOL) structures and action buttons (VIEW) launching an external viewer²² embedded (HTML layer) in a floating window. In the case of Figure 14-B, the interface of the molecular collection (macromolecules) is multimodal and could launch local/hyperlinks ('FILE'/'LINK' columns), PyMol²³ generated images ('IMAGE') of structure, a viewer ('JMOL') as other information.

From a CSV file point of view (RAW data) the samples of Figure 13-14-15 are of the same complexity despite the kind of objects displayed. All is taken in account by a different JavaScript framework: the CSV file data were converted in JavaScript tables and were processed by dedicated JavaScript components. The Following Code shows the contents of this kind of JavaScript table and the first row for the molecular structure displayed in the case of Figure 14-A:

Code 3. – A JavaScript table descending from a CSV file.

```
// GXMLPARSER interface, download these data before,
// and use GXMLDISPLAY for use with browsers

function jmol(mol,button)
{
    var src = "<INPUT onclick=" + "'" + "loadwindow('jmol.html',250,287,'" + mol +
    "');" + "'" + "type=button value='" + button + "'>";
    return(src);
}

var flags_array = new Array ("TYPE", "SECTOR", "IMG", "NAME", "SKETCH", "MOL", "VIEW");
var flags_n=7;
var getf_array = new Array ("TYPE", "IMG", "NAME", "SKETCH", "MOL", "VIEW");
var getf_n=6;
var data_c=7;

var data_array = new Array();
var data_r=0;
...
data_array[data_r++] = new Array ("inh", "glccerase", "<IMG SRC='sketch/mol-22.png'>",
"IFM<br>5-hydroxymethyl-3,4-dihydropiperidine<br>isofagomine<br>inhibiteur BglA, 1OIF",
"<A HREF='sketch/mol-22.sk'>mol-22.sk</A>", " <A HREF='lig/ifm.pdb'>ifm.pdb</A>",
jmol('ifm-h.mol','ifm-h'));
...
```

It is easy to transform a CSV file in another tabular data (i.e. JavaScript, CSV, XML tables) format because all that is needed to do the transformation is embedded in the CSV file itself.

The corresponding index CSV files can be edited manually by users (as they were using spreadsheet tables) or automatically generated, but the CSV mechanism has the advantage of providing a lot of features for further processing and transformations.

¹⁹ M. Catala - Etude structurale et arrimage moléculaire (Docking) *in silico* de prodrogues dans la β -glucocérébrosidase. Rapport M1BBT (2006) Université Paul Sabatier.

²⁰ S. Gau, F. Rodriguez, C. Lherbet, C. Menendez, M. Baltas. Approches interdisciplinaires pour la conception rationnelle d'inhibiteurs. RECOB 13, Aussois (2010).

²¹ New potent bisubstrate inhibitors of histone acetyltransferase p300: design, synthesis and biological evaluation. F.H.A. Kwie, M. Briet, D. Soupaya, P. Hoffmann, M. Maturano, F. Rodriguez, C. Blonski, C. Lherbet, C. Baudoin-Dehoux. *Chem Biol Drug Des.* (2011) 77:1, 86-92.

²² JMOL: An open-source Java viewer for chemical structures in 3S – <http://jmol.sourceforge.net/>

²³ PyMol – <http://www.pymol.org/>

4.2. Environmental tabular data

This sample shows a way to dig inside files when the indexed files are also CSV files. An index was also created but each entry was linked to a CSV file rather than a text document (Open Office, XML, HTML, PDF ...) or a data source (data file, image, hyperlinks).

In this case, the CSV data (table of measurements, not index) were written manually using a standard spreadsheet. *Courtesy from Magali Gerino, Maya Lauriol, Sabine Sauvage*²⁴ and Gerome Beyries²⁵.

As in previous case, end users can download corresponding CSV files using an hyperlink in index display, but also they can view (after another CSV conversion in JavaScript tables of each leaf files) directly inside these files using another JavaScript framework²⁶, as shown by the following figure:

Figure 15. – A dynamic view of a CSV file after its conversion (JavaScript table, columns at right not shown).

[malause1.js] Malause station 1

date	heure	numero station	coordonnees gps	profondeur	methode de mesure	te
2003/07/07	12:00	1	0346600-4884900	0	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	0.25	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	1.75	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	2	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	2.25	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	2.5	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	2.75	sonde manuel	
2003/07/07	12:00	1	0346600-4884900	3	sonde manuel	

Showing No. 1 to 13 of 13

Navigation controls: |< < Query * > >|

In this table the #WIDTH information is used for column's widths and #TYPE values are injected as column's names in the JavaScript tables.

CSV can be used as well as for index and for tables. This approach has a great potential for complex processing of files because one parser and one format are used independently of tabular objects and information.

²⁴ M. Gerino & coll. Bilan et dynamique de la matière organique et des contaminants au sein d'une discontinuité ex de la retenue de Malause. Restitution des travaux scientifiques du projet de recherche ECOBAG P2 (2006).

²⁵ G. Beyries, F. Rodriguez. Quels outils informatiques pour les collections de données ? Restitution du Programme ECOBAG P1, Agen (2004).

²⁶ A modified version of Dieter Bunger's (<http://www.infovation.de>) Cross Browser Dynamic HTML Tables code - http://www.insidedhtml.com/tips/databinding/ts06/paper/dhtml_table.htm.

4.3. Working with RAW tables

Typically a case encountered in collaborative projects, in which the aggregation of RAW data from different scientific fields (and different data types by definition) is expected. The CSVM concept makes easy to compute intersection or union of tables which could be often useful. These operations are possible even if the tables cover the 1) same, 2) a common core, and 3) different sets of data. In Python CSVM toolkit the union or intersection of CSVM tables use `csvm_ptr()` objects :

```
print "\n*** Test completely different CSVM files"
c1 = csvm_ptr()
c1 = csvm_ptr_read_extended_csvm(c1, "test/test1.csvm", "\t")
c2 = csvm_ptr()
c2 = csvm_ptr_read_extended_csvm(c2, "test/test2.csvm", "\t")
```

The two CSVM files were loaded in two `csvm_ptr()` objects `c1` and `c2`, now the intersection of the two tables can take place, the result is a new `csvm_ptr()` named `r` :

```
print "=> Compute INTERSECTION"
r = csvm_ptr_intersect(c1, c2)
if (r != None):
    r.csvm_ptr_dump(0,0)
    r.csvm_ptr_clear()
else:
    print "No data found"
```

The union of tables is coded as same, and after operation all data structures are destroyed ²⁷.

```
print "\n=> compute UNION"
r = csvm_ptr_union(c1, c2)
r.csvm_ptr_dump(0,0)
r.csvm_ptr_clear()
c1.csvm_ptr_clear()
c2.csvm_ptr_clear()
print "\n*** Test csvmutil done."
```

Notice that after (and sometimes before) union of tables with different (names and types) columns, the RAW data can be sparse, typically a big number of columns, without data for some (or nearly all) rows. If same columns are found in `c1` and `c2` tables, the union mechanism preserves data because data blocks (rows) are added successively and not interleaved.

For an attentive reader, just remark something on [Figure 13](#): the presence of a file named *pinguicula_vulgaris*. This file contains information about a carnivorous plant (butterworts) which seems not to be included in the same category than the others (bibliographic digests about stations around Malause: *malause_biblio*): this is the result of union of tables from different CSVM files.

In the case of *pinguicula* a year was not extracted or assigned from data file, and the resulting cell in final CSVM file is marked by the empty character (‘-’ in this case), the CSVM *pinguicula* branch has not this column. The last column is an excerpt of the [abstract/beginning of the text] found in files. The *pinguicula* type CSVM files have not this kind of information. But, the text recognized as title (not filename) inside the document was assigned to a particular column. This column was later and prior to union of tables set as equivalent to the ‘RESUME’ column of *malause_biblio* branch. So this column is filled in the resulting CSVM file that was processed in the dataflow. Another mechanism was also used: adding ²⁸ a column ‘FLORAIISON’ (flowering date) not found in the *malause_biblio* branch and obviously found in *pinguicula* branch.

This shows how it is possible to manipulate and aggregate RAW data easily in CSVM space without any data modeling

²⁷ Directly taken from the *build.parsers.csvm.csvmutil* Python module (unitary test) of the CSVM toolkit.

²⁸ Python CSVM toolkit makes it possible using one line of code.

5. From collections to calculations: the case of time series

The case of enzyme kinetics is interesting because the CSVm format can be used in different ways: to store *i)* data (time series); *ii)* parameters for calculation (modeling, curve fitting); *iii)* results (best fit, residuals, parameter's values) and sometimes *iv)* the model itself. These steps can be found in a lot of other scientific fields (i.e. Physics, Chemistry) in which modeling is used as a core technique.

Enzyme kinetics measurements or calculations are typical of that we call short time series (less than 5000 rows, typical: 10 to 1000). This kind of data-island is widely found in a lot of biological, biochemical, chemical measurements, *courtesy from Sabine Gavalda*^{29 30} and Casimir Blonski.

This example, taken from real laboratory work, is typical for large sets of kinetics, a set of 800 files (RAW data) were collected and processed³¹ without using any database system.

5.1. A canonical format

The corresponding data is a molar (or equivalent measurements: i.e. values of optical densities taken with a spectrophotometer) concentration ([S] decreasing or [P] increasing) of a Substrate (P) or a Product (P) depending from time (the first column tagged 'Time (s)'). Each column at right of time, is an evolution of [P] for a particular molar concentration of an Inhibitor, giving a table of $[P] = f(t, [I])$ which can be modeled later to give parameters of an underlying kinetic model.

A typical example is given in the following figure with the first column devoted to time base and the others at right to different experiments (increase of Product corresponding to a particular value of inhibitor concentration).

Figure 16-A. – A partial view of kinetics CSVm data, a lot of rows are not shown (marked by '...' characters).

```

0.000 -0.0001 -0.0007 0.0015 -0.0002 0.0003 0.0010 -0.0013 -0.0017 0.0008 0.0009 -0.0017 0.0002 0.0010 -0.0004
0.010 0.0001 -0.0013 0.0004 -0.0004 -0.0011 -0.0000040 -0.0014 -0.0015 0.0001 0.0006 -0.0018 0.0005 0.0004 -0.0010
0.020 0.0001 -0.0009 0.0014 -0.0003 -0.0004 0.0005 -0.0013 -0.0014 -0.0004 0.0021 -0.0025 -0.0003 0.0008 -0.0003
...
6.970 0.4313 0.4479 - - - - - - - - - - - - -
6.980 0.4324 0.4470 - - - - - - - - - - - - -
6.990 0.4328 0.4484 - - - - - - - - - - - - -

#TITLE csvm\sab150403.csvm (CSVm) from [reports\sab150403.txt]
#HEADER Time (s) A1 A2 A3 A4 A5 A6 A7 Set-8 Set-9 Set-10 Set-11 Set-12 Set-13 Set-14
#TYPE NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC
#WIDTH 50 50 50 50 50 50 50 50 50 50 50 50 50 50 50
#META

```

This **primary** CSVm file can be generated *de novo* by users or acquisition system, but it can also result from another data format specific of a method or an apparatus. This was the case for this sample and the original format is shown in the following figure:

Figure 16-B. – Original format issued from a spectrophotometer..

```

X      Y      X      Y      X      Y      X      Y
-----
0.000  0.1112  0.050  0.1111  0.100  0.1111  0.150  0.1113
0.200  0.1116  0.250  0.0875  0.300  0.0514  0.350  0.0378
0.400  0.0696  0.450  0.1014  0.500  0.1126  0.550  0.1135
0.600  0.1145  0.650  0.1155  0.700  0.1164  0.750  0.1173
0.800  0.1181  0.850  0.1188  0.900  0.1194  0.950  0.1202
...

```

This example show one interest of the CSVm format: to be used as a canonical data format for RAW tabular data. This permits to use data independently of the data source and in a better way than pure CSV files that cannot embed metadata and annotations.

²⁹ S. Gavalda. L'Enolase de *Trypanosoma brucei* : Synthèse et étude du mode d'action d'inhibiteurs. Thèse de doctorat de l'Université Paul Sabatier Toulouse III (2005).

³⁰ S. Gavalda, F. Rodriguez, G. Beyries, C. Blonski. Software design and components for enzymology,. 21ème CBSO, Ax-Les-Thermes (2006).

³¹ S. Gavalda, F. Rodriguez, G. Beyries, C. Blonski. Méthodes et composants logiciels pour l'interface chimie-biologie. RECOB11 Aussois (2006).

In the CSV file shown in [Figure 16-A](#), the amount of information seems to be low: the values of inhibitor concentration (#HEADER field) are tagged as A1, A2, ... Set-14 which is not very explicit if the enzymologist wants to reuse the data later. In this case, the main information is the number of file: 150403 (date of experiment) stored in #TITLE field, this record number give access to another file with contains full information about experimental conditions.

This approach was used because we were not sure, at this time, if it was pertinent to store all parameters of the experiment with the data in the same CSV file. It was possible to use the #META field for that, in another way the insertion of annotations in the CSV table could be used³².

Notice also that some of the last rows of this file are not filled with data. This is normal, all kinetics are not recorded with same duration (but with same time origin and intervals). The CSV convention lets you to mark these empty positions by a particular character (in this case ` `) or not.

5.2. Data extraction from RAW values

Very often calculation programs cannot handle multi-column files (not only CSV) but uses a two-column file: time and a variation of a given quantity (here the product of reaction, for a particular value of inhibitor concentration). The following figure shows this kind of file with: a number of experimental points in the first row, followed by two columns using spaces as delimiters, one for time, the other for a biochemical quantity, and at last a block (rows beginning by a # character) for some metadata.

Figure 17. – A two-column data file specific of a given software.

```
691
0          0
0.05      0.67669
0.1       1.42857
0.15      2.18045
0.2       2.85714
...
34.383    281.50376
34.433    281.8797
34.483    282.25564

# LPZ file . lpz\0234-1.lpz
# Npts ..... 691
...
# Colx no .. 0
# Header x . Time
# Coly no .. 1
# Header y . B
...
```

In this case, we need to use software components to extract columns from the [primary](#) CSV file, make this kind of dataset: {Time (s); A1}, {Time (s); A2} ... {Time(s); Set-14} and save it using the corresponding format. With CSV it is easy to split the multi-column file (one time column, $i=1..n$ columns $[P]_i$) into n CSV files (each on with time and $[P]_i$). Column extractors are included in Python toolkit, a Python list is returned, depending on a search string applied on #HEADER values:

Code 4. – Simple CSV column extractor in Python language.

```
print """ Extract columns on the value of headers"
print "-> the column named 'fichier_mol' in strict mode (equal)"
ls = csv_ptr_getcol(c, 'fichier_mol', 1)
print "found %d column in CSV stream" % (ls[0])
print ls[1]
print "-> the columns in which string 'no_' is found (include)"
ls = csv_ptr_getcol(c, 'no_', 0)
print "found %d columns in CSV stream" % (ls[0])
```

³² And easy to do: if the first character of a row is a # not followed by a reserved keyword (TITLE, META, HEADER, TYPE or WIDTH) the row is not read by parsers and is available to store annotations, this kind of annotation provides a high level of information (but not immediately available for processing) for the related table.

For large sets it is also possible to make queries based on all parts of CSVM (metadata and data) following the columns or rows. The following output shows on a simple string (green) matrix (the data block of a CSVM Python class) how to make a search or a combined search (AND, OR) on columns or rows:

Code 5. – Example of using a CSVM table for a file catalog (this output is a dump of a CSVM Python class).

```

*** Test EQ column queries samples
uses matrix : [['PDB', '3.24', 'AB4'], ['- ', '1.0', '46'], ['PDB', '1.01', '4']]
query_col_eq(matrix, '4', 2, 1) => strict : rows [2]
query_col_eq(matrix, '4', 2, 0) => inc : rows [0, 1, 2]
query_col_eqs(matrix, 'PDB -', 'and', ' ', 0) => strict/and : rows []
query_col_eqs(matrix, 'PDB -', 'or', ' ', 0) => strict/or : rows [0, 2, 1]
query_col_eqs(matrix, '4 46', 'and', ' ', 2) => strict/and : rows []
query_col_eqsv(matrix, ['4', '46'], 'or', 2) => strict/or : rows [2, 1]
query_col_eqsv(matrix, ['4', '46'], 'or', 2, 0) => inc/or : rows [0, 1, 2]

*** Test NOT.EQ column queries samples
uses matrix : [['PDB', '3.24', 'AB4'], ['- ', '1.0', '46'], ['PDB', '1.01', '4']]
query_col_not_eq(matrix, '4', 2) => strict : rows [0, 1]
query_col_not_eq(matrix, '4', 0, 0) => inc : rows [0, 1, 2]
query_col_not_eqs(matrix, '4 LIG', 'or', ' ', 0, 0) => inc : rows [0, 1, 2]
query_col_not_eqs(matrix, 'B 6', 'and', ' ', 2, 0) => inc : rows [2]

*** Test EQ row queries samples
get the first row of matrix : ['PDB', '3.24', 'AB4']
query_row_eq(row, '4') => strict : columns []
query_row_eq(row, '4', 0) => inc : columns [1, 2]
query_row_eqsv(row, ['4', 'PDB'], 'or', 0) => inc/or : columns [1, 2, 0]

*** Test NOT.EQ row queries samples
get the first row of matrix : ['PDB', '3.24', 'AB4']
query_row_not_eq(row, 'PDB') => strict : columns [1, 2]
query_row_not_eqs(row, 'B 4', 'and', ' ', 0) => inc/and : columns []
query_row_not_eqs(row, 'B 4', 'or', ' ', 0) => inc/or : columns [1, 0]
query_row_not_eqs(row, 'LIG -', 'or', ' ', 0) => inc/or : columns [0, 1, 2]

```

The results (red) of each tested functions (blue) are enclosed in brackets at right of lines. In strict mode the search string (or the list of search strings) must be found as a plain value of a cell, if not strict (in mode), the search string can be also a substring of a cell value.

With this minimal toolkit it is easy to extract columns or rows and to make files at a given format usable by software that cannot handle multi-column CSVM files. CSVM files and processors can simplify greatly the handling of complex dataflows.

5.3. Mixed time series and information

In this case, two column curves were extracted from the CSVM file (the **primary** data source) shown in [Figure 16-A](#). Each extracted two-column files is formerly a curve $[P]=f(t)$ and it can be used for calculation (curve fit). But what about for the results ? In the following sample, not only the results of minimization (data, model key, residuals) were stored in one CSVM file but also all parameters used for calculation (initial values, solution, confidence intervals ...) by adding new columns at right of data.

The following output is a typical result after a calculation. The four first columns code for time series: column 1 for time ('X0'), column 2 for product [P] concentration ('Y0'), column 3 for curve fit using model C23 ('Y0 CALC') and column 4 for residuals ('RES') between experimental (Y0) and calculated (Y0 CALC) data.

Figure 18. – Results of a curve fitting calculation enclosed in a CSVM file.

0	0	0	0	LFILE	A020304\LPZ\0234-17.LPZ
.05	.97744	.6794984	.2979416	PROG	Model C23
.1	1.8797	1.345653	.5340471	VER	v:(1.0)
.15	2.85714	1.998803	.8583375	NPTS	687
.2	3.83459	2.639278	1.195312	NPAR	3
.25	4.3609	3.267402	1.093498	NVAL	1
.3	4.66165	3.883489	.7781613	F	182.0163
.35	5.11278	4.487844	.6249361	ET	.5158544
.4	5.86466	5.080766	.7838941	PNUM	1
.45	6.61654	5.662545	.9539948	PNAME	k
.5	7.21805	6.233466	.9845843	PSOL	.5151757

```

.55 7.81955 6.793804 1.025746 PMIN .5145575
.6 8.42105 7.343827 1.077223 PMAX .5157939
.65 8.79699 7.883799 .9131913 PNUM 2
.7 9.24812 8.413976 .8341446 PNAME Vo
.75 9.62406 8.934606 .6894541 PSOL 8.571785
.8 10.07519 9.445931 .6292582 PMIN 8.558927
.85 10.67669 9.94819 .7285004 PMAX 8.584643
.9 11.2782 10.44161 .8365898 PNUM 3
.95 11.65414 10.92642 .7277212 PNAME Vs
1 11.8797 11.40284 .4768639 PSOL 3.095204
1.05 12.10526 11.87107 .2341881 PMIN 3.092419
1.1 12.85714 12.33134 .5258017 PMAX 3.09799
1.15 13.60902 12.78383 .8251867 CNUM 1
1.2 14.13534 13.22876 .9065809 CNAME Quality
1.25 14.58647 13.6663 .920166 CVAL 991
1.3 15.03759 14.09666 .9409323 END -
1.45 16.01504 15.34638 .6686592 - -
1.5 16.39098 15.74975 .6412268 - -
1.55 16.69173 16.1468 .5449276 - -
...
34.183 125.4135 126.4381 -1.024567 - -
34.233 125.5639 126.5929 -1.028954 - -
34.283 125.8647 126.7476 -.8829575 - -
#
#TITLE / Model C23
#HEADER X0 Y0 Y0 calc RES KEY VALUE
#TYPE NUMERIC NUMERIC NUMERIC NUMERIC TEXT NUMERIC
#WIDTH 50 50 50 50 50 50
#META CSVM Result / VA04A QB BUILD

```

The two last columns are used to store calculation parameters and some results as a *key-values* scheme. This information are known as columns 'KEY', 'VALUE' and of type TEXT and NUMERIC and include primary data file (LFILE), program used (PROG), number of experimental points (NPTS), calculation values such as standard deviation (ET) and solution (PSOL) of model's parameters (PNAME) with confidence intervals (PMIN, PMAX), or other post processing calculations (Quality).

The KEY-VALUE block is closed by a KEY='END', after this point the 'KEY' and 'VALUE' columns are filled with blank chars ('-' in this case).

5.4. Combine different layers of data and calculations

This kind of file is considered a *secondary* data source, and these files could be aggregated to do other calculations. But, in fact, the most frequent use of *secondary* data is to pick some results inside (typically a solution value for a given parameter to be optimized). In this case, the value (PSOL=0.5151757) of an apparent kinetic constant *k* (PNUM=1, PNAME=k) was extracted, and stored in a new CSVM file: the *tertiary* data source.

More precisely, this file was used to store the variation of *k* value (*secondary* data) with concentrations of [S] or [I] (substrate and inhibitor) stored in the *primary* data source (CSVM file shown in [Figure 16](#)). An example of *tertiary* data source is given in the [Figure 19](#) :

Figure 19. – Aggregated results in a CSVM file and used as a new data source in the next calculation layer. Some columns are not shown and are marked by '...' symbols.

```

A020304 0234-1.LPZ Model C23 691 3 1 186.8595 .5211507 1 k ...
A020304 0234-10.LPZ Model C23 686 3 1 224.6174 .5734709 1 k ...
A020304 0234-11.LPZ Model C23 691 3 1 1373.348 1.41285 1 k ...
A020304 0234-12.LPZ Model C23 690 3 1 152.6696 .4714091 1 k ...
A020304 0234-13.LPZ Model C23 684 3 1 509.5747 .8650284 1 k ...
...
TB310304 3134-8.LPZ Model C23 623 3 1 516.7548 .9129487 1 k ...
TB310304 3134-9.LPZ Model C23 649 3 1 682.8931 1.028159 1 k ...

#TITLE Results in [model_c23\][CSV]
#HEADER LFILE PROG NPTS NPAR NVAL F ET PNUM PNAME ...
#TYPE UNDEF UNDEF UNDEF UNDEF UNDEF UNDEF UNDEF UNDEF UNDEF ...
#WIDTH 50 50 50 50 50 50 50 50 50 ...

```

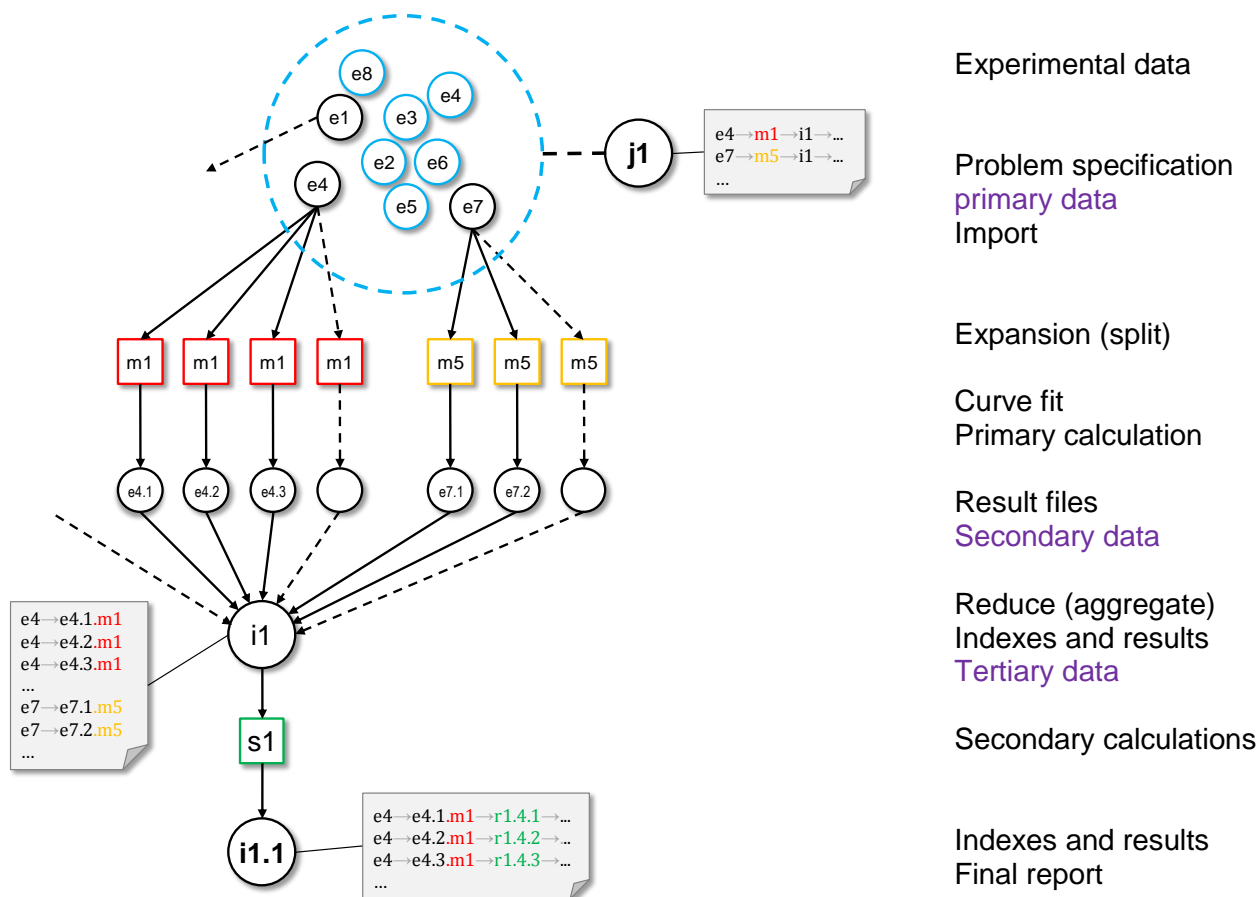
The tertiary data source was then used to make other calculations in order to compute an inhibition constant (K_I or K_I^*) significant of the affinity of a given inhibitor relative to this enzyme and that is used to rank the efficiency of different inhibitors.

The details of equations used in this software pipe are given in Annex-1 with the same color conventions than in this section.

5.5. Data driven parallelism

This previous sample shows how CSVM could be used to follow closely data parallelism, the [Figure 19](#) generalize this kind of concept:

Figure 20. – Data parallelism in the case of this Enzyme kinetics processing.



The files in set $j1$ are native files (apparatus or other formats). These files ($e4$, $e7$...) are converted to the canonical format in order to create a CSVM data space ($e1$, $e2$, $e3$... $e8$). For each file of this space the n data sets (two-column sets *time* vs. *quantity*) are extracted, and n suitable files for calculation (CSVM or not depending on the software used) are created (i.e. $m1$ or $m5$ squares).

After calculation the results are stored in n CSVM files mixing original data and results (i.e. $e4.1$, $e4.2$, $e4.3$... for columns $m1$, $m2$, $m3$... of file $e4$).

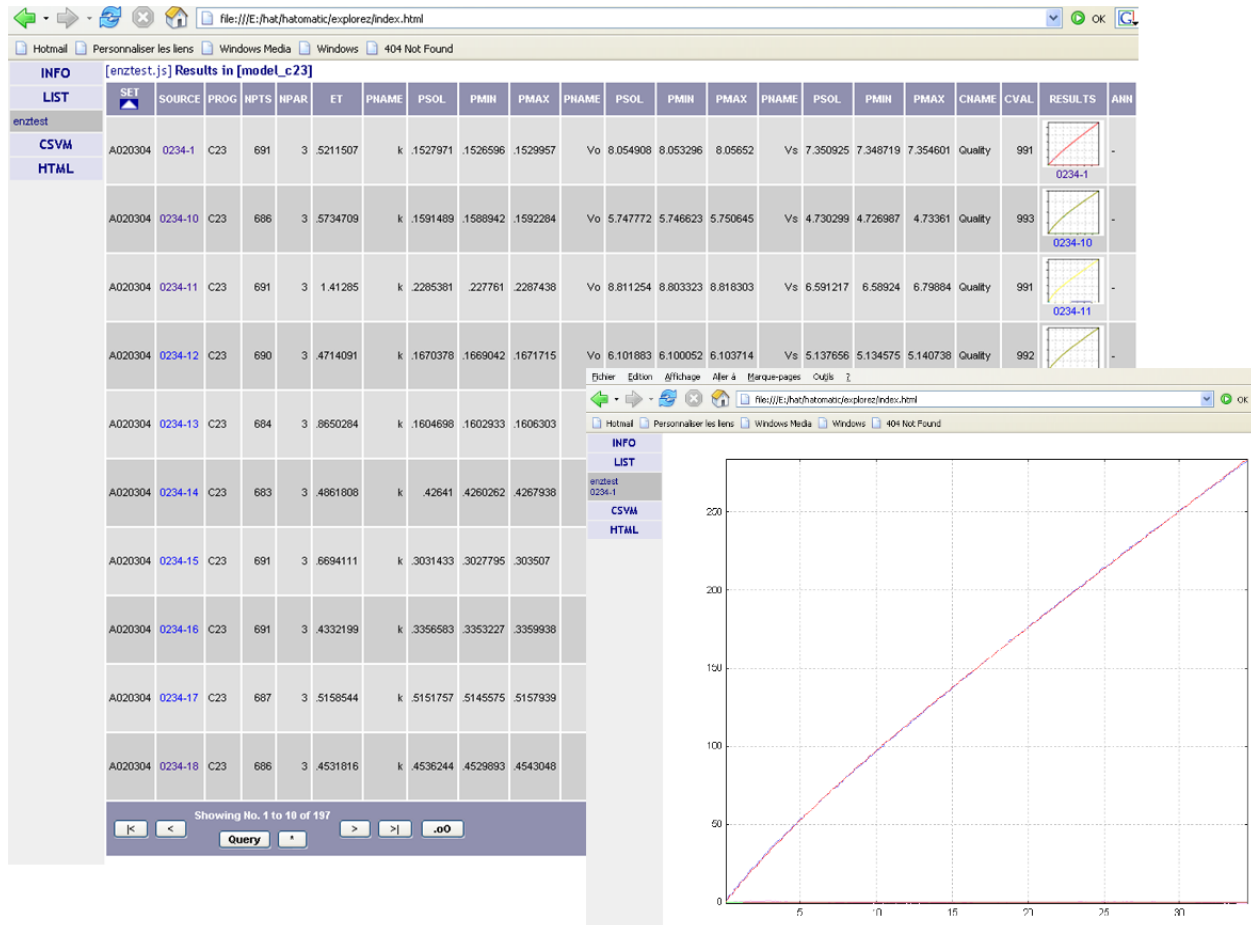
Some data inside these files are then extracted and aggregated inside a new data source $i1$. This source can be used as a report or for a new calculation run. In this case a new file can be created or columns can be added to $i1$ that gives a modified $i1.1$.

All calculations are independent and could be attributed to a pool of nodes/threads depending on the parallel model used. We can use CSVM as support for a data driven parallelism and it could be possible to implement an abstraction layer upon CSVM to make operations such as: split, scatter/bcast, reduce ... (typically MPI-like operations on data at different calculation steps) in order to implement a model of high throughput enzymology.

5.6. From calculations to reports

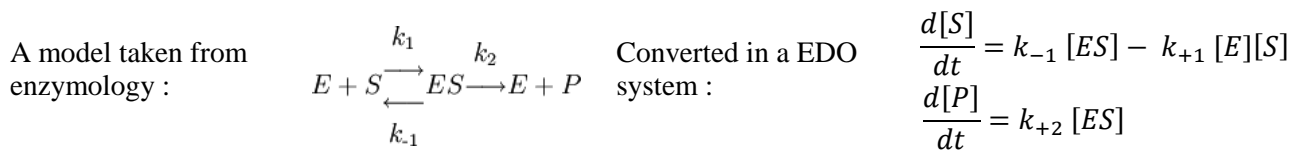
In the previous sections we have shown that CSVM is well suited to support publication chains. This also the case here, the final (i.e. **il.1**) or intermediate reports (**il**) can be used as input for a new dataflow (calculation or data exposition). The following figure shows a proof of concept for a report layer at the level **il** that includes a JavaScript framework and software components for extracting data (i.e. make graphs) from the CSVM files.

Figure 21. – Output of a publishing chain using aggregated data from calculations.



5.7. Including models in the scheme

Calculations uses a phenomenological model for data fitting, now what about the model itself ? The following sample shows how a model could be stored³³ in a CSV file. The chosen model is very used in enzyme kinetics: a variation of Michaelis-Menten mechanism that is converted as an EDO system and stored with all needed parameters for numerical integration inside a CSV file using a key/values scheme.



The corresponding CSV file shown in the following figure includes other equations (mass conservation, enzyme conservation), values for rate constants and molar concentrations, integration algorithm used (in this case RK4 stands for Runge-Kutta 4) and its parameters, other non-numerical data ...

Figure 22. – EDO system and parameters stored in a CSV file (human readable form).

```

ALGO  rk4      -      -      -      EDO solver
TIME  0        1      0.000001  0.01  Seconds
SPEC  S        1.0e-3  -      -      Substrat (M, publi= 10.10-6)
SPEC  P        0.0    -      -      Product (M)
SPEC  ES       0.0    -      -      Enzyme-substrat complex (M)
SPEC  E        1.0e-006 -      -      Free Enzyme (M, publi= 45.5 10-12 M)
RATE  k1       2.0e+005 -      -      ct. ES formation (M-1.s-1)
RATE  k-1      1.1e+003 -      -      ct. ES disparition (s-1)
RATE  k2       900    -      -      ct. P formation (s-1)
# Theoric KM of 10.10-3 M
PATH  S        <k -1>.ES - <k1>.E.S -      -      dS/dt
PATH  P        <k2>.ES -      -      dP/dt
PATH  ES       <k1>.E.S - <k -1>.ES - <k2>.ES -      -      dES/dt
PATH  E        - <k1>.E.S + <k -1>.ES + <k2>.ES -      -      dE/dt
MONI  Cm       S+P+ES+E -      -      Matter conservation
MONI  Etotal  E+ES   -      -      Enzyme conservation

#TITLE  Prob 3.3 of Shuler and Kargi S>>E
#HEADER KEY      UNDEF  UNDEF  UNDEF  UNDEF  UNDEF
#TYPE   TEXT     TEXT   TEXT   TEXT   TEXT   TEXT
#WIDTH  50      50     50     50     50     50

```

This is a proposal for coding the model in order to ensure genericity (a same formalism is usable for all models) and to keep human readable data (a formalism closely related to EDO system). Perhaps it is not the best solution for presenting data closely from the syntax and calling arguments of the numerical function used to make numerical integration. But in two cases, the CSV file will be easily parsed independently of model's complexity, and a one problem will be eliminated.

*Similar examples could be found in chemistry, environmental sciences, etc. In these cases, CSV shows that it is possible to store in the **same format**, the data and intermediate data, the results, the models. This is useful in a lot of experimental sciences because the development effort for computer scientists is reduced and they can focus on the problem itself.*

³³ Stored and parsed. We have used this approach for simulation, not for modeling because we use mainly analytical solutions rather than numerical solutions in our work.

6. Limit the number of data formats you use!

We have spoken about the use of CSV as a canonical format for RAW data conversion and intermediate results. Extending the concept to other kind of data is fairly immediate. In this section some samples taken from structural bioinformatics are given.

6.1. Map of molecular interactions

The previous section has shown that CSV can be used to store results of calculations. This ensures that these results could be used by another program or could be aggregated to larger dataset without any modifications. This is also true for non-time dependant data, here is a case for a map of molar interaction between a ligand (COA standing for Coenzyme A) and the water (HOH) or amino acid residues (LYS, ARG, ALA, SER, ...) of an enzyme (1KRU PDB structure), *courtesy from Vincent Mariaule*³⁴.

A software pipe generates a custom 2D map of interaction using different programs: LIGPLOT³⁵ package and related programs^{36 37}, the results of different calculations are filtered, extracted, converted and aggregated in the final CSV file.

Figure 23. – A 2D interaction map stored as a CSV table.

1KRU	1KRU	LYS	195	N	d	HOH	2053	0	a	2.83
1KRU	1KRU	COA	204	S1P	d	IPT	207	06	a	3.14
1KRU	1KRU	HOH	2053	0	d	COA	204	04A	a	2.92
1KRU	1KRU	COA	204	OAP	d	COA	204	N7A	a	2.52
1KRU	1KRU	ARG	183	NH2	d	COA	204	05A	a	3.09
1KRU	1KRU	ARG	183	NE	d	COA	204	05A	a	3.22
1KRU	1KRU	COA	204	N6A	d	ALA	160	0	a	2.95
1KRU	1KRU	ALA	160	N	d	COA	204	09P	a	3.16
1KRU	1KRU	SER	142	N	d	COA	204	05P	a	2.81
1KRU	1KRU	IPT	207	C6	-	COA	204	S1P	-	3.60
1KRU	1KRU	COA	204	C2A	-	VAL	177	CG1	-	3.84
1KRU	1KRU	COA	204	S1P	-	HIS	115	CD2	-	3.48
1KRU	1KRU	COA	204	C6A	-	ALA	175	CB	-	3.89
1KRU	1KRU	COA	204	C7P	-	TRP	139	CH2	-	3.76
1KRU	1KRU	COA	204	CEP	-	TRP	139	CH2	-	3.65
1KRU	1KRU	COA	204	CEP	-	TRP	139	CZ3	-	3.60
1KRU	1KRU	COA	204	C6P	-	TRP	139	CZ2	-	3.74
1KRU	1KRU	COA	204	C7P	-	TRP	139	CZ2	-	3.57
1KRU	1KRU	COA	204	C3P	-	ALA	105	CB	-	3.45

#TITLE	Ligplot summary!ligplot2csvm.py (v:1.04)!1KRU											
#HEADER	PDB	CHAIN	RESNAME	RESSEQ	ATYPE	FROM	RESNAME	RESSEQ	ATYPE	TO	DIST	
#TYPE	TEXT	TEXT	TEXT	INTEGER	TEXT	TEXT	TEXT	INTEGER	TEXT	TEXT	TEXT	FLOAT
#WIDTH	10	10	10	10	10	10	10	10	10	10	10	10

The *d* and *a* acronyms stands for donor and acceptor (hydrogen interactions), the columns RESSEQ are used to store the residue number in the protein sequence or in the crystallographic structure. The columns ATYPE store the atom types implied in interactions. The formalism is closely related to the Protein Data Bank³⁸ (PDB) formalism. The last column is the distance between the two atoms found by row.

Is structural data (a PDB³⁹ file) or sequence data could be stored in CSV files ? Definitely yes, but the interest could be limited because a lot of file formats exists and are widely used.

³⁴ V. Mariaule. Etude structurale des histones acétyl transférasés. Rapport M1 BBT, Université Paul Sabatier Toulouse III (2008).
³⁵ A.C. Wallace, R.A. Laskowski, J.M. Thornton. LIGPLOT: a program to generate schematic diagrams of protein-ligand interactions. Protein Eng. (1995) 8:2, 127-134.
³⁶ HBPLUS: Hydrogen Bond Calculation Program - I.K. McDonald, J.M. Thornton. Satisfying Hydrogen Bonding Potential in Proteins (1994) JMB 238, 777-793.
³⁷ NACCESS: Solvent accessible area calculations – S. Hubbard, J. Thornton (1992).
³⁸ RCSB Protein Data Bank - <http://www.rcsb.org/pdb/>
³⁹ In the case of a PDB file, the coordinates PDB block is immediately transposable, in the case of CSV the use of delimiters permits to add new columns in any order and to add useful annotations. For the header PDB block using CSV could simplify the metadata organization because CSV is not restricted to a particular number of columns or characters per row. Sometimes, for intermediate results in calculations or protein coordinate normalization we had used CSV for coordinate block.

6.2. BLOSSUM and PAM matrixes

For people working in genomics or proteomics, BLOSUM or PAM substitution matrixes⁴⁰ are well known objects. The next figure shows a CSVM version of a BLOSUM 62 matrix. For readability, some columns of the metadata block are missing (and marked by `...` characters at right).

The order of amino acid names (1-letter code in #HEADER) is the same in columns (left to right) and rows (top to bottom), data taken from Hiroto Saigo⁴¹.

Figure 24. – A Blossum 62 matrix encoded in a CSVM table.

```
4 - - - - - - - - - - - - - - - - - - - - - -
-1 5 - - - - - - - - - - - - - - - - - - - - -
-2 0 6 - - - - - - - - - - - - - - - - - - - -
-2 -2 1 6 - - - - - - - - - - - - - - - - - - -
0 -3 -3 -3 9 - - - - - - - - - - - - - - - - -
-1 1 0 0 -3 5 - - - - - - - - - - - - - - - -
-1 0 0 2 -4 2 5 - - - - - - - - - - - - - - -
0 -2 0 -1 -3 -2 -2 6 - - - - - - - - - - - -
-2 0 1 -1 -3 0 0 -2 8 - - - - - - - - - - -
-1 -3 -3 -3 -1 -3 -3 -4 -3 4 - - - - - - - -
-1 -2 -3 -4 -1 -2 -3 -4 -3 2 4 - - - - - -
-1 2 0 -1 -3 1 1 -2 -1 -3 -2 5 - - - - - -
-1 -1 -2 -3 -1 0 -2 -3 -2 1 2 -1 5 - - - - -
-2 -3 -3 -3 -2 -3 -3 -3 -1 0 0 -3 0 6 - - -
-1 -2 -2 -1 -3 -1 -1 -2 -2 -3 -1 -2 -4 7 - -
1 -1 1 0 -1 0 0 0 -1 -2 -2 0 -1 -2 -1 4 - -
0 -1 0 -1 -1 -1 -1 -2 -2 -1 -1 -1 -1 -2 -1 1 5 -
-3 -3 -4 -4 -2 -2 -3 -2 -2 -3 -2 -3 -1 1 -4 -3 -2 11 -
-2 -2 -2 -3 -2 -1 -2 -3 2 -1 -1 -2 -1 3 -3 -2 -2 2 7 -
0 -3 -3 -3 -1 -2 -2 -3 -3 3 1 -2 1 -1 -2 -2 0 -3 -1 4

#TITLE BLOSSUM62
#HEADER A R N D C Q E G H I ...
#TYPE NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC NUMERIC ...
#WIDTH 10 10 10 10 10 10 10 10 10 10 10 ...
#META from http://sunflower.kuicr.kyoto-u.ac.jp/~hiroto/project/optaa.html
```

Using same and unified file format for different parameters sets is interesting to minimize error and developments. Standard or reference data/charts are used in a lot of scientific fields, and it could be useful to take advantage from a canonical and generic format.

6.3. PDB hetero compound dictionary and variations

The following figure shows the bottom of the PDB's databank hetero dictionary⁴² encoded in CSVM. This file of about 2000 rows, is used to store PDB and chemical component identifier correspondences.

In example, the chemical component encoded with name ZZM (1-propan-2-yl-3-(2-pyridin-3-ylethynyl)pyrazolo[4,5-e]pyrimidin-4-amine) is found in PDB structure 2WXM (phosphoinositide-3-OH kinase). So, the first column is for the chemical compound PDB code (3 chars) and the column 2 is for a list (values separated by a space) of PDB entries (4 chars) in which the compound is found.

The CSVM format is not locked to tabular data of a given number of columns, formerly the following table is constituted of two columns (**HETNAME**, **INPDB**) but in the second column, the values could be single (last lines shown) or multiple (first line with 3 values: **2wtv**, **2wtw**, **2x81**).

It is possible to make 3 columns to store these values, but if the average amount of values by line is near 1, you will have a lot of unused columns needed by lines in which the number of values is high. This is precisely the case in this example; and it is better to define lines with two columns and to use a secondary delimiter in the **INPDB** column. Here a ` ` (space) character is used, but another could be used, the best in this case is a character⁴³ not found in current scientific data.

⁴⁰ Substitution matrix (Wikipedia) - http://en.wikipedia.org/wiki/Substitution_matrix

⁴¹ Akutsu Laboratory - <http://sunflower.kuicr.kyoto-u.ac.jp/index.html.en>

⁴² Ligand Expo Downloads - <http://ligand-expo.rcsb.org/ld-download.html>

⁴³ The '\$' could be used and the corresponding string with 3 values in INPDB column becomes '2wtv\$2wtw\$2x81'.

The CSV parser split the file according to the main delimiter (generally a TAB) and the application program, read CSV matrix cells and split them knowing the secondary (or ternary) delimiter, if multiple values are expectable inside.

The CSV formalism allows to send a signal through the parser about that (multiple values in cells). The #META row could be used with a specific keyword inside (the signal and the secondary delimiter), in example: '#META MULTIPLE_VALUES_IN_CELLS[\$] ...' or '#META CODE104 CODE108 ...' etc.

This signal can be interpreted (post-parsing) by the application layer. As usual we don't define in CSV specification this kind of mechanism; the interest of CSV is precisely to provide architecture for data and metadata but without any normalization.

Figure 25. – PDB and chemical component identifier (column HETNAME) correspondences encoded in a CSV file.

```
...
ZZL 2wtv 2wtw 2x81
ZZM 2wxm
ZZN 2wxg
...
ZZU 2wbp 2wbq
ZZY 2wd1
ZZZ 2cfi

#TITLE cc-to-pdb.tdd
#HEADER HETNAME INPDB
#TYPE TEXT TEXT
#WIDTH 10 10
#META http://ligand-expo.rcsb.org/ld-download.html|20jul2010
```

In this sample the #META field is used to store the download location, and a release date (using another delimiter the '|' character). The CSV format let the user free from defining characters used as primary or secondary delimiters, both for data or metadata block, the parser knows only column delimiter, others separators are used by application.

Real case: how to merge CSV tables

The previous file (Figure 25) is very close from original file (the dictionary released by the PDB): only a metadata block has been added. This is also the case for the next file (Figure 26): another dictionary taken from the PDB listing the same hetero compounds with the same 3 letter code (column HETNAME) and two other columns for a chemical formula written in SMILES format (SMI column) and a chemical name (MOLNAME):

Figure 26. – Last rows of a SMILES data file including PDB chemical component identifier (column HETNAME).

```
...
C(CNC(=N)N)C(C(C(=O)O)N)O ZZU (2s,3s)-hydroxyarginine
c1ccc(c(c1)[N+](=O)[O-])S(=O)(=O)n2ccc3c2cc(cn3)C(=O)N ZZY 1-[(2-nitrophenyl)sulfonyl]-1h-pyrrolo[3,2-
b]pyridine-6-carboxamide
C1C(NC2=C(N1)N=C(NC2=O)N)C=O ZZZ 6-formyltetrahydropterin

#TITLE Components-smiles-oe.smi
#HEADER SMI HETNAME MOLNAME
#TYPE TEXT TEXT TEXT
#WIDTH 10 10 10
#META http://ligand-expo.rcsb.org/ld-download.html|20jul2010
```

Even if the only modifications were to add a metadata block, this enables complex operations such as the merging of the files from Figure 25 and of Figure 26. This can be done with a few lines of code and functions from the Python CSV toolkit. This example is given in the following and it will be used to illustrate some subtleties related to the manipulation of RAW data sets.

After importation of some modules of the Python toolkit, the first step is to create a `csvm_ptr` Python class `c1` and to read the first CSVM file (Figure 25). The columns `HETNAME` and `INPDB` can exist (only one or more column than one) or not. The indices (column number beginning with zero) of 'HETNAME' and 'INPDB' are extracted from `c1` object, the number of rows of `c1` are also counted:

```
c1 = csvm_ptr()
c1 = csvm_ptr_read_extended_csvm(c1,"figure25.csvm","\t")
h1 = csvm_ptr_getcolind(c1,'HETNAME')
i1 = csvm_ptr_getcolind(c1,'INPDB')
r1 = c1.DATA_R
print "c1: column 'HETNAME' found %d times - first occurrence at index [%d]" % (len(h1), h1[0])
print "c1: column 'INPDB' found %d times - first occurrence at index [%d]" % (len(i1), i1[0])
print "c1: found %d rows" % (c1.DATA_R)
```

The same process is done for the second file (Figure 26). But before, a row `INPDB` is added in the second `csvm_ptr` class `c2`, at the right of the data block:

```
c2 = csvm_ptr()
c2 = csvm_ptr_read_extended_csvm(c2,"figure26.csvm","\t")
c2 = csvm_ptr_add_hash(c2,['INPDB'],['TEXT'],['10'],'')
h2 = csvm_ptr_getcolind(c2,'HETNAME')
i2 = csvm_ptr_getcolind(c2,'INPDB')
r2 = c2.DATA_R
print "c2: column 'HETNAME' found %d times - first occurrence at index [%d]" % (len(h2), h2[0])
print "c2: column 'INPDB' found %d times - first occurrence at index [%d]" % (len(i2), i2[0])
print "c2: found %d rows" % (r2)
```

The script output shows that the number of rows is not the same in `c1` (12107) and in (15071) `c2`:

```
c1: column 'HETNAME' found 1 times - first occurrence at index [0]
c1: column 'INPDB' found 1 times - first occurrence at index [1]
c1: found 12107 rows
c2: column 'HETNAME' found 1 times - first occurrence at index [1]
c2: column 'INPDB' found 1 times - first occurrence at index [3]
c2: found 15071 rows
```

This can be explained by the fact that a particular compound is known in PDB (Figure 26, `c2`) but not yet found as ligand or hetero compound inside a protein structure (Figure 25, `c1`).

In this case (`c1.DATA_R < c2.DATA_R` and `c2` is strictly a subset of `c1`) the choice is to extract values `INPDB` from `c1` and to fill column `INPDB` of `c2` when the same `HETNAME` value is found in the two data blocks.

For each value 'HETNAME' of `c2` a check is performed in `c1` (column `HETNAME`) in order to find if the same value exists. *If no*: the value is added to the `notfound` Python list, *if yes*: the counter `hetcount` is incremented. *If yes*: the corresponding row indice (`ind`) in `c1` column `HETNAME` is recorded.

The value of `ind` is used to get the `INPDB` value in the corresponding column of `c1`, and the value is then copied in `c2` at the current row indice (given by `i`) and in the corresponding column `INPDB` of `c2`:

```
"""
Extract the column 'HETNAME' from c1.
The column will be used to check if a 'HETNAME' value is found in c1
rather than using directly the data block.
"""

col = csvm_ptr_getcol(c1,'HETNAME')
hetcount = 0
notfound = []

"""
The loop: for each 'HETNAME' of c2
Check if 'HETNAME' value is found in column 'HETNAME' of c1.
If yes, get the indice of this value in c1 and add the 'INPDB' value
in the same row (of c1) to the current row and 'INPDB' column of c2
"""

for i in range(0, c2.DATA_R, 1):
    hetname2 = c2.DATA[i][h2[0]]
    if hetname2 in col[1]:
        ind = col[1].index(hetname2)
        # print "found %s at %d" % (hetname2, ind)
        c2.DATA[i][i2[0]] = c1.DATA[ind][i1[0]]
        hetcount += 1
    else:
        # print "not found %s" % (hetname2)
        notfound.append(hetname2)
```

```
print "c2: added %d hetnames" % (hetcount)
print "c1: %d hetnames not found" % (len(notfound))
```

The following script output shows that all **HETNAME** values of **c1** have been checked, and the summation of *added* and *not found* gives the good value of 15071 (number of rows for **c2** data block):

```
c2: added 12107 hetnames
c1: 2964 hetnames not found
```

Now it is possible to make an (optional) dump of the modified **c2**, to save a new CSV file and to clean data structures in memory:

```
#c2.csvm_ptr_dump(0,0)
s = csvm_ptr_make_csvm(c2, "\n", "\t")
file_str2file("result.csvm", s)
c1.csvm_ptr_clear()
c2.csvm_ptr_clear()
del(col[1])
```

The following table shows a part of the resulting CSV file. The files taken for the programming example are more recent than the two used for previous figures. If a comparison is done with the file of [Figure 25](#) we see obviously than 3 new compounds (**ZZV**, **ZZW**, **ZZX**) have been added:

Table 1. – Result of table merging (only the 6 last rows are shown).

SMI	HETNAME	MOLNAME	INPDB
C(CNC(=N)N)C(C(C(=O)O)N)O	ZZU	(2s,3s)-hydroxyarginine	2wbp 2wbq
CN(c1c2cccnc2c(c3c1CN(C3=O)Cc4cc(c(cc4)F)O)S(=O)(=O)C	ZZV	n-[7-(4-fluorobenzyl)-9-hydroxy-8-oxo-7,8-dihydro-6h-pyrrolo[3,4-g]quinolin-5-yl]-n-methylmethanesulfonamide	3oyd
c1ccc2c(c1)c3cc(ncc3n2Cc4ccc(cc4)F)C(=O)NO	ZZW	9-(4-fluorobenzyl)-n-hydroxy-9h-beta-carboline-3-carboxamide	3oyc
CCN1CC(n2c3c(c(c2C1=O)O)C(=O)N(N=C3C(=O)NC)Cc4ccc(c(c4)Cl)F)C	ZZX	(6s)-2-(3-chloro-4-fluorobenzyl)-8-ethyl-10-hydroxy-n,6-dimethyl-1,9-dioxo-1,2,6,7,8,9-hexahydropyrazino[1',2':1,5]pyrrolo[2,3-d]pyridazine-4-carboxamide	3oyb 3oyj 3oyl 3oyn
c1ccc(c(c1)[N+](=O)[O-])S(=O)(=O)n2ccc3c2cc(cn3)C(=O)N	ZZY	1-[(2-nitrophenyl)sulfonyl]-1h-pyrrolo[3,2-b]pyridine-6-carboxamide	2wd1
C1C(NC2=C(N1)N=C(NC2=O)N)C=O	ZZZ	6-formyltetrahydropterin	2cfi

The next sample ⁴⁴ is the output (using the **csvm_ptr_dump** method) of the first 5 rows of **c2** and it shows some compounds not found in **c1**. This is the case of compounds [000] and [005], the corresponding rows (indices 0 and 5) have the last field empty (no value between brackets).

Figure 27. – Dump of the CSV class corresponding to the Table.1 (only the 6 first rows of data block are shown).

```
DUMP: CSVm info {
SOURCE result.csvm
CSV CSVm
META [http://ligand-expo.rcsb.org/ld-download.html]
TITLE_N 1
TITLE Components-smiles-oe.smi with INPDB field added
HEADER_N 4
TYPE_N 4
WIDTH_N 4
0 10 TEXT {SMI}
1 10 TEXT {HETNAME}
2 10 TEXT {MOLNAME}
3 10 TEXT {INPDB}
DATA_R 15071
DATA_C 4
15071 4

0 [COC(=O)O][000][methyl hydrogen carbonate] []
1 [Coc1cc(cc(c1OC)OC)C(C(=O)N2CCCC2C(=O)OC(CCCc3ccccc3)CCCC4cccnc4)(F)F][001][1-[2,2-difluoro-2-(3,4,5-trimethoxy-phenyl)-acetyl]-piperidine-2-carboxylic acid 4-phenyl-1-(3-pyridin-3-yl-propyl)-butyl ester][1]4r ]
2 [CCC(C)C(C(=O)NC(CC(C)C)C(=O)O)NC(=O)C(Cc1ccccc1)CC(=O)NO][002][n-[(2R)-2-benzyl-4-(hydroxyamino)-4-oxobutanoyl]-l-isoleucyl-l-leucine][2fv9 ]
```

⁴⁴ The rows of the data block (grey and gray background, black or yellow foreground) are wrapped. The row indices (yellow) at left and INPDB values (yellow) at right, compound names (HETNAME, yellow) in the middle of the row.

```

3 [CC(C)CN1c2c(c(n(n2)Cc3cccc4c3cccc4)c5ccncc5)C(=O)N(C1=O)C] [003] [5-methyl-7-(2-methylpropyl)-2-
(naphthalen-1-ylmethyl)-3-pyridin-4-yl-2h-pyrazolo[3,4-d]pyrimidine-4,6(5h,7h)-dione] [2jtz ]
4 [c1ccc(cc1)C(C(=O)O)N] [004] [(2s)-amino(phenyl)ethanoic acid] [1sm1 1yit 1yiw 2z2p ]
5 [c1ccc(cc1)CC(C(C(=O)O)O)N] [005] [(2s,3s)-3-amino-2-hydroxy-4-phenylbutanoic acid] [ ]

```

This sample demonstrates that it is possible to make an operation that mixes the intersection and union (Section 4.3) of RAW tables in a simple manner.

Nevertheless this is a very favorable situation because all columns of **c1** have the same number of rows, no duplicates are found in **HETNAME** column of **c1** and **c2**, no blank values inserted between values ... if it is not the case, the algorithm must be modified.

Conclusion

This document is based on about ten years of CSVSM usage in laboratory and does not cover all samples we have encountered. The examples of using CSVSM are now numerous and all seem to demonstrate the plasticity of this format. This feature was essential to implement a SFMD (Single Format – Multiple Data) paradigm for tabular data.

This format was also found to operate very well with tabular like data (data that can be organized in a table, but are not inherently tabular) such as multiple key/values groups in the same data block.

The use of CSVSM as a canonical format is of a great interest in a lot of processes involving manipulation of tables: files conversion, normalization of data and data types, intermediate files in software pipes, automatic processing of big tables with sparse or heterogeneous data.

CSVSM is also a suitable format for gathering scientific objects and for exposition of data through a web framework, without any database system.

A generalized application could be the long term storage of RAW data. We have defined in this case the concept of ‘data museum’. The data is stored with a minimal annotation but enough to ensure its reuse some time (month, years and more) after. This kind annotation is nearly the same as the tags on drawers (that contains minerals or fossils samples) of a collection. It provides a view on data, enough to enable a research of a sample by a specialist of the domain. It is not suitable in this case to embed data in a database system, because these data islands are too heterogeneous to be taken in account by a given system (that is not guaranteed to be functional for a long duration). In the same way, to focus efforts on RAW data is very important because, very often, all RAW data is not integrated exhaustively in databases. Later, these ‘not used’ data could be requested for future studies, but at this time these data will be lost. This risk is particularly encountered in environmental sciences (loss of ‘zero time’).

The use of CSVSM in the short term could be also useful. Typically RAW data managers in laboratories cannot, in the same time, organize data collection; curate data and build a database. Perhaps in a lot of cases, a better approach could be to store CSVSM files, and at demand to build (a RDBMS or another system) a specific view of the data space (particularly in order to respond to a scientific question). In this case, the CSVSM files could give all information needed by IT teams to build the database, independently of the work of RAW data managers. A well curated flat database is often better (from a scientific point of view) than any RDBMS. The evolution of JavaScript frameworks and search engines shows also than a collection of CSVSM files could be exposed using a web interface with non negligible search capabilities. So, CSVSM could be useful to resolve the gap between people which wait a new database to store data, and people which wait data to model and build the same database (the use of AGILE techniques is not ever enough to solve this problem).

CSVSM could also give support in the case of not published scientific data, because the CSVSM format provides a way to store these tables (perhaps 80 per cent of data in a lot of experimental sciences) and allow a normalization effort, without too much complexity for the scientific community. Clearly, the research of funds in order to build databases for never (perhaps) used data could be a harder work than storing data in an adequate format.

Beyond of scientific use, this approach could be interesting in the Open Data. In this case, some questions are common with long term storage and unpublished data fields.

Specific References

CSVSM as a derivative of CSV is an Open file Format. The CSVSM-1 specification and use of CSVSM dictionaries are covered in the two following documents:

- G. Beyries, F. Rodriguez (2012) Technical Report: CSVSM format for scientific tabular data – <http://fr.arxiv.org/abs/1207.5711> [arXiv:1207.5711v1].
- F. Rodriguez (2012) Technical report: CSVSM dictionaries – <http://fr.arxiv.org/abs/1208.1934> [arXiv:1208.1934v1].

A CSVSM toolkit (Python programming language, Open Source: CeCILL free software license agreement) is available from the corresponding author of this manuscript.

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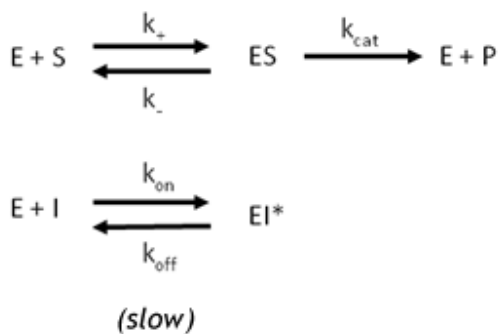
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Annex-1

In this scheme the use of CSV files makes easy the mapping of manual processing to automatic processing. The equations^{45 46} corresponding to these kinetic mechanisms (slow binding) are studied from the 60's and the enzymatic parameters can be resumed in the following way:

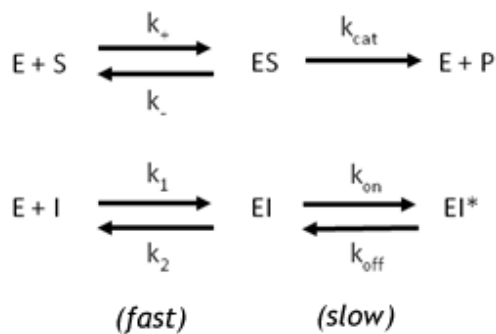
Mechanism A



$$K_M = \frac{[E] \cdot [S]}{[ES]} = \frac{k_- + k_{cat}}{k_+}$$

$$K_I^* = \frac{[E] \cdot [I]}{[EI^*]} = \frac{k_{off}}{k_{on}}$$

Mechanism B

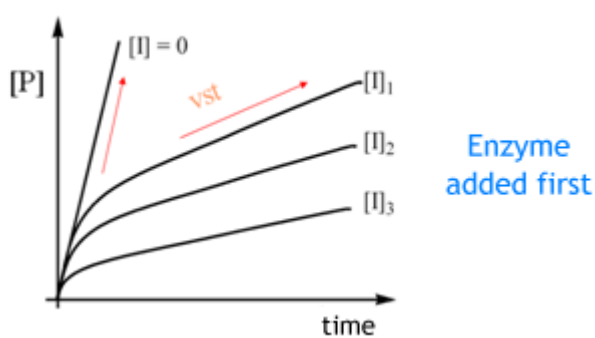


$$K_M = \frac{[E] \cdot [S]}{[ES]} = \frac{k_- + k_{cat}}{k_+}$$

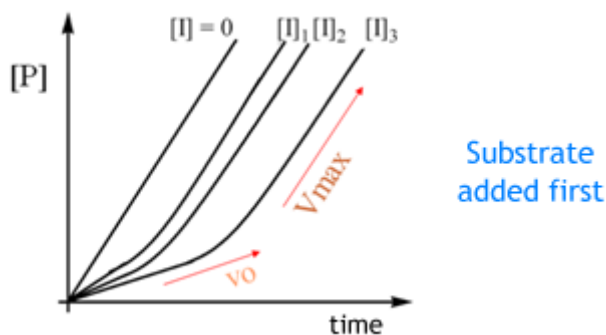
$$K_I^* = \frac{[E] \cdot [I]}{[EI]} = \frac{k_2}{k_1}$$

$$K_I^* = \frac{[E] \cdot [I]}{[EI] + [EI^*]} = \frac{K_I \cdot k_{off}}{k_{on} + k_{off}}$$

For a given mechanism (i.e. the **mechanism A**) the mixing order of the compounds (Substrate, Enzyme, Inhibitor) in the reaction cell induces different progress curves (and corresponding equation schemes) for the variation of $[P]$ vs. time t :



$$[P] = v_{st} \cdot t + \frac{(V_0 - v_{st})(1 - e^{-k \cdot t})}{k}$$



$$[P] = v_{max} \cdot t + \frac{(V_0 - V_{max})(1 - e^{-k_{off} \cdot t})}{k_{off}}$$

This is a particularity of slow binding mechanisms compared to more classic inhibition scheme found in enzyme kinetics science.

⁴⁵ Morrison JF. Kinetics of the reversible inhibition of enzyme-catalysed reactions by tight-binding inhibitors. *Biochim Biophys Acta*. (1969) 185(2): 269–286.

⁴⁶ Morrison JF, Walsh CT. The behavior and significance of slow-binding enzyme inhibitors. *Adv Enzymol Relat Areas Mol Biol*. (1988) 61: 201–301.

The analytic equations of $[P]=f(t)$ for each case of mixture show an apparent constant k that can be computed by a direct fit of the progress curves (two column curves extracted from **primary** data). After calculation, this apparent constant found in **secondary** CSV files and its value stored in key **PSOL** of the first key block (**PNUM=1**).

Knowing that the K_M constant (characteristic of the Enzyme-Substrate affinity) is determined using another way (kinetics without inhibitor in the mixture) it is possible to compute the variation of k_{on} , k_{off} and K_I^* using a linearization. Another curve fitting is done that uses the generated curves $k=f([I])$ stored in **tertiary** CSV files and computed as a *secondary* calculation layer:

Secondary calculation layer sample

$$1/v_{st} = f([I]) \text{ et } k=f([I])$$

$$[P] = v_{max} \cdot t + \frac{(v_0 - V_{max})(1 - e^{-k_{off} \cdot t})}{k_{off}}$$

$$k = k_{off} + k_{on} \frac{[I]}{1 + \frac{[S]}{K_M}}$$

Secondary calculation layer sample

$$V_{max}/v_0 = f([I]) \text{ ou } v_0/(V_{max} - v_0) = f([I])$$

$$\frac{V_{max}}{v_0} = 1 + \frac{[I]}{K_I^*}$$

$$\frac{v_0}{V_{max} - v_0} = \frac{K_I^*}{[I]}$$