# Proposed API for tech.ml.dataset

# GenerateMe

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#### Introduction

tech.ml.dataset is a great and fast library which brings columnar dataset to the Clojure. Chris Nuernberger has been working on this library for last year as a part of bigger tech.ml stack.

I've started to test the library and help to fix uncovered bugs. My main goal was to compare functionalities with the other standards from other platforms. I focused on R solutions: dplyr, tidyr and data.table.

During conversions of the examples I've come up how to reorganized existing tech.ml.dataset functions into simple to use API. The main goals were:

- Focus on dataset manipulation functionality, leaving other parts of tech.ml like pipelines, datatypes, readers, ML, etc.
- Single entry point for common operations one function dispatching on given arguments.
- group-by results with special kind of dataset a dataset containing subsets created after grouping as a column.
- Most operations recognize regular dataset and grouped dataset and process data accordingly.
- One function form to enable thread-first on dataset.

All proposed functions are grouped in tabs below. Select group to see examples and details.

If you want to know more about tech.ml.dataset and tech.ml.datatype please refer their documentation:

- Datatype
- Date/time
- Dataset

INFO: The future of this API is not known yet. Two directions are possible: integration into tech.ml or development under Scicloj organization. For the time being use this repo if you want to try. Join the discussion on Zulip

Let's require main namespace and define dataset used in most examples:

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	3	1.500	$\mathbf{C}$

:V1	:V2	:V3	:V4
2	4	0.5000	A
1	5	1.000	В
2	6	1.500	$\mathbf{C}$
1	7	0.5000	A
2	8	1.000	В
1	9	1.500	$\mathbf{C}$

# **Functionality**

#### Dataset

Dataset is a special type which can be considered as a map of columns implemented around tech.ml.datatype library. Each column can be considered as named sequence of typed data. Supported types include integers, floats, string, boolean, date/time, objects etc.

#### Dataset creation

Dataset can be created from various of types of Clojure structures and files:

- single values
- sequence of maps
- map of sequences or values
- sequence of columns (taken from other dataset or created manually)
- sequence of pairs
- file types: raw/gzipped csv/tsv, json, xls(x) taken from local file system or URL
- input stream

# api/dataset accepts:

- data
- options (see documentation of tech.ml.dataset/->dataset function for full list):
- :dataset-name name of the dataset
- :num-rows number of rows to read from file
- :header-row? indication if first row in file is a header
- :key-fn function applied to column names (eg. keyword, to convert column names to keywords)
- :separator column separator
- :single-value-column-name name of the column when single value is provided

Empty dataset.	
(api/dataset)	
-	
_unnamed [0 0]:	
1 1	
Dataset from single value.	
(api/dataset 999)	

\_unnamed [1 1]:

 $\frac{:\$ value}{999}$ 

Set column name for single value. Also set the dataset name.

 $\underline{\quad}$  unnamed [1 1]:

 $\frac{\text{my-single-value}}{999}$ 

Single value [1 1]:

 $\frac{0}{999}$ 

Sequence of pairs (first = column name, second = value(s)).

```
(api/dataset [[:A 33] [:B 5] [:C :a]])
```

\_unnamed [1 3]:

 $\frac{\text{:A :B :C}}{33 \quad 5 \quad \text{:a}}$ 

Not sequential values are repeated row-count number of times.

```
(api/dataset [[:A [1 2 3 4 5 6]] [:B "X"] [:C :a]])
```

 $\underline{\quad}$  unnamed [6 3]:

:A	:В	:(
1	Χ	:a
2	X	:a
3	X	:a
4	X	:a
5	X	:a
6	X	:a

Dataset created from map (keys = column name, second = value(s)). Works the same as sequence of pairs.

```
(api/dataset {:A 33})
(api/dataset {:A [1 2 3]})
(api/dataset {:A [3 4 5] :B "X"})
_unnamed [1 1]:
                                                         <u>:A</u>
                                                         33
_unnamed [3 1]:
                                                         <u>:A</u>
                                                         1
                                                         2 3
_unnamed [3 2]:
                                                      :A
                                                            :В
                                                      3
                                                            Χ
                                                      4
                                                            \mathbf{X}
                                                            \mathbf{X}
You can put any value inside a column
(api/dataset {:A [[3 4 5] [:a :b]] :B "X"})
\underline{\quad} unnamed [2 2]:
                                                     :A
                                                              :В
                                                              X
                                                     [3\ 4\ 5]
                                                     [:a :b]
                                                              \mathbf{X}
Sequence of maps
(api/dataset [{:a 1 :b 3} {:b 2 :a 99}])
(api/dataset [{:a 1 :b [1 2 3]} {:a 2 :b [3 4]}])
\underline{\phantom{a}}unnamed [2 2]:
                                                            :b
                                                            3
                                                       99
                                                           2
\underline{\phantom{a}}unnamed [2 2]:
```

Missing values are marked by nil

```
(api/dataset [{:a nil :b 1} {:a 3 :b 4} {:a 11}])
```

 $\underline{\phantom{a}}$ unnamed [3 2]:

3 4 11

Import CSV file

```
(api/dataset "data/family.csv")
```

data/family.csv [5 5]:

family	${\rm dob\_child1}$	${\rm dob\_child2}$	${\rm gender\_child1}$	${\rm gender\_child2}$
1	1998-11-26	2000-01-29	1	2
2	1996-06-22		2	
3	2002-07-11	2004-04-05	2	2
4	2004-10-10	2009-08-27	1	1
5	2000 - 12 - 05	2005 - 02 - 28	2	1

Import from URL

```
(defonce ds (api/dataset "https://vega.github.io/vega-lite/examples/data/seattle-weather.csv"))
```

ds

https://vega.github.io/vega-lite/examples/data/seattle-weather.csv [1461 6]:

date	precipitation	temp_max	temp_min	wind	weather
2012-01-01	0.000	12.80	5.000	4.700	drizzle
2012-01-02	10.90	10.60	2.800	4.500	rain
2012-01-03	0.8000	11.70	7.200	2.300	rain
2012-01-04	20.30	12.20	5.600	4.700	rain
2012-01-05	1.300	8.900	2.800	6.100	rain
2012-01-06	2.500	4.400	2.200	2.200	rain
2012-01-07	0.000	7.200	2.800	2.300	rain
2012-01-08	0.000	10.00	2.800	2.000	sun
2012-01-09	4.300	9.400	5.000	3.400	rain
2012-01-10	1.000	6.100	0.6000	3.400	rain
2012-01-11	0.000	6.100	-1.100	5.100	sun

date	precipitation	temp_max	temp_min	wind	weather
2012-01-12	0.000	6.100	-1.700	1.900	sun
2012-01-13	0.000	5.000	-2.800	1.300	sun
2012-01-14	4.100	4.400	0.6000	5.300	snow
2012-01-15	5.300	1.100	-3.300	3.200	snow
2012-01-16	2.500	1.700	-2.800	5.000	snow
2012-01-17	8.100	3.300	0.000	5.600	snow
2012-01-18	19.80	0.000	-2.800	5.000	snow
2012-01-19	15.20	-1.100	-2.800	1.600	snow
2012-01-20	13.50	7.200	-1.100	2.300	snow
2012-01-21	3.000	8.300	3.300	8.200	rain
2012-01-22	6.100	6.700	2.200	4.800	rain
2012-01-23	0.000	8.300	1.100	3.600	rain
2012-01-24	8.600	10.00	2.200	5.100	rain
2012-01-25	8.100	8.900	4.400	5.400	rain

# Saving

Export dataset to a file or output stream can be done by calling api/write-csv!. Function accepts:

- dataset
- file name with one of the extensions: .csv, .tsv, .csv.gz and .tsv.gz or output stream
- options:
- :separator string or separator char.

```
(api/write-csv! ds "output.tsv.gz")
(.exists (clojure.java.io/file "output.csv.gz"))
```

nil true

# Dataset related functions

Summary functions about the dataset like number of rows, columns and basic stats.

Number of rows		
(api/row-count ds)		
1461		
Number of columns		
(api/column-count ds)		
6		
Shape of the dataset, [row	count, column count]	
(api/shape ds)		

[1461 6]

General info about dataset. There are three variants:

- $\bullet\,$  default containing information about columns with basic statistics
- :basic just name, row and column count and information if dataset is a result of group-by operation
- :columns columns' metadata

(api/info ds)

(api/info ds :basic)
(api/info ds :columns)

https://vega.github.io/vega-lite/examples/data/seattle-weather.csv: descriptive-stats [6 10]:

:col- name	:datatype	:n- valid	:n- missing	:min	:mean	:mode :max	:standard- deviation	:skew
date	:packed-	1461	0	2012-	2013-	2015-		
	local-date			01-01	12-31	12-31		
precipita	tionfloat32	1461	0	0.000	3.029	55.90	6.680	3.506
temp m	ax:float32	1461	0	-1.600	16.44	35.60	7.350	0.2809
temp_m	in:float32	1461	0	-7.100	8.235	18.30	5.023	- 0.2495
weather	:string	1461	0			sun		0.2493
wind	:float $32$	1461	0	0.4000	3.241	9.500	1.438	0.8917

https://vega.github.io/vega-lite/examples/data/seattle-weather.csv:basic info [14]:

:name	:grouped?	:rows	:columns
https://vega.github.io/vega-lite/examples/data/seattle-weather.csv	false	1461	6

https://vega.github.io/vega-lite/examples/data/seattle-weather.csv :column info [6 4]:

:name	:size	:datatype	:categorical?
date precipitation	1461 1461	:packed-local-date :float32	
temp_max	1461	:float32	
temp_min wind	$1461 \\ 1461$	:float32 :float32	
weather	1461	:string	true

Getting a dataset name

(api/dataset-name ds)

"https://vega.github.io/vega-lite/examples/data/seattle-weather.csv"

Setting a dataset name (operation is immutable).

```
(->> "seattle-weather"
     (api/set-dataset-name ds)
     (api/dataset-name))
```

```
Columns and rows
Get columns and rows as sequences. column, columns and rows treat grouped dataset as regular one. See
Groups to read more about grouped datasets.
Select column.
(ds "wind")
(api/column ds "date")
#tech.ml.dataset.column<float32>[1461]
[4.700, 4.500, 2.300, 4.700, 6.100, 2.200, 2.300, 2.000, 3.400, 3.400, 5.100, 1.900, 1.300, 5.300, 3.20
#tech.ml.dataset.column<packed-local-date>[1461]
[2012-01-01,\ 2012-01-02,\ 2012-01-03,\ 2012-01-04,\ 2012-01-05,\ 2012-01-06,\ 2012-01-07,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 20
Columns as sequence
(take 2 (api/columns ds))
(#tech.ml.dataset.column<packed-local-date>[1461]
[2012-01-01,\ 2012-01-02,\ 2012-01-03,\ 2012-01-04,\ 2012-01-05,\ 2012-01-06,\ 2012-01-07,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 2012-01-08,\ 20
precipitation
[0.000, 10.90, 0.8000, 20.30, 1.300, 2.500, 0.000, 0.000, 4.300, 1.000, 0.000, 0.000, 0.000, 4.100, 5.3
Columns as map
(keys (api/columns ds :as-map))
("date" "precipitation" "temp_max" "temp_min" "wind" "weather")
Rows as sequence of sequences
(take 2 (api/rows ds))
([#object[java.time.LocalDate 0x7ba8c0d1 "2012-01-01"] 0.0 12.8 5.0 4.7 "drizzle"] [#object[java.time.L
Rows as sequence of maps
(clojure.pprint/pprint (take 2 (api/rows ds :as-maps)))
```

({"date" #object[java.time.LocalDate 0x56706259 "2012-01-01"], "precipitation" 0.0, "temp\_min" 5.0,

<sup>&</sup>quot;seattle-weather"

```
"weather" "drizzle",
"temp_max" 12.8,
"wind" 4.7}
{"date" #object[java.time.LocalDate 0x6b1bab71 "2012-01-02"],
"precipitation" 10.9,
"temp_min" 2.8,
"weather" "rain",
"temp_max" 10.6,
"wind" 4.5})
```

# **Printing**

Dataset is printed using dataset->str or print-dataset functions. Options are the same as in tech.ml.dataset/dataset-data->str. Most important is :print-line-policy which can be one of the: :single, :repl or :markdown.

unnamed [2 3]:

```
| :name | :group-id |
                                      :data |
|-----|
             0 | Group: 1 [5 4]:
    1 |
              | \| :V1 \| :V2 \| :V3 \| :V4 \| |
      | \|----\|----\| |
               1 \1
                    1 \|
                         1 \| 0.5000 \| A \| |
               1 \1
                    1 \|
                         3 \| 1.500 \|
                                       C \ | |
               | \| 1 \| 5 \| 1.000 \| B \| |
      1
               | \| 1 \| 7 \| 0.5000 \| A \| |
               | \| 1 \| 9 \| 1.500 \|
    2 |
             1 | Group: 2 [4 4]:
               | \| :V1 \| :V2 \| :V3 \| :V4 \| |
              | \|----\|----\||
               | \cdot |
                    2 \|
                          2 \| 1.000 \| B \| |
      1
               I \setminus I
                    2 \|
                         4 \| 0.5000 \|
                                       A \setminus I
               | \|
                    2 \|
                          6 \| 1.500 \|
                                       C /| |
               | \|
                    2 \|
                         8 \| 1.000 \|
```

### Group-by

Grouping by is an operation which splits dataset into subdatasets and pack it into new special type of... dataset. I distinguish two types of dataset: regular dataset and grouped dataset. The latter is the result of grouping.

Grouped dataset is annotated in by :grouped? meta tag and consist following columns:

- :name group name or structure
- :group-id integer assigned to the group
- :data groups as datasets

Almost all functions recognize type of the dataset (grouped or not) and operate accordingly.

You can't apply reshaping or join/concat functions on grouped datasets.

# Grouping

Grouping is done by calling group-by function with arguments:

- ds dataset
- grouping-selector what to use for grouping
- options:
  - :result-type what to return:
    - \* :as-dataset (default) return grouped dataset
    - \* :as-indexes return rows ids (row number from original dataset)
    - \* :as-map return map with group names as keys and subdataset as values
    - \* :as-seq return sequens of subdatasets
  - :limit-columns list of the columns which should be returned during grouping by function.

All subdatasets (groups) have set name as the group name, additionally group-id is in meta.

Grouping can be done by:

- single column name
- seq of column names
- map of keys (group names) and row indexes
- value returned by function taking row as map

Note: currently dataset inside dataset is printed recursively so it renders poorly from markdown. So I will use :as-seq result type to show just group names and groups.

```
____
```

List of columns in groupd dataset

```
(api/column-names (api/group-by DS :V1))

(:name :group-id :data)
```

Content of the grouped dataset

```
(api/columns (api/group-by DS :V1) :as-map)
```

```
{:name #tech.ml.dataset.column<int64>[2]
:name
[1, 2, ], :group-id #tech.ml.dataset.column<int64>[2]
:group-id
[0, 1, ], :data #tech.ml.dataset.column<object>[2]
:data
[Group: 1 [5 4]:
| :V1 | :V2 | :V3 | :V4 |
|----|----|----|
| 1 | 1 | 0.5000 | A |
```

```
3 | 1.500 |
        5 | 1.000 |
                      ΒΙ
        7 | 0.5000 |
   1 |
        9 | 1.500 |
                      CI
, Group: 2 [4 4]:
| :V1 | :V2 |
               : V3 | : V4 |
|----|
   2 |
        2 | 1.000 |
                      ΒΙ
   2 |
        4 | 0.5000 |
   2 |
        6 | 1.500 |
                      Cl
   2 |
        8 |
             1.000 |
, ]}
```

Grouped dataset as map

```
(keys (api/group-by DS :V1 {:result-type :as-map}))
```

(1 2)

```
(vals (api/group-by DS :V1 {:result-type :as-map}))
```

(Group: 1 [5 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	3	1.500	$\mathbf{C}$
1	5	1.000	В
1	7	0.5000	A
1	9	1.500	$\mathbf{C}$

Group: 2 [4 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
2	4	0.5000	A
2	6	1.500	$\mathbf{C}$
2	8	1.000	В

)

Group dataset as map of indexes (row ids)

```
(api/group-by DS :V1 {:result-type :as-indexes})
```

```
{1 [0 2 4 6 8], 2 [1 3 5 7]}
```

Grouped datasets are printed as follows by default.

```
(api/group-by DS :V1)
```

\_unnamed [2 3]:

:name	:group-id	:data
1	0	Group: 1 [5 4]:
2	1	Group: 2 [4 4]:

To get groups as sequence or a map can be done from grouped dataset using <code>groups->seq</code> and <code>groups->map</code> functions.

Groups as seq can be obtained by just accessing :data column.

I will use temporary dataset here.

(Group: 1 [2 2]:

a b
1 a
1 b

Group: 2 [2 2]:

 $\begin{array}{ccc} a & b \\ \hline 2 & c \\ 2 & d \end{array}$ 

```
(-> {"a" [1 1 2 2]

"b" ["a" "b" "c" "d"]}

(api/dataset)

(api/group-by "a")

(api/groups->seq))
```

(Group: 1 [2 2]:

a b1 a1 b

Group: 2 [2 2]:

a b
2 c
2 d

)

Groups as map

Grouping by more than one column. You can see that group names are maps. When ungrouping is done these maps are used to restore column names.

```
(api/group-by DS [:V1 :V3] {:result-type :as-seq})
```

(Group: {:V3 1.0, :V1 1} [1 4]:

:V1	:V2	:V3	:V4
1	5	1.000	В

Group: {:V3 0.5, :V1 1} [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	7	0.5000	A

Group: {:V3 0.5, :V1 2} [1 4]:

:V1	:V2	:V3	:V4
2	4	0.5000	A

Group: {:V3 1.0, :V1 2} [2 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
2	8	1.000	В

Group: {:V3 1.5, :V1 1} [2 4]:

:V1	:V2	:V3	:V4
1	3	1.500	С
1	9	1.500	$\mathbf{C}$

Group: {:V3 1.5, :V1 2} [1 4]:

:V1	:V2	:V3	:V4
2	6	1.500	С

)

Grouping can be done by providing just row indexes. This way you can assign the same row to more than one group.

(Group: group-a [4 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
2	2	1.000	В
1	3	1.500	$\mathbf{C}$

Group: group-b [4 4]:

:V1	:V2	:V3	:V4
2	6	1.500	С
2	6	1.500	$^{\mathrm{C}}$
2	6	1.500	$\mathbf{C}$
2	2	1.000	В

)

You can group by a result of gruping function which gets row as map and should return group name. When map is used as a group name, ungrouping restore original column names.

(Group: 1.0 [2 4]:

:V1	:V2	:V3	:V4
2	4	0.5000	A
1	5	1.000	В

Group: 2.0 [2 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
2	8	1.000	В

Group: 0.5 [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	7	0.5000	A

Group: 3.0 [1 4]:

:V1	:V2	:V3	:V4
2	6	1.500	$^{\rm C}$

Group: 1.5 [2 4]:

:V1	:V2	:V3	:V4
1	3	1.500	С
1	9	1.500	$\mathbf{C}$

)

You can use any predicate on column to split dataset into two groups.

```
(api/group-by DS (comp #(< % 1.0) :V3) {:result-type :as-seq})</pre>
```

(Group: false [6 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
1	5	1.000	В
2	6	1.500	$\mathbf{C}$
2	8	1.000	В
1	9	1.500	$\mathbf{C}$

Group: true [3 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	4	0.5000	A
1	7	0.5000	A

)

juxt is also helpful

(api/group-by DS (juxt :V1 :V3) {:result-type :as-seq})

(Group: [1 1.0] [1 4]:

:V1	:V2	:V3	:V4
1	5	1.000	В

Group: [1 0.5] [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	7	0.5000	A

Group: [2 1.5] [1 4]:

:V1	:V2	:V3	:V4
2	6	1.500	С

Group: [1 1.5] [2 4]:

:V1	:V2	:V3	:V4
1	3	1.500	С
1	9	1.500	$\mathbf{C}$

Group: [2 0.5] [1 4]:

:V1	:V2	:V3	:V4
2	4	0.5000	A

Group: [2 1.0] [2 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
2	8	1.000	В

)

tech.ml.dataset provides an option to limit columns which are passed to grouping functions. It's done for performance purposes.

(Group: {:V1 1} [5 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	3	1.500	$\mathbf{C}$
1	5	1.000	В
1	7	0.5000	A
1	9	1.500	С

Group: {:V1 2} [4 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
2	4	0.5000	A
2	6	1.500	$\mathbf{C}$
2	8	1.000	В

)

# Ungrouping

Ungrouping simply concats all the groups into the dataset. Following options are possible

- :order? order groups according to the group name ascending order. Default: false
- :add-group-as-column should group name become a column? If yes column is created with provided name (or :\$group-name if argument is true). Default: nil.
- :add-group-id-as-column should group id become a column? If yes column is created with provided name (or :\$group-id if argument is true). Default: nil.
- :dataset-name to name resulting dataset. Default: nil (\_unnamed)

If group name is a map, it will be splitted into separate columns. Be sure that groups (subdatasets) doesn't contain the same columns already.

If group name is a vector, it will be splitted into separate columns. If you want to name them, set vector of target column names as :add-group-as-column argument.

After ungrouping, order of the rows is kept within the groups but groups are ordered according to the internal storage.

Grouping and ungrouping.

```
(-> DS
    (api/group-by :V3)
    (api/ungroup))
```

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
1	5	1.000	В
2	8	1.000	В
1	1	0.5000	A
2	4	0.5000	A
1	7	0.5000	A
1	3	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$
1	9	1.500	$\mathbf{C}$

Groups sorted by group name and named.

Ordered by V3 [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	4	0.5000	A
1	7	0.5000	A
2	2	1.000	В
1	5	1.000	В
2	8	1.000	В
1	3	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$
1	9	1.500	С

Let's add group name and id as additional columns

 $\underline{\quad}$  unnamed [9 6]:

:\$group-name	:\$group-id	:V1	:V2	:V3	:V4
false	0	2	4	0.5000	A
false	0	1	5	1.000	В
false	0	2	6	1.500	$^{\mathrm{C}}$
false	0	1	7	0.5000	A
false	0	2	8	1.000	В
false	0	1	9	1.500	$\mathbf{C}$
true	1	1	1	0.5000	A
true	1	2	2	1.000	В
true	1	1	3	1.500	$\mathbf{C}$

Let's assign different column names

\_unnamed [9 6]:

Is V2 less than 4?	group id	:V1	:V2	:V3	:V4
false	0	2	4	0.5000	A
false	0	1	5	1.000	В
false	0	2	6	1.500	$\mathbf{C}$
false	0	1	7	0.5000	A
false	0	2	8	1.000	В
false	0	1	9	1.500	$\mathbf{C}$
true	1	1	1	0.5000	A
true	1	2	2	1.000	В
true	1	1	3	1.500	$\mathbf{C}$

If we group by map, we can automatically create new columns out of group names.

\_unnamed [9 6]:

V1 and V3 multiplied	V4 as lowercase	:V1	:V2	:V3	:V4
1.000	a	2	4	0.5000	A

V1 and V3 multiplied	V4 as lowercase	:V1	:V2	:V3	:V4
0.5000	a	1	1	0.5000	A
0.5000	a	1	7	0.5000	A
1.000	b	1	5	1.000	В
2.000	b	2	2	1.000	В
2.000	b	2	8	1.000	В
3.000	c	2	6	1.500	$\mathbf{C}$
1.500	c	1	3	1.500	$\mathbf{C}$
1.500	$\mathbf{c}$	1	9	1.500	$\mathbf{C}$

We can add group names without separation

# $\underline{\phantom{a}}$ unnamed [9 5]:

just map	:V1	:V2	:V3	:V4
{"V1 and V3 multiplied" 1.0, "V4 as lowercase" "a"}	2	4	0.5000	A
{"V1 and V3 multiplied" 0.5, "V4 as lowercase" "a"}	1	1	0.5000	A
{"V1 and V3 multiplied" 0.5, "V4 as lowercase" "a"}	1	7	0.5000	A
{"V1 and V3 multiplied" 1.0, "V4 as lowercase" "b"}	1	5	1.000	В
{"V1 and V3 multiplied" 2.0, "V4 as lowercase" "b"}	2	2	1.000	В
{"V1 and V3 multiplied" 2.0, "V4 as lowercase" "b"}	2	8	1.000	В
{"V1 and V3 multiplied" 3.0, "V4 as lowercase" "c"}	2	6	1.500	$\mathbf{C}$
{"V1 and V3 multiplied" 1.5, "V4 as lowercase" "c"}	1	3	1.500	$\mathbf{C}$
{"V1 and V3 multiplied" 1.5, "V4 as lowercase" "c"}	1	9	1.500	С

The same applies to group names as sequences

```
(-> DS
    (api/group-by (juxt :V1 :V3))
    (api/ungroup {:add-group-as-column "abc"}))
```

# \_unnamed [9 6]:

:abc-0	:abc-1	:V1	:V2	:V3	:V4
1	1.000	1	5	1.000	В
1	0.5000	1	1	0.5000	A
1	0.5000	1	7	0.5000	A
2	1.500	2	6	1.500	$\mathbf{C}$
1	1.500	1	3	1.500	$\mathbf{C}$
1	1.500	1	9	1.500	$\mathbf{C}$
2	0.5000	2	4	0.5000	A
2	1.000	2	2	1.000	В
2	1.000	2	8	1.000	В

Let's provide column names

```
(-> DS
    (api/group-by (juxt :V1 :V3))
    (api/ungroup {:add-group-as-column ["v1" "v3"]}))
```

\_unnamed [9 6]:

v1	v3	:V1	:V2	:V3	:V4
1	1.000	1	5	1.000	В
1	0.5000	1	1	0.5000	A
1	0.5000	1	7	0.5000	A
2	1.500	2	6	1.500	$\mathbf{C}$
1	1.500	1	3	1.500	$\mathbf{C}$
1	1.500	1	9	1.500	$\mathbf{C}$
2	0.5000	2	4	0.5000	A
2	1.000	2	2	1.000	В
2	1.000	2	8	1.000	В

Also we can supress separation

 $\underline{\text{unnamed } [9\ 5]}$ :

:\$group-name	:V1	:V2	:V3	:V4
[1 1.0]	1	5	1.000	В
$[1 \ 0.5]$	1	1	0.5000	A
$[1 \ 0.5]$	1	7	0.5000	A
$[2\ 1.5]$	2	6	1.500	$\mathbf{C}$
$[1 \ 1.5]$	1	3	1.500	$\mathbf{C}$
$[1 \ 1.5]$	1	9	1.500	$\mathbf{C}$
$[2\ 0.5]$	2	4	0.5000	A
$[2 \ 1.0]$	2	2	1.000	В
$[2 \ 1.0]$	2	8	1.000	В

# Other functions

To check if dataset is grouped or not just use grouped? function.

```
(api/grouped? DS)
```

nil

```
(api/grouped? (api/group-by DS :V1))
```

true

If you want to remove grouping annotation (to make all the functions work as with regular dataset) you can use unmark-group or as-regular-dataset (alias) functions.

It can be important when you want to remove some groups (rows) from grouped dataset using drop-rows or something like that.

```
(-> DS
    (api/group-by :V1)
    (api/as-regular-dataset)
    (api/grouped?))
```

nil

This is considered internal.

If you want to implement your own mapping function on grouped dataset you can call process-group-data and pass function operating on datasets. Result should be a dataset to have ungrouping working.

```
(-> DS
    (api/group-by :V1)
    (api/process-group-data #(str "Shape: " (vector (api/row-count %) (api/column-count %))))
    (api/as-regular-dataset))
```

 $\underline{\quad}$  unnamed [2 3]:

:name	:group-id	:data
1	0	Shape: [5 4]
2	1	Shape: [4 4]

#### Columns

Column is a special tech.ml.dataset structure based on tech.ml.datatype library. For our purposes we cat treat columns as typed and named sequence bound to particular dataset.

Type of the data is inferred from a sequence during column creation.

#### Names

To select dataset columns or column names columns-selector is used. columns-selector can be one of the following:

- :all keyword selects all columns
- column name for single column
- sequence of column names for collection of columns
- regex to apply pattern on column names or datatype
- $\bullet\,$  filter predicate to filter column names or data type

Column name can be anything.

column-names function returns names according to columns-selector and optional meta-filed. meta-field is one of the following:

- :name (default) to operate on column names
- :datatype to operated on column types

• :all - if you want to process all metadata

```
To select all column names you can use column-names function.
(api/column-names DS)
(:V1 :V2 :V3 :V4)
(api/column-names DS :all)
(:V1 :V2 :V3 :V4)
In case you want to select column which has name :all (or is sequence or map), put it into a vector. Below
code returns empty sequence since there is no such column in the dataset.
(api/column-names DS [:all])
()
Obviously selecting single name returns it's name if available
(api/column-names DS : V1)
(api/column-names DS "no such column")
(:V1)
()
Select sequence of column names.
(api/column-names DS [:V1 "V2" :V3 :V4 :V5])
(:V1 :V3 :V4)
Select names based on regex, columns ends with 1 or 4
(api/column-names DS #".*[14]")
(:V1:V4)
Select names based on regex operating on type of the column (to check what are the column types, call
(api/info DS :columns). Here we want to get integer columns only.
(api/column-names DS #"^:int.*" :datatype)
(:V1:V2)
And finally we can use predicate to select names. Let's select double precision columns.
(api/column-names DS #(= :float64 %) :datatype)
(:V3)
```

If you want to select all columns but given, use complement function. Works only on a predicate.

```
(api/column-names DS (complement #{:V1}))
(api/column-names DS (complement #(= :float64 %)) :datatype)

(:V2 :V3 :V4)
(:V1 :V2 :V4)
```

You can select column names based on all column metadata at once by using :all metadata selector. Below we want to select column names ending with 1 which have long datatype.

#### Select

select-columns creates dataset with columns selected by columns-selector as described above. Function works on regular and grouped dataset.

Select only float64 columns

```
(api/select-columns DS #(= :float64 %) :datatype)
```

\_unnamed [9 1]:

:V3 0.5000 1.000 1.500 0.5000 1.000 1.500 0.5000 1.000 1.500

Select all but :V1 columns

```
(api/select-columns DS (complement #{:V1}))
```

 $\underline{\quad}$  unnamed [9 3]:

:V2	:V3	:V4
1	0.5000	A
2	1.000	В
3	1.500	$\mathbf{C}$
4	0.5000	A
5	1.000	В
6	1.500	$\mathbf{C}$

:V2	:V3	:V4
7	0.5000	A
8	1.000	В
9	1.500	$\mathbf{C}$

If we have grouped data set, column selection is applied to every group separately.

```
(-> DS
    (api/group-by :V1)
    (api/select-columns [:V2 :V3])
    (api/groups->map))
```

{1 Group: 1 [5 2]:

:V2	:V3
1	0.5000
3	1.500
5	1.000
7	0.5000
9	1.500

, 2 Group: 2 [4 2]:

:V2	:V3
2	1.000
4	0.5000
6	1.500
8	1.000

}

# Drop

drop-columns creates dataset with removed columns.

Drop float64 columns

```
(api/drop-columns DS #(= :float64 %) :datatype)
```

\_unnamed [9 3]:

:V1	:V2	:V4
1	1	A
2	2	В
1	3	$\mathbf{C}$
2	4	A
1	5	В
2	6	$\mathbf{C}$

:V1	:V2	:V4
1	7	A
2	8	В
1	9	С

Drop all columns but  $:\! V1$  and  $:\! V2$ 

```
(api/drop-columns DS (complement #{:V1 :V2}))
```

\_unnamed [9 2]:

:V1	:V2
1	1
2	2
1	3
2	4
1	5
2	6
1	7
2	8
1	9

If we have grouped data set, column selection is applied to every group separately. Selected columns are dropped.

```
(-> DS
    (api/group-by :V1)
    (api/drop-columns [:V2 :V3])
    (api/groups->map))
```

{1 Group: 1 [5 2]:

:V1	:V4
1	A
1	$\mathbf{C}$
1	В
1	A
1	$\mathbf{C}$

, 2 Group: 2 [4 2]:

4

}

# Rename

If you want to rename colums use rename-columns and pass map where keys are old names, values new ones.

\_unnamed [9 4]:

v1	v2	$[1 \ 2 \ 3]$	java.lang. Object@3467017b
1	1	0.5000	A
2	2	1.000	В
1	3	1.500	С
2	4	0.5000	A
1	5	1.000	В
2	6	1.500	С
1	7	0.5000	A
2	8	1.000	В
1	9	1.500	C

Function works on grouped dataset

{1 Group: 1 [5 4]:

v1	v2	$[1 \ 2 \ 3]$	java.lang.Object@410e4101
1	1	0.5000	A
1	3	1.500	C
1	5	1.000	В
1	7	0.5000	A
1	9	1.500	C

, 2 Group: 2 [4 4]:

v1	v2	$[1 \ 2 \ 3]$	java.lang. Object@410e4101
2	2	1.000	В
2	4	0.5000	A
2	6	1.500	С

v1	v2	[1 2 3]	java.lang.Object@410e4101
2	8	1.000	В

}

# Add or update

To add (or update existing) column call add-or-update-column function. Function accepts:

- ds a dataset
- column-name if it's existing column name, column will be replaced
- column can be column (from other dataset), sequence, single value or function. Too big columns are always trimmed. Too small are cycled or extended with missing values (according to size-strategy argument)
- size-strategy (optional) when new column is shorter than dataset row count, following strategies are applied:
  - :cycle (default) repeat data
  - :na append missing values

Function works on grouped dataset.

Add single value as column

```
(api/add-or-update-column DS : V5 "X")
```

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:V5
1	1	0.5000	A	X
2	2	1.000	В	X
1	3	1.500	$\mathbf{C}$	X
2	4	0.5000	A	X
1	5	1.000	В	X
2	6	1.500	$\mathbf{C}$	X
1	7	0.5000	A	X
2	8	1.000	В	X
1	9	1.500	$\mathbf{C}$	X

Replace one column (column is trimmed)

```
(api/add-or-update-column DS :V1 (repeatedly rand))
```

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
0.3955	1	0.5000	A
0.7369	2	1.000	В
0.8222	3	1.500	$\mathbf{C}$
0.5205	4	0.5000	A
0.4425	5	1.000	В
0.7587	6	1.500	$\mathbf{C}$

:V1	:V2	:V3	:V4
0.7391	7	0.5000	A
0.9806	8	1.000	В
0.9969	9	1.500	$\mathbf{C}$

# ${\rm Copy\ column}$

```
(api/add-or-update-column DS : V5 (DS : V1))
```

# $\underline{\text{unnamed } [9\ 5]}$ :

:V1	:V2	:V3	:V4	:V5
1	1	0.5000	A	1
2	2	1.000	В	2
1	3	1.500	$\mathbf{C}$	1
2	4	0.5000	A	2
1	5	1.000	В	1
2	6	1.500	С	2
1	7	0.5000	A	1
2	8	1.000	В	2
1	9	1.500	$\mathbf{C}$	1

When function is used, argument is whole dataset and the result should be column, sequence or single value (api/add-or-update-column DS :row-count api/row-count)

# \_unnamed [9 5]:

:V1	:V2	:V3	:V4	:row-count
1	1	0.5000	A	9
2	2	1.000	В	9
1	3	1.500	$\mathbf{C}$	9
2	4	0.5000	A	9
1	5	1.000	В	9
2	6	1.500	$\mathbf{C}$	9
1	7	0.5000	A	9
2	8	1.000	В	9
1	9	1.500	$\mathbf{C}$	9

Above example run on grouped dataset, applies function on each group separately.

```
(-> DS
     (api/group-by :V1)
     (api/add-or-update-column :row-count api/row-count)
     (api/ungroup))
```

 $\underline{\phantom{a}}$ unnamed [9 5]:

:V1	:V2	:V3	:V4	:row-count
1	1	0.5000	A	5
1	3	1.500	$\mathbf{C}$	5
1	5	1.000	В	5
1	7	0.5000	A	5
1	9	1.500	$\mathbf{C}$	5
2	2	1.000	В	4
2	4	0.5000	A	4
2	6	1.500	$\mathbf{C}$	4
2	8	1.000	В	4

When column which is added is longer than row count in dataset, column is trimmed. When column is shorter, it's cycled or missing values are appended.

```
(api/add-or-update-column DS :V5 [:r :b])
```

 $\underline{\phantom{a}}$ unnamed [9 5]:

:V1	:V2	:V3	:V4	:V5
1	1	0.5000	A	:r
2	2	1.000	В	:b
1	3	1.500	С	:r
2	4	0.5000	A	:b
1	5	1.000	В	:r
2	6	1.500	$\mathbf{C}$	:b
1	7	0.5000	A	:r
2	8	1.000	В	:b
1	9	1.500	$\mathbf{C}$	:r

```
(api/add-or-update-column DS :V5 [:r :b] :na)
```

 $\underline{\text{unnamed } [9 5]}$ :

:V1	:V2	:V3	:V4	:V5
1	1	0.5000	A	:r
2	2	1.000	В	:b
1	3	1.500	$\mathbf{C}$	
2	4	0.5000	A	
1	5	1.000	В	
2	6	1.500	$\mathbf{C}$	
1	7	0.5000	$\mathbf{A}$	
2	8	1.000	В	
1	9	1.500	$\mathbf{C}$	

The same applies for grouped dataset

```
(-> DS
    (api/group-by :V3)
    (api/add-or-update-column :V5 [:r :b] :na)
```

```
(api/ungroup))
```

 $\underline{\text{unnamed } [9\ 5]}$ :

:V1	:V2	:V3	:V4	:V5
2	2	1.000	В	:r
1	5	1.000	В	:b
2	8	1.000	В	
1	1	0.5000	A	:r
2	4	0.5000	A	:b
1	7	0.5000	A	
1	3	1.500	$\mathbf{C}$	:r
2	6	1.500	$\mathbf{C}$	:b
1	9	1.500	С	

Let's use other column to fill groups

```
(-> DS
     (api/group-by :V3)
     (api/add-or-update-column :V5 (DS :V2))
     (api/ungroup))
```

 $\underline{\text{unnamed } [9\ 5]}$ :

:V1	:V2	:V3	:V4	:V5
2	2	1.000	В	1
1	5	1.000	В	2
2	8	1.000	В	3
1	1	0.5000	A	1
2	4	0.5000	A	2
1	7	0.5000	A	3
1	3	1.500	$\mathbf{C}$	1
2	6	1.500	$\mathbf{C}$	2
1	9	1.500	С	3

In case you want to add or update several columns you can call add-or-update-columns and provide map where keys are column names, vals are columns.

 $\underline{\phantom{a}}$ unnamed [9 6]:

:V1	:V2	:V3	:V4	:V5	:V6
2	1	0.5000	A	:A	11
3	2	1.000	В	:В	11
2	3	1.500	$\mathbf{C}$	:C	11
3	4	0.5000	A	:A	11
2	5	1.000	В	:В	11

:V1	:V2	:V3	:V4	:V5	:V6
3	6	1.500	С	:C	11
2	7	0.5000	A	:A	11
3	8	1.000	В	:В	11
2	9	1.500	$\mathbf{C}$	:C	11

# Map

The other way of creating or updating column is to map columns as regular map function. The arity of mapping function should be the same as number of selected columns.

# Arguments:

- ds dataset
- column-name target column name
- map-fn mapping function
- columns-selector columns selected
- meta-field (optional) column selector option

Let's add numerical columns together

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:sum-of-numbers
1	1	0.5000	A	2.500
2	2	1.000	В	5.000
1	3	1.500	С	5.500
2	4	0.5000	A	6.500
1	5	1.000	В	7.000
2	6	1.500	$\mathbf{C}$	9.500
1	7	0.5000	A	8.500
2	8	1.000	В	11.00
1	9	1.500	С	11.50

The same works on grouped dataset

 $\underline{\text{unnamed } [9\ 5]}$ :

:V1	:V2	:V3	:V4	:sum-of-numbers
1	1	0.5000	A	2.500
2	4	0.5000	A	6.500
1	7	0.5000	$\mathbf{A}$	8.500
2	2	1.000	В	5.000

:V1	:V2	:V3	:V4	:sum-of-numbers
1	5	1.000	В	7.000
2	8	1.000	В	11.00
1	3	1.500	$\mathbf{C}$	5.500
2	6	1.500	$\mathbf{C}$	9.500
1	9	1.500	$\mathbf{C}$	11.50

#### Reorder

To reorder columns use columns selectors to choose what columns go first. The unseleted columns are appended to the end.

```
(api/reorder-columns DS :V4 [:V3 :V2] :V1)
```

\_unnamed [9 4]:

:V4	:V2	:V3	:V1
A	1	0.5000	1
В	2	1.000	2
$\mathbf{C}$	3	1.500	1
A	4	0.5000	2
В	5	1.000	1
$\mathbf{C}$	6	1.500	2
A	7	0.5000	1
В	8	1.000	$^2$
С	9	1.500	1

This function doesn't let you select meta field, so you have to call column-names in such case. Below we want to add integer columns at the end.

```
(api/reorder-columns DS (api/column-names DS (complement #{:int64}) :datatype))
```

\_unnamed [9 4]:

:V3	:V4	:V1	:V2
${0.5000}$	A	1	1
1.000	В	2	2
1.500	С	1	3
0.5000	A	2	4
1.000	В	1	5
1.500	$\mathbf{C}$	2	6
0.5000	A	1	7
1.000	В	2	8
1.500	$\mathbf{C}$	1	9

### Type conversion

To convert column into given datatype can be done using convert-column-type function. Not all the types can be converted automatically also some types require slow parsing (every conversion from string). In case where conversion is not possible you can pass conversion function.

# Arguments:

- ds dataset
- Two options:
  - coltype-map in case when you want to convert several columns, keys are column names, vals are new types
  - colname and new-type column name and new datatype

### new-type can be:

- a type like :int64 or :string
- or pair of datetime and conversion function

After conversion additional infomation is given on problematic values.

The other conversion is casting column into java array (->array) of the type column or provided as argument. Grouped dataset returns sequence of arrays.

#### Basic conversion

```
(-> DS
     (api/convert-column-type :V1 :float64)
     (api/info :columns))
```

\_unnamed :column info [4 6]:

:name	:size	:datatype	:unparsed-indexes	:unparsed-data	:categorical?
:V1 :V2 :V3	9 9 9	:float64 :int64 :float64	{}		4
:V4	9	:string			true

Using custom converter. Let's treat : V4 as haxadecimal values. See that this way we can map column to any value.

```
(-> DS
     (api/convert-column-type :V4 [:int16 #(Integer/parseInt % 16)]))
```

# \_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	10
2	2	1.000	11
1	3	1.500	12
2	4	0.5000	10
1	5	1.000	11
2	6	1.500	12
1	7	0.5000	10
2	8	1.000	11
1	9	1.500	12

You can process several columns at once

\_unnamed :column info [4 5]:

:name	:size	:datatype	:unparsed-indexes	:unparsed-data
:V1 :V2 :V3 :V4	9 9 9	:float64 :object :boolean :object	{} {} {}	

Function works on the grouped dataset

```
(-> DS
     (api/group-by :V1)
     (api/convert-column-type :V1 :float32)
     (api/ungroup)
     (api/info :columns))
```

\_unnamed :column info [4 6]:

:name	:size	:datatype	:unparsed-indexes	:unparsed-data	:categorical?
:V1	9	:float32	{}		
:V2	9	:int $64$			
:V3	9	:float64			
:V4	9	:string			true

Double array conversion.

```
(api/->array DS :V1)

#object["[J" 0x1bc6993 "[J@1bc6993"]
```

Function also works on grouped dataset

```
(-> DS
    (api/group-by :V3)
    (api/->array :V2))
```

(#object["[J" 0x49cd7336 "[J@49cd7336"] #object["[J" 0x2b4acf41 "[J@2b4acf41"] #object["[J" 0x29483e1c

You can also cast the type to the other one (if casting is possible):

```
(api/->array DS :V4 :string)
(api/->array DS :V1 :float32)
```

```
#object["[Ljava.lang.String;" 0x31b8c04c "[Ljava.lang.String;@31b8c04c"]
#object["[F" 0x2abc68e0 "[F@2abc68e0"]
```

#### Rows

Rows can be selected or dropped using various selectors:

- row id(s) row index as number or sequence of numbers (first row has index 0, second 1 and so on)
- sequence of true/false values
- filter by predicate (argument is row as a map)

When predicate is used you may want to limit columns passed to the function (limit-columns option).

Additionally you may want to precalculate some values which will be visible for predicate as additional columns. It's done internally by calling add-or-update-columns on a dataset. :pre is used as a column definitions.

#### Select

Select fourth row

```
(api/select-rows DS 4)
```

 $\underline{\quad}$  unnamed [1 4]:

:V1	:V2	:V3	:V4
1	5	1.000	В

Select 3 rows

```
(api/select-rows DS [1 4 5])
```

 $\underline{\quad}$  unnamed [3 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
1	5	1.000	В
2	6	1.500	С

Select rows using sequence of true/false values

```
(api/select-rows DS [true nil nil true])
```

 $\underline{\text{unnamed } [2\ 4]}$ :

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	4	0.5000	A

Select rows using predicate

```
(api/select-rows DS (comp #(< % 1) :V3))
```

 $\underline{\quad}$  unnamed [3 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	4	0.5000	A
1	7	0.5000	A

The same works on grouped dataset, let's select first row from every group.

```
(-> DS
    (api/group-by :V1)
    (api/select-rows 0)
    (api/ungroup))
```

 $\underline{\quad}$  unnamed [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В

If you want to select : V2 values which are lower than or equal mean in grouped dataset you have to precalculate it using :pre.

 $\underline{\quad}$  unnamed [6 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	4	0.5000	A
2	2	1.000	В
1	5	1.000	В
1	3	1.500	$\mathbf{C}$
2	6	1.500	С

### Drop

drop-rows removes rows, and accepts exactly the same parameters as select-rows

Drop values lower than or equal : V2 column mean in grouped dataset.

 $\underline{\phantom{a}}$  unnamed [3 4]:

:V1	:V2	:V3	:V4
1	7	0.5000	A
2	8	1.000	В
1	9	1.500	$\mathbf{C}$

### Other

There are several function to select first, last, random rows, or display head, tail of the dataset. All functions work on grouped dataset.

First row

(api/first DS)

\_unnamed [1 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A

Last row

(api/last DS)

 $\underline{\phantom{a}}$ unnamed [1 4]:

:V1	:V2	:V3	:V4
1	9	1.500	С

Random row (single)

(api/rand-nth DS)

 $\underline{\quad}$  unnamed [1 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В

Random n (default: row count) rows with repetition.

# (api/random DS)

 $\underline{\phantom{a}}$ unnamed [9 4]:

:V1	:V2	:V3	:V4
1	9	1.500	$\overline{C}$
2	6	1.500	$\mathbf{C}$
1	3	1.500	$\mathbf{C}$
2	2	1.000	В
2	2	1.000	В
2	6	1.500	$\mathbf{C}$
1	1	0.5000	A
1	5	1.000	В
1	9	1.500	$\mathbf{C}$

Five random rows with repetition

# (api/random DS 5)

 $\underline{\phantom{a}}$ unnamed [5 4]:

:V1	:V2	:V3	:V4
2	4	0.5000	A
2	4	0.5000	A
1	5	1.000	В
1	1	0.5000	A
2	4	0.5000	A

Five random, non-repeating rows

(api/random DS 5 {:repeat? false})

 $\underline{\quad}$  unnamed [5 4]:

:V1	:V2	:V3	:V4
2	8	1.000	В
1	5	1.000	В
2	2	1.000	В
1	1	0.5000	A
2	6	1.500	С

Shuffle dataset

(api/shuffle DS)

 $\underline{\phantom{a}}$ unnamed [9 4]:

:V1	:V2	:V3	:V4
2	8	1.000	В
2	6	1.500	$\mathbf{C}$
1	5	1.000	В
1	9	1.500	$\mathbf{C}$
1	7	0.5000	A
2	4	0.5000	A
1	3	1.500	$\mathbf{C}$
1	1	0.5000	A
2	2	1.000	В

First n rows (default 5)

(api/head DS)

 $\underline{\phantom{a}}$ unnamed [5 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
2	4	0.5000	A
1	5	1.000	В

Last n rows (default 5)

(api/tail DS)

 $\underline{\phantom{a}}$ unnamed [5 4]:

:V1	:V2	:V3	:V4
1	5	1.000	В
2	6	1.500	$\mathbf{C}$
1	7	0.5000	A
2	8	1.000	В
1	9	1.500	$\mathbf{C}$

Select 5 random rows from each group

```
(-> DS
     (api/group-by :V4)
     (api/random 5)
     (api/ungroup))
```

\_unnamed [15 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A

:V1	:V2	:V3	:V4
2	4	0.5000	A
1	1	0.5000	A
2	4	0.5000	A
2	4	0.5000	A
2	8	1.000	В
2	8	1.000	В
2	2	1.000	В
1	5	1.000	В
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$
1	9	1.500	$\mathbf{C}$
1	3	1.500	С

## Aggregate

Aggregating is a function which produces single row out of dataset.

Aggregator is a function or sequence or map of functions which accept dataset as an argument and result single value, sequence of values or map.

Where map is given as an input or result, keys are treated as column names.

Grouped dataset is ungrouped after aggreation. This can be turned of by setting :ungroup to false. In case you want to pass additional ungrouping parameters add them to the options.

By default resulting column names are prefixed with summary prefix (set it with :default-column-name-prefix option).

Sequential result is spread into separate columns

```
(api/aggregate DS #(take 5(% :V2)))
```

 $\underline{\quad}$  unnamed [1 5]:

:summary-0	:summary-1	:summary-2	:summary-3	:summary-4
1	2	3	4	5

You can combine all variants and rename default prefix

\_unnamed [1 5]:

:V2-value-0-0	:V2-value-0-1	:V2-value-0-2	:sum-v1	:prod-v3
1	2	3	13	0.4219

Processing grouped dataset

\_unnamed [3 6]:

:V4	:V2-value-0-0	:V2-value-0-1	:V2-value-0-2	:sum-v1	:prod-v3
В	2	5	8	5	1.000
$\mathbf{C}$	3	6	9	4	3.375
A	1	4	7	4	0.1250

Result of aggregating is automatically ungrouped, you can skip this step by stetting :ungroup option to false.

 $\underline{\phantom{a}}$ unnamed [3 3]:

:name	:group-id	:data
{:V3 1.0}	0	unnamed [1 5]:
$\{: V3 \ 0.5\}$	1	$\underline{}$ unnamed [1 5]:

:name	:group-id	:data
{:V3 1.5}	2	_unnamed [1 5]:

#### Order

Ordering can be done by column(s) or any function operating on row. Possible order can be:

- :asc for ascending order (default)
- :desc for descending order
- custom comparator

:limit-columns limits row map provided to ordering functions.

Order by single column, ascending

(api/order-by DS :V1)

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	3	1.500	$\mathbf{C}$
1	5	1.000	В
1	7	0.5000	A
1	9	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$
2	4	0.5000	A
2	8	1.000	В
2	2	1.000	В

Descending order

(api/order-by DS :V1 :desc)

 $\underline{\text{unnamed } [9 \ 4]}$ :

:V1	:V2	:V3	:V4
2	2	1.000	В
2	4	0.5000	A
2	6	1.500	$\mathbf{C}$
2	8	1.000	В
1	5	1.000	В
1	3	1.500	$\mathbf{C}$
1	7	0.5000	A
1	1	0.5000	A
1	9	1.500	$\mathbf{C}$

Order by two columns

# (api/order-by DS [:V1 :V2])

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	3	1.500	$\mathbf{C}$
1	5	1.000	В
1	7	0.5000	A
1	9	1.500	$\mathbf{C}$
2	2	1.000	В
2	4	0.5000	A
2	6	1.500	$\mathbf{C}$
2	8	1.000	В

Use different orders for columns

```
(api/order-by DS [:V1 :V2] [:asc :desc])
```

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	9	1.500	С
1	7	0.5000	A
1	5	1.000	В
1	3	1.500	$\mathbf{C}$
1	1	0.5000	A
2	8	1.000	В
2	6	1.500	$\mathbf{C}$
2	4	0.5000	A
2	2	1.000	В

# (api/order-by DS [:V1 :V2] [:desc :desc])

 $\underline{\phantom{a}}$ unnamed [9 4]:

:V1	:V2	:V3	:V4
2	8	1.000	В
2	6	1.500	$\mathbf{C}$
2	4	0.5000	A
2	$^2$	1.000	В
1	9	1.500	$\mathbf{C}$
1	7	0.5000	A
1	5	1.000	В
1	3	1.500	$\mathbf{C}$
1	1	0.5000	A

```
(api/order-by DS [:V1 :V3] [:desc :asc])
```

### $\underline{\quad}$ unnamed [9 4]:

:V1	:V2	:V3	:V4
2	4	0.5000	A
2	2	1.000	В
2	8	1.000	В
2	6	1.500	$^{\mathrm{C}}$
1	1	0.5000	A
1	7	0.5000	A
1	5	1.000	В
1	3	1.500	$\mathbf{C}$
1	9	1.500	$\mathbf{C}$

Custom function can be used to provied ordering key. Here order by :V4 descending, then by product of other columns ascending.

### $\underline{\text{unnamed } [9 \ 4]}$ :

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	7	0.5000	A
2	4	0.5000	A
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
1	5	1.000	В
1	9	1.500	$\mathbf{C}$
2	8	1.000	В
2	6	1.500	С

Custom comparator also can be used in case objects are not comparable by default. Let's define artificial one: if euclidean distance is lower than 2, compare along  ${\bf z}$  else along  ${\bf x}$  and  ${\bf y}$ . We use first three columns for that.

### #'user/dist

(compare [x1 y1] [x2 y2])))))

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	5	1.000	В
1	7	0.5000	A
1	9	1.500	$\mathbf{C}$
2	2	1.000	В
2	4	0.5000	A
1	3	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$
2	8	1.000	В

## Unique

Remove rows which contains the same data. By default unique-by removes duplicates from whole dataset. You can also pass list of columns or functions (similar as in group-by) to remove duplicates limited by them. Default strategy is to keep the first row. More strategies below.

unique-by works on groups

Remove duplicates from whole dataset

(api/unique-by DS)

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
2	4	0.5000	A
1	5	1.000	В
2	6	1.500	$\mathbf{C}$
1	7	0.5000	A
2	8	1.000	В
1	9	1.500	С

Remove duplicates from each group selected by column.

(api/unique-by DS : V1)

 $\underline{\quad}$  unnamed [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В

Pair of columns

```
(api/unique-by DS [:V1 :V3])
```

 $\underline{\text{unnamed } [6 \ 4]}$ :

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	3	1.500	$\mathbf{C}$
2	4	0.5000	A
1	5	1.000	В
2	6	1.500	$\mathbf{C}$

Also function can be used, split dataset by modulo 3 on columns : V2

```
(api/unique-by DS (fn [m] (mod (:V2 m) 3)))
```

 $\underline{\phantom{a}}$ unnamed [3 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	3	1.500	$\mathbf{C}$

The same can be achived with group-by

```
(-> DS
    (api/group-by (fn [m] (mod (:V2 m) 3)))
    (api/first)
    (api/ungroup))
```

 $\underline{\quad}$  unnamed [3 4]:

:V1	:V2	:V3	:V4
1	3	1.500	$\overline{C}$
1	1	0.5000	A
2	2	1.000	В

Grouped dataset

```
(-> DS
    (api/group-by :V4)
    (api/unique-by :V1)
    (api/ungroup))
```

\_unnamed [6 4]:

:V1	:V2	:V3	·V4
. V 1	: V Z	. v o	. V 4
1	1	0.5000	A
2	4	0.5000	A
2	2	1.000	В
1	5	1.000	В
1	3	1.500	$\mathbf{C}$
2	6	1.500	$\mathbf{C}$

## Strategies

There are 4 strategies defined:

- :first select first row (default)
- :last select last row
- :random select random row
- any function apply function to a columns which are subject of uniqueness

Last

```
(api/unique-by DS :V1 {:strategy :last})
```

\_unnamed [2 4]:

:V1	:V2	:V3	:V4
2	8	1.000	В
1	9	1.500	$\mathbf{C}$

Random

```
(api/unique-by DS :V1 {:strategy :random})
```

\_unnamed [2 4]:

:V1	:V2	:V3	:V4
2	2	1.000	В
1	7	0.5000	A

Pack columns into vector

```
(api/unique-by DS : V4 {:strategy vec})
```

\_unnamed [3 3]:

:V1	:V2	:V3
$   \begin{array}{c c}     \hline     [2 \ 1 \ 2] \\     [1 \ 2 \ 1] \\     [1 \ 2 \ 1]   \end{array} $	[2 5 8] [3 6 9] [1 4 7]	[1.0 1.0 1.0] [1.5 1.5 1.5] [0.5 0.5 0.5]

Sum columns

```
(api/unique-by DS : V4 {:strategy (partial reduce +)})
```

 $\underline{\quad}$  unnamed [3 3]:

:V1	:V2	:V3
5	15	3.000
4	18	4.500
4	12	1.500

Group by function and apply functions

```
(api/unique-by DS (fn [m] (mod (:V2 m) 3)) {:strategy vec})
```

 $\underline{\quad}$  unnamed [3 4]:

:V1	:V2	:V3	:V4
$   \begin{array}{c c}     \hline     [1 \ 2 \ 1] \\     [1 \ 2 \ 1] \\     [2 \ 1 \ 2]   \end{array} $	[3 6 9]	[1.5 1.5 1.5]	["C" "C" "C"]
	[1 4 7]	[0.5 0.5 0.5]	["A" "A" "A"]
	[2 5 8]	[1.0 1.0 1.0]	["B" "B" "B"]

Grouped dataset

```
(-> DS
    (api/group-by :V1)
    (api/unique-by (fn [m] (mod (:V2 m) 3)) {:strategy vec})
    (api/ungroup {:add-group-as-column :from-V1}))
```

\_unnamed [6 5]:

:from-V1	:V1	:V2	:V3	:V4
1	[1 1]	[3 9]	$[1.5 \ 1.5]$	["C" "C"]
1	$[1 \ 1]$	$[1 \ 7]$	$[0.5 \ 0.5]$	["A" "A"]
1	[1]	[5]	[1.0]	["B"]
2	[2]	[6]	[1.5]	["C"]
2	[2]	[4]	[0.5]	["A"]
2	$[2\ 2]$	$[2 \ 8]$	$[1.0 \ 1.0]$	["B" "B"]

## Missing

When dataset contains missing values you can select or drop rows with missing values or replace them using some strategy.

column-selector can be used to limit considered columns

Let's define dataset which contains missing values

 $\mathsf{DSm}$ 

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
	3		$\mathbf{C}$
1	4	1.500	A
2	5	0.5000	В
	6	1.000	$\mathbf{C}$
1	7		A
2	8	1.500	В
	9	0.5000	С

## Select

Select rows with missing values

(api/select-missing DSm)

 $\underline{\quad}$  unnamed [4 4]:

3 C 6 1.000 C	:V1	:V2 :	v3	:V4
0 2.000 0		3		С
4 <del>-</del> A		6	1.000	$\mathbf{C}$
1 7 A	1	7		A
9   0.5000   C		9 (	0.5000	С

Select rows with missing values in :V1

(api/select-missing DSm :V1)

 $\underline{\phantom{a}}$ unnamed [3 4]:

:V1	:V2	:V3	:V4
	3		$^{\rm C}$
	6	1.000	$\mathbf{C}$
	9	0.5000	$\mathbf{C}$

The same with grouped dataset

```
(-> DSm
    (api/group-by :V4)
    (api/select-missing :V3)
```

# (api/ungroup))

 $\underline{\phantom{a}}$ unnamed [2 4]:

:V1	:V2	:V3	:V4
1	7		A
	3		$\mathbf{C}$

## Drop

Drop rows with missing values

(api/drop-missing DSm)

 $\underline{\phantom{a}}$  unnamed [5 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	4	1.500	A
2	5	0.5000	В
2	8	1.500	В

Drop rows with missing values in  $: \mathtt{V1}$ 

(api/drop-missing DSm :V1)

 $\underline{\phantom{a}}$ unnamed [6 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
2	2	1.000	В
1	4	1.500	A
2	5	0.5000	В
1	7		A
2	8	1.500	В

The same with grouped dataset

```
(-> DSm
     (api/group-by :V4)
     (api/drop-missing :V1)
     (api/ungroup))
```

 $\underline{\phantom{a}}$ unnamed [6 4]:

:V1	:V2	:V3	:V4
1	1	0.5000	A
1	4	1.500	A

:V1	:V2	:V3	:V4
1	7		A
2	2	1.000	В
2	5	0.5000	В
2	8	1.500	В

### Replace

Missing values can be replaced using several strategies. replace-missing accepts:

- dataset
- column selector
- value
  - single value
  - sequence of values (cycled)
  - function, applied on column(s) with stripped missings
- strategy (optional)

Strategies are:

- :value replace with given value (default)
- :up copy values up
- :down copy values down

Let's define special dataset here:

```
(def DSm2 (api/dataset {:a [nil nil nil 1.0 2 nil 4 nil 11 nil nil] :b [2.0 2 2 nil nil 3 nil 3 4 5 5]}))
```

#### DSm2

\_unnamed [11 2]:

:a	:b
	2.000
	2.000
	2.000
1.000	
2.000	
	3.000
4.000	
	3.000
11.00	4.000
	5.000
	5.000

Replace missing with single value in whole dataset

```
(api/replace-missing DSm2 999)
```

:a	:b
999.0	2.000

:a	:b
999.0	2.000
999.0	2.000
1.000	999.0
2.000	999.0
999.0	3.000
4.000	999.0
999.0	3.000
11.00	4.000
999.0	5.000
999.0	5.000

Replace missing with single value in :a column

```
(api/replace-missing DSm2 :a 999)
```

\_unnamed [11 2]:

:a	:b
999.0	2.000
999.0	2.000
999.0	2.000
1.000	
2.000	
999.0	3.000
4.000	
999.0	3.000
11.00	4.000
999.0	5.000
999.0	5.000

Replace missing with sequence in :a column

```
(api/replace-missing DSm2 :a [-999 -998 -997])
```

:a	:b
-999.0	2.000
-998.0	2.000
-997.0	2.000
1.000	
2.000	
-999.0	3.000
4.000	
-998.0	3.000
11.00	4.000
-997.0	5.000
-999.0	5.000

Replace missing with a function (mean)

```
(api/replace-missing DSm2 :a tech.v2.datatype.functional/mean)
```

\_unnamed [11 2]:

:a	:b
4.500	2.000
4.500	2.000
4.500	2.000
1.000	
2.000	
4.500	3.000
4.000	
4.500	3.000
11.00	4.000
4.500	5.000
4.500	5.000

Using :down strategy, fills gaps with values from above. You can see that if missings are at the beginning, they are left missing.

```
(api/replace-missing DSm2 [:a :b] nil :down)
```

\_unnamed [11 2]:

:a	:b
	2.000
	2.000
	2.000
1.000	2.000
2.000	2.000
2.000	3.000
4.000	3.000
4.000	3.000
11.00	4.000
11.00	5.000
11.00	5.000

To fix above issue you can provide value

```
(api/replace-missing DSm2 [:a :b] 999 :down)
```

:a	:b
999.0	2.000
999.0	2.000
999.0	2.000

:a	:b
1.000	2.000
2.000	2.000
2.000	3.000
4.000	3.000
4.000	3.000
11.00	4.000
11.00	5.000
11.00	5.000

The same applies for :up strategy which is opposite direction.

```
(api/replace-missing DSm2 [:a :b] 999 :up)
```

\_unnamed [11 2]:

:a	:b
1.000	2.000
1.000	2.000
1.000	2.000
1.000	3.000
2.000	3.000
4.000	3.000
4.000	3.000
11.00	3.000
11.00	4.000
999.0	5.000
999.0	5.000

We can use a function which is applied after applying :up or :down

```
(api/replace-missing DSm2 [:a :b] tech.v2.datatype.functional/mean :down)
```

:a	:b
4.500	2.000
4.500	2.000
4.500	2.000
1.000	2.000
2.000	2.000
2.000	3.000
4.000	3.000
4.000	3.000
11.00	4.000
11.00	5.000
11.00	5.000

### Join/Separate Columns

Joining or separating columns are operations which can help to tidy messy dataset.

- join-columns joins content of the columns (as string concatenation or other structure) and stores it in new column
- separate-column splits content of the columns into set of new columns

#### Join

join-columns accepts:

- dataset
- column selector (as in select-columns)
- options
  - :separator (default "-")
  - $-\,$  :drop-columns? whether to drop source columns or not (default  $\mathtt{true})$
  - :result-type
    - \* :map packs data into map
    - \* :seq packs data into sequence
    - \* :string join strings with separator (default)
    - \* or custom function which gets row as a vector
  - :missing-subst substitution for missing value

Default usage. Create : joined column out of other columns.

```
(api/join-columns DSm :joined [:V1 :V2 :V4])
```

 $\underline{\phantom{a}}$ unnamed [9 2]:

:V3	:joined
0.5000	1-1-A
1.000	2-2-B
	3-C
1.500	1-4-A
0.5000	2-5-B
1.000	6-C
	1-7-A
1.500	2 - 8 - B
0.5000	9-C

Without dropping source columns.

```
(api/join-columns DSm :joined [:V1 :V2 :V4] {:drop-columns? false})
```

 $\underline{\text{unnamed } [9 5]}$ :

:V1	:V2	:V3	:V4	:joined
1	1	0.5000	A	1-1-A
2	2	1.000	В	2-2-B
	3		$\mathbf{C}$	3-C
1	4	1.500	A	1-4-A

:V1	:V2	:V3	:V4	:joined
2	5	0.5000	В	2-5-B
	6	1.000	$\mathbf{C}$	6-C
1	7		A	1-7-A
2	8	1.500	В	2 - 8 - B
	9	0.5000	$\mathbf{C}$	9-C

Let's replace missing value with "NA" string.

```
(api/join-columns DSm :joined [:V1 :V2 :V4] {:missing-subst "NA"})
```

\_unnamed [9 2]:

:V3	:joined
0.5000	1-1-A
1.000	2-2-B
	NA-3-C
1.500	1-4-A
0.5000	$2\text{-}5\text{-}\mathrm{B}$
1.000	NA-6-C
	1-7-A
1.500	2 - 8 - B
0.5000	NA-9-C

We can use custom separator.

 $\underline{\phantom{a}}$ unnamed [9 2]:

:V3	:joined
0.5000	1/1/A
1.000	2/2/B
	./3/C
1.500	1/4/A
0.5000	2/5/B
1.000	./6/C
	1/7/A
1.500	2/8/B
0.5000	./9/C

Or even sequence of separators.

:V3	:joined
0.5000	1-1/A
1.000	2-2/B
	3/C
1.500	1-4/A
0.5000	2 - 5/B
1.000	-6/C
	1-7/A
1.500	2 - 8/B
0.5000	9/C

The other types of results, map:

```
(api/join-columns DSm :joined [:V1 :V2 :V4] {:result-type :map})
```

 $\underline{\phantom{a}}$ unnamed [9 2]:

:V3	:joined
0.5000	{:V1 1, :V2 1, :V4 "A"}
1.000	{:V1 2, :V2 2, :V4 "B"}
	{:V1 nil, :V2 3, :V4 "C"}
1.500	{:V1 1, :V2 4, :V4 "A"}
0.5000	{:V1 2, :V2 5, :V4 "B"}
1.000	{:V1 nil, :V2 6, :V4 "C"}
	{:V1 1, :V2 7, :V4 "A"}
1.500	{:V1 2, :V2 8, :V4 "B"}
0.5000	{:V1 nil, :V2 9, :V4 "C"}

Sequence

```
(api/join-columns DSm :joined [:V1 :V2 :V4] {:result-type :seq})
```

\_unnamed [9 2]:

joined
1 1 "A")
2 2 "B")
[nil 3 "C"]
(1 4 "A")
(2 5 "B")
[nil 6 "C"]
(1 7 "A")
(2 8 "B")
nil 9 "C"

Custom function, calculate hash

```
(api/join-columns DSm :joined [:V1 :V2 :V4] {:result-type hash})
```

# $\underline{\quad}$ unnamed [9 2]:

:V3	:joined
0.5000	535226087
1.000	1128801549
	-1842240303
1.500	2022347171
0.5000	1884312041
1.000	-1555412370
	1640237355
1.500	-967279152
0.5000	1128367958

# Grouped dataset

```
(-> DSm
    (api/group-by :V4)
    (api/join-columns :joined [:V1 :V2 :V4])
    (api/ungroup))
```

# $\underline{\phantom{a}}$ unnamed [9 2]:

:V3	:joined
0.5000	1-1-A
1.500	1-4-A
	1-7-A
1.000	2-2-B
0.5000	$2\text{-}5\text{-}\mathrm{B}$
1.500	2 - 8 - B
	3-C
1.000	6-C
0.5000	9-C

# Tidyr examples

source

#'user/df

df

 $\underline{\phantom{a}}$ unnamed [4 2]:

 $\underline{\quad}$  unnamed [4 3]:

 $\underline{\phantom{a}}$ unnamed [4 3]:

#### Separate

Column can be also separated into several other columns using string as separator, regex or custom function. Arguments:

- dataset
- source column
- target columns
- separator as:
  - string it's converted to regular expression and passed to clojure.string/split function
  - regex
  - or custom function (default: identity)
- options
  - :drop-columns? whether drop source column or not (default: true)
  - :missing-subst values which should be treated as missing, can be set, sequence, value or function (default: "")

Custom function (as separator) should return sequence of values for given value.

Separate float into integer and factional values

\_unnamed [9 5]:

:V1	:V2	:int-part	:frac-part	:V4
1	1	0.000	0.5000	A
2	2	1.000	0.000	В
1	3	1.000	0.5000	$\mathbf{C}$
2	4	0.000	0.5000	A
1	5	1.000	0.000	В
2	6	1.000	0.5000	$\mathbf{C}$
1	7	0.000	0.5000	A
2	8	1.000	0.000	В
1	9	1.000	0.5000	$\mathbf{C}$

Source column can be kept

\_unnamed [9 6]:

:V1	:V2	:V3	:int-part	:frac-part	:V4
1	1	0.5000	0.000	0.5000	A
2	2	1.000	1.000	0.000	В
1	3	1.500	1.000	0.5000	$\mathbf{C}$
2	4	0.5000	0.000	0.5000	A
1	5	1.000	1.000	0.000	В
2	6	1.500	1.000	0.5000	$\mathbf{C}$
1	7	0.5000	0.000	0.5000	A
2	8	1.000	1.000	0.000	В
1	9	1.500	1.000	0.5000	$\mathbf{C}$

We can treat 0 or 0.0 as missing value

\_unnamed [9 5]:

:V1	:V2	:int-part	:frac-part	:V4
1	1		0.5000	A
2	$^2$	1		В
1	3	1	0.5000	$\mathbf{C}$
2	4		0.5000	A
1	5	1		В

:V1	:V2	:int-part	:frac-part	:V4
2	6	1	0.5000	$\overline{C}$
1	7		0.5000	A
2	8	1		В
1	9	1	0.5000	$\mathbf{C}$

Works on grouped dataset

 $\underline{\phantom{a}}$ unnamed [9 5]:

:V1	:V2	:int-part	:fract-part	:V4
1	1	0.000	0.5000	A
2	4	0.000	0.5000	A
1	7	0.000	0.5000	A
2	2	1.000	0.000	В
1	5	1.000	0.000	В
2	8	1.000	0.000	В
1	3	1.000	0.5000	$\mathbf{C}$
2	6	1.000	0.5000	$\mathbf{C}$
1	9	1.000	0.5000	С

Join and separate together.

```
(-> DSm
    (api/join-columns : joined [:V1 :V2 :V4] {:result-type :map})
    (api/separate-column : joined [:v1 :v2 :v4] (juxt :V1 :V2 :V4)))
```

 $\underline{\phantom{a}}$ unnamed [9 4]:

:V3	:v1	:v2	:v4
0.5000	1.000	1.000	A
1.000	2.000	2.000	В
		3.000	$\mathbf{C}$
1.500	1.000	4.000	A
0.5000	2.000	5.000	В
1.000		6.000	$\mathbf{C}$
	1.000	7.000	A
1.500	2.000	8.000	В
0.5000		9.000	$\mathbf{C}$

```
(-> DSm
    (api/join-columns : joined [:V1 :V2 :V4] {:result-type :seq})
    (api/separate-column : joined [:v1 :v2 :v4] identity))
```

\_unnamed [9 4]:

:V3	:v1	:v2	:v4
0.5000	1.000	1.000	A
1.000	2.000	2.000	В
		3.000	$\mathbf{C}$
1.500	1.000	4.000	A
0.5000	2.000	5.000	В
1.000		6.000	$\mathbf{C}$
	1.000	7.000	A
1.500	2.000	8.000	В
0.5000		9.000	С

## Tidyr examples

separate source extract source

```
(def df-separate (api/dataset {:x [nil "a.b" "a.d" "b.c"]}))
(def df-separate2 (api/dataset {:x ["a" "a b" nil "a b c"]}))
(def df-separate3 (api/dataset {:x ["a?b" nil "a.b" "b:c"]}))
(def df-extract (api/dataset {:x [nil "a-b" "a-d" "b-c" "d-e"]}))

#'user/df-separate
#'user/df-separate2
#'user/df-separate3
#'user/df-extract
df-separate
```

\_unnamed [4 1]:

 $\begin{array}{c}
\underline{\mathbf{a}} \\
\underline{\mathbf{a}} \\
\underline{\mathbf{b}} \\
\underline{\mathbf{b}} \\
\underline{\mathbf{c}}
\end{array}$ 

## df-separate2

\_unnamed [4 1]:

:x a a b a b c

## ${\tt df\text{-}separate3}$

 $\underline{\phantom{a}}$ unnamed [4 1]: :xa?b a.b b:c df-extract  $\underline{\phantom{a}}$ unnamed [5 1]: :x a-b a-d b-c  $\operatorname{d-e}$ (api/separate-column df-separate :x [:A :B] "\\.")  $\underline{\quad}$  unnamed [4 2]: :В b d  $\mathbf{c}$ You can drop columns after separation by setting nil as a name. We need second value here. (api/separate-column df-separate :x [nil :B] "\\.")  $\underline{\quad}$  unnamed [4 1]: :В b d Extra data is dropped (api/separate-column df-separate2 :x ["a" "b"] " ")  $\underline{\phantom{a}}$ unnamed [4 2]: b

```
Split with regular expression
(api/separate-column df-separate3 :x ["a" "b"] "[?\\.:]")
\underline{\phantom{a}}unnamed [4 2]:
                                                         b
                                                         b
                                                     a
Or just regular expression to extract values
(api/separate-column df-separate3 :x ["a" "b"] #"(.).(.)")
\underline{\phantom{a}}unnamed [4 2]:
                                                         b
                                                         b
Extract first value only
(api/separate-column df-extract :x ["A"] "-")
_unnamed [5 1]:
                                                       a
                                                       \mathbf{a}
                                                       b
                                                       d
Split with regex
(api/separate-column df-extract :x ["A" "B"] #"(\p{Alnum})-(\p{Alnum})")
\underline{\phantom{a}}unnamed [5 2]:
```

b

a b

a b a d b c
<u>d</u> <u>e</u>
Only a,b,c,d strings
(api/separate-column df-extract :x ["A" "B"] #"([a-d]+)-([a-d]+)") _unnamed [5 2]:
<u>A B</u>
$\begin{array}{ccc} a & b \\ a & d \\ b & c \end{array}$
b с ——
Fold/Unroll Rows
Fold
Unroll
Reshape
Longer
Wider
Join/Concat
Left
Right
Inner
Hash
Concat

 $\overline{A}$  B

# Functions