# Dataset manipulation API for tech.ml.dataset library

### GenerateMe

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#### tech-ml-version

"5.02"

#### 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.

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

- tech.ml.dataset walkthrough
- dtype-next overview
- dtype-next cheatsheet

### SOURCE CODE

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	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$^{\mathrm{C}}$
1	7	0.5	Α
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

# **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: [string column-data] or [keyword column-data]
- array of native arrays
- 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
  - :column-names in case you want to name columns only works for sequential input (arrays)
  - :layout for numberical, native array of arrays treat entries :as-rows or :as-columns (default)

api/let-dataset accepts bindings symbol-column-data to simulate R's tibble function. Each binding is converted into a column. You can refer previous columns to in further bindings (as in let).

Empty dataset.		
(api/dataset)		
_unnamed [0 0]		

Dataset from single value.

```
(api/dataset 999)
```

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

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

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

 $\underline{\phantom{a}}$ 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]])
```

 $\underline{\quad}$  unnamed [1 3]:

:A :B :C 33 5 :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	X	:a
2	X	:a
3	X	:a
4	X	:a
5	X	:a
6	X	:a

Dataset created from map (keys = column names, vals = 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]:

 $\frac{\overline{A}}{33}$ 

 $\underline{\quad}$  unnamed [3 1]:

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

:A	:В
3	X
4	X
5	X

You can put any value inside a column

```
(api/dataset {:A [[3 4 5] [:a :b]] :B "X"})
```

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

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]:

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

:a	:b
1	[1 2 3]
2	$[3 \ 4]$

Missing values are marked by nil

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

\_unnamed [3 2]:

 $\begin{array}{ccc} & :& & :& \\ \hline & & & 1 \\ 3 & & 4 \\ 11 & & & \\ \end{array}$ 

Reading from arrays, by default :as-columns

```
(-> (map int-array [[1 2] [3 4] [5 6]])
     (into-array)
     (api/dataset))
```

:\_unnamed [2 3]:

```
\begin{array}{cccc}
0 & 1 & 2 \\
\hline
1 & 3 & 5 \\
2 & 4 & 6
\end{array}
```

:as-rows

:\_unnamed [3 2]:

```
    \begin{array}{c|c}
      0 & 1 \\
      \hline
      1 & 2 \\
      3 & 4 \\
      5 & 6
    \end{array}
```

:as-rows with names

:\_unnamed [3 2]:

Create dataset using macro let-dataset to simulate R tibble function. Each binding is converted into a column.

 $\underline{\phantom{a}}$  unnamed [5 3]:

:x	<b>:</b> y	:z
1	1	2
2	1	3
3	1	4
4	1	5
5	1	6

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	0.0	12.8	5.0	4.7	drizzle
2012-01-02	10.9	10.6	2.8	4.5	rain
2012-01-03	0.8	11.7	7.2	2.3	$_{ m rain}$

date	precipitation	$temp\_max$	$temp\_min$	wind	weather
2012-01-04	20.3	12.2	5.6	4.7	rain
2012-01-05	1.3	8.9	2.8	6.1	rain
2012-01-06	2.5	4.4	2.2	2.2	rain
2012 - 01 - 07	0.0	7.2	2.8	2.3	rain
2012-01-08	0.0	10.0	2.8	2.0	sun
2012-01-09	4.3	9.4	5.0	3.4	rain
2012-01-10	1.0	6.1	0.6	3.4	rain
2012-01-11	0.0	6.1	-1.1	5.1	sun
2012 - 01 - 12	0.0	6.1	-1.7	1.9	sun
2012 - 01 - 13	0.0	5.0	-2.8	1.3	sun
2012 - 01 - 14	4.1	4.4	0.6	5.3	snow
2012 - 01 - 15	5.3	1.1	-3.3	3.2	snow
2012-01-16	2.5	1.7	-2.8	5.0	snow
2012 - 01 - 17	8.1	3.3	0.0	5.6	snow
2012-01-18	19.8	0.0	-2.8	5.0	snow
2012 - 01 - 19	15.2	-1.1	-2.8	1.6	snow
2012-01-20	13.5	7.2	-1.1	2.3	snow
2012 - 01 - 21	3.0	8.3	3.3	8.2	rain
2012 - 01 - 22	6.1	6.7	2.2	4.8	rain
2012-01-23	0.0	8.3	1.1	3.6	rain
2012-01-24	8.6	10.0	2.2	5.1	rain
2012-01-25	8.1	8.9	4.4	5.4	rain

# Saving

Export dataset to a file or output stream can be done by calling api/write!. 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! ds "output.tsv.gz")
(.exists (clojure.java.io/file "output.tsv.gz"))
```

nil true

### Nippy

```
(api/write! DS "output.nippy.gz")
```

nil

```
(api/dataset "output.nippy.gz")
```

output.nippy.gz [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В

:V1	:V2	:V3	:V4
2	6	1.5	С
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

#### 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 12]:

:col-		:n-	:n-				:standard-			
name	:datatype	valid	missin	g:min	:mean	:mode :max	deviation	:skew	:first	:last
date	:packed-	1461	0	2012-	2013-	2015-	3.64520463	E <b>4.80</b> 606	880240-12-	2015-
	local-			01-01	12-31	12-31		15	01-01	12-31
	date									
precipit	tat <b>ibo</b> at64	1461	0	0.000	3.029	55.90	6.68019432	E <b>3.60</b> 564	372 <b>E00</b> 00	0.000
temp_1	maxloat64	1461	0	-	16.44	35.60	7.34975810	E <b>2.80</b> 929	992 <b>E2</b> -80	5.600
				1.600				01		
temp_r	mirfloat64	1461	0	-	8.235	18.30	5.02300418	E+00	5.000	-
				7.100				2.49458	552E-	2.100
								01		

:col-		:n-	:n-					:standard-			
name	:datatype	valid	missing	:min	:mean	:mode	:max	deviation	:skew	:first	:last
weather	:string	1461	0			rain				drizzle	sun
wind	:float64	1461	0	0.4000	3.241		9.500	1.437825061	E <b>8.90</b> 166751	L9 <b>1</b> E700	3.500
									01		

https://vega.github.io/vega-lite/examples/data/seattle-weather.csv :basic info [1 4]:

: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	:datatype	$: \! n\text{-}elems$	:categorical?
date	:packed-local-date	1461	
precipitation	:float64	1461	
temp_max	:float64	1461	
temp_min	:float64	1461	
wind	:float64	1461	
weather	:string	1461	true

Getting a dataset name

(api/dataset-name ds)

 $\verb|"https://vega.github.io/vega-lite/examples/data/seattle-weather.csv"|$ 

\_\_\_\_

Setting a dataset name (operation is immutable).

"seattle-weather"

### 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.

Possible result types:

- :as-seq or :as-seqs sequence of seqences (default)
- :as-maps sequence of maps (rows)
- :as-map map of sequences (columns)
- :as-double-arrays array of double arrays

Select column.

```
(ds "wind")
 (api/column ds "date")
#tech.v3.dataset.column<float64>[1461]
wind
[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, 3.200, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400, 3.400,
#tech.v3.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.v3.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, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 2012-01-08, 201
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 0x517777d9 "2012-01-01"] 0.0 12.8 5.0 4.7 "drizzle"] [#object[java.time.L
Select rows/columns as double-double-array
(-> ds
                       (api/select-columns :type/numerical)
                        (api/head)
                       (api/rows :as-double-arrays))
#object["[[D" 0x353b9210 "[[D@353b9210"]
(-> ds
                       (api/select-columns :type/numerical)
                        (api/head)
                       (api/columns :as-double-arrays))
#object["[[D" 0x30d05934 "[[D@30d05934"]
Rows as sequence of maps
(clojure.pprint/pprint (take 2 (api/rows ds :as-maps)))
```

```
({"date" #object[java.time.LocalDate 0x39ef9139 "2012-01-01"],
   "precipitation" 0.0,
   "temp_min" 5.0,
   "weather" "drizzle",
   "temp_max" 12.8,
   "wind" 4.7}
{"date" #object[java.time.LocalDate 0x6f1a0e2e "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.

```
(api/print-dataset (api/group-by DS :V1) {:print-line-policy :markdown})
```

\_unnamed [2 3]:

(api/print-dataset (api/group-by DS :V1) {:print-line-policy :repl})

\_unnamed [2 3]:

```
:name | :group-id |
                                                                        :data |
                   0 | Group: 1 [5 4]:
    1 |
       1
                     I \setminus I
                             :V1 \|
                                          : V2 \|
                                                         : \\ \|
                     | \|----\|----\||
                                 1 \|
                                             1 \|
                                                         0.5 \|
                     I \setminus I
                                                                        A \| |
                                             3 \|
                     I \setminus I
                                 1 \|
                                                         1.5 \|
                                                                        C \ | |
                                                                        B \| |
                     I \setminus I
                                 1 \|
                                             5 \|
                                                         1.0 \|
                     I \setminus I
                                 1 \|
                                             7\|
                                                         0.5 \|
                                                                        A \| |
                     I \setminus I
                                 1 \|
                                             9 \|
                                                         1.5 \|
                                                                        C \ | |
                 1 | Group: 2 [4 4]:
    2 |
                     I \setminus I
                               :V1 \|
                                           :V2 \|
                                                         : \\3 \|
                                                                      :V4 \| |
                     | \|----\|----\|----
                                                        ----\|--
                                                                   ----\| |
                     | \ | \ |
                                 2 \|
                                             2 \|
                                                         1.0 \|
                                                                        B \setminus I
                     1 \1
                                 2 \ [
                                             4 \|
                                                         0.5 \|
                                                                        A \ | |
                     I \setminus I
                                 2 \|
                                             6 \1
                                                         1.5 \|
                                                                        C \| |
                                                         1.0 \|
                     I \setminus I
                                 2 \|
                                             8 \1
                                                                        B \ | |
```

(api/print-dataset (api/group-by DS :V1) {:print-line-policy :single})

\_unnamed [2 3]:

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

#### 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
  - :select-keys list of the columns passed to a grouping selector 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 (limited to :select-keys)

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 grouped dataset

List of columns in grouped dataset treated as regular dataset

```
(-> DS
    (api/group-by :V1)
    (api/as-regular-dataset)
    (api/column-names))
(:name :group-id :data)
Content of the grouped dataset
(api/columns (api/group-by DS :V1) :as-map)
{:name #tech.v3.dataset.column<int64>[2]
:name
[1, 2], :group-id #tech.v3.dataset.column<int64>[2]
:group-id
[0, 1], :data #tech.v3.dataset.column<dataset>[2]
:data
[Group: 1 [5 4]:
                          :V3 |
     :V1 |
              : 72 |
                                    : 74
       1 |
                1 |
                         0.5 |
                                      Αl
       1 |
                3 |
                         1.5
                                      Cl
       1 |
                5 |
                         1.0 |
                                      ΒΙ
       1 |
                7 |
                         0.5
                                      Αl
                         1.5 |
       1 |
                9 |
                                      Cl
, Group: 2 [4 4]:
     :V1 |
              :V2 |
                          :V3 |
                                    :V4 |
Ι
       2 |
                2 |
                         1.0 |
                                      ΒΙ
       2 |
                4 |
                         0.5 |
                                      A |
       2 |
                                      Cl
6 |
                         1.5 |
       2 |
                8 |
                         1.0 |
                                      ΒΙ
]}
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
                                               0.5
                                                     Α
                                          3
                                               1.5
                                                     \mathbf{C}
                                    1
                                                   В
                                    1
                                          5
                                               1.0
                                     1
                                          7
                                               0.5
                                                    Α
```

1.5

 $\mathbf{C}$ 

9

# Group: 2 [4 4]:

:V1	:V2	:V3	:V4
2	2	1.0	В
2	4	0.5	A
2	6	1.5	$\mathbf{C}$
2	8	1.0	В

)

Group dataset as map of indexes (row ids)

```
(api/group-by DS :V1 {:result-type :as-indexes})
{1 #list<int32>[5]
[0, 2, 4, 6, 8], 2 #list<int32>[4]
[1, 3, 5, 7]}
```

Grouped datasets are printed as follows by default.

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

 $\underline{\phantom{a}}$ 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 groups->seq and groups->map 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]:

```
d
(-> {"a" [1 1 2 2]
     "b" ["a" "b" "c" "d"]}
     (api/dataset)
     (api/group-by "a")
     (api/groups->seq))
(Group: 1 [2 2]:
                                                   b
                                                  a
                                               1 b
Group: 2 [2 2]:
                                                  b
                                               \frac{}{2} c
                                                  d
)
Groups as map
(-> {"a" [1 1 2 2]
      "b" ["a" "b" "c" "d"]}
     (api/dataset)
     (api/group-by "a")
     (api/groups->map))
{1 Group: 1 [2 2]:
                                                  b
                                               1
                                                  \mathbf{a}
                                                  b
, 2 Group: 2 [2 2]:
                                                   b
                                               2 c
                                               2 d
}
```

 $\frac{\mathrm{b}}{\mathrm{c}}$ 

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	1.0	В

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

:V1	:V2	:V3	:V4
1	1	0.5	A
1	7	0.5	A

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

:V1	:V2	:V3	:V4
2	4	0.5	A

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

:V1	:V2	:V3	:V4
2	2	1.0	В
2	8	1.0	В

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

:V1	:V2	:V3	:V4
1	3	1.5	С
1	9	1.5	$\mathbf{C}$

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

:V1	:V2	:V3	:V4
2	6	1.5	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	1.0	В
1	3	1.5	$\mathbf{C}$
2	2	1.0	В
1	3	1.5	$\mathbf{C}$

Group: group-b [4 4]:

:V1	:V2	:V3	:V4
2	6	1.5	$\overline{C}$
2	6	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
2	2	1.0	В

)

You can group by a result of grouping 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	0.5	A
1	5	1.0	В

Group: 2.0 [2 4]:

:V1	:V2	:V3	:V4
2	2	1.0	В
2	8	1.0	В

Group: 0.5 [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
1	7	0.5	A

Group: 3.0 [1 4]:

:V1	:V2	:V3	:V4
2	6	1.5	С

Group: 1.5 [2 4]:

:V1	:V2	:V3	:V4
1	3	1.5	С
1	9	1.5	$\mathbf{C}$

)

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	1.0	В
1	3	1.5	$\mathbf{C}$
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

Group: true [3 4]:

:V2	:V3	:V4
1	0.5	A
4	0.5	A
7	0.5	A
	1 4	1 0.5 4 0.5

)

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	1.0	В

Group: [1 0.5] [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
1	7	0.5	A

Group: [2 1.5] [1 4]:

:V1	:V2	:V3	:V4
2	6	1.5	С

Group: [1 1.5] [2 4]:

:V1	:V2	:V3	:V4
1	3	1.5	$\overline{C}$
1	9	1.5	$\mathbf{C}$

Group: [2 0.5] [1 4]:

:V1	:V2	:V3	:V4
2	4	0.5	A

Group: [2 1.0] [2 4]:

:V1	:V2	:V3	:V4
2	2	1.0	В
2	8	1.0	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	0.5	A
1	3	1.5	$\mathbf{C}$
1	5	1.0	В
1	7	0.5	A
1	9	1.5	$\mathbf{C}$

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

:V1	:V2	:V3	:V4
2	2	1.0	В
2	4	0.5	A
2	6	1.5	$\mathbf{C}$
2	8	1.0	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))
```

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

:V1	:V2	:V3	:V4
2	2	1.0	В
1	5	1.0	В
2	8	1.0	В
1	1	0.5	A
2	4	0.5	A
1	7	0.5	A
1	3	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$

Groups sorted by group name and named.

Ordered by V3 [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	4	0.5	A
1	7	0.5	A
2	2	1.0	В
1	5	1.0	В
2	8	1.0	В
1	3	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$

Groups sorted descending by group name and named.

Ordered by V3 descending [9 4]:

:V1	:V2	:V3	:V4
1	3	1.5	$^{\rm C}$
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$
2	2	1.0	В
1	5	1.0	В
2	8	1.0	В
1	1	0.5	A
2	4	0.5	A
1	7	0.5	A

Let's add group name and id as additional columns

:group-name :group-id	:V1	:V2	:V3	:V4	
false	0	2	4	0.5	A
false	0	1	5	1.0	В

$\overline{:group-name :group-id}$	:V1	:V2	:V3	:V4	
false	0	2	6	1.5	$\overline{C}$
false	0	1	7	0.5	A
false	0	2	8	1.0	В
false	0	1	9	1.5	$\mathbf{C}$
true	1	1	1	0.5	A
true	1	2	2	1.0	В
true	1	1	3	1.5	$\mathbf{C}$

Let's assign different column names

\_unnamed [9 6]:

Is V2 less than 4?	group id	:V1	:V2	:V3	:V4
false	0	2	4	0.5	A
false	0	1	5	1.0	В
false	0	2	6	1.5	$\mathbf{C}$
false	0	1	7	0.5	A
false	0	2	8	1.0	В
false	0	1	9	1.5	$\mathbf{C}$
true	1	1	1	0.5	A
true	1	2	2	1.0	В
true	1	1	3	1.5	$\mathbf{C}$

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

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

We can add group names without separation

# \_unnamed [9 5]:

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

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	1.0	1	5	1.0	В
1	0.5	1	1	0.5	A
1	0.5	1	7	0.5	A
2	1.5	$^2$	6	1.5	$\mathbf{C}$
1	1.5	1	3	1.5	$\mathbf{C}$
1	1.5	1	9	1.5	$\mathbf{C}$
2	0.5	$^2$	4	0.5	A
2	1.0	$^2$	2	1.0	В
2	1.0	2	8	1.0	В

Let's provide column names

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

$\overline{v1}$	v3	:V1	:V2	:V3	:V4
1	1.0	1	5	1.0	В
1	0.5	1	1	0.5	A
1	0.5	1	7	0.5	A
2	1.5	2	6	1.5	$\mathbf{C}$
1	1.5	1	3	1.5	$\mathbf{C}$
1	1.5	1	9	1.5	$\mathbf{C}$
2	0.5	2	4	0.5	A
2	1.0	2	$^2$	1.0	В
2	1.0	2	8	1.0	В

Also we can supress separation

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

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

You can also operate on grouped dataset as a regular one in case you want to access its columns using without-grouping-> threading macro.

```
(-> DS
    (api/group-by [:V4 :V1])
    (api/without-grouping->
        (api/order-by (comp (juxt :V4 :V1) :name))))
```

 $\underline{\phantom{a}}$  unnamed [6 3]:

:name	:group-id	:data
{:V4 "A", :V1 1}	1	Group: {:V4 "A", :V1 1} [2 4]:
{:V4 "A", :V1 2}	4	Group: {:V4 "A", :V1 2} [1 4]:
{:V4 "B", :V1 1}	2	Group: {:V4 "B", :V1 1} [1 4]:
{:V4 "B", :V1 2}	5	Group: {:V4 "B", :V1 2} [2 4]:
{:V4 "C", :V1 1}	3	Group: {:V4 "C", :V1 1} [2 4]:
{:V4 "C", :V1 2}	0	Group: {:V4 "C", :V1 2} [1 4]:

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))
```

\_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
- $\bullet\,\,$  column name for single column
- $\bullet\,$  sequence of column names for collection of columns

- regex to apply pattern on column names or datatype
- filter predicate to filter column names or datatype
- type namespaced keyword for specific datatype or group of datatypes

Column name can be anything.

column-names function returns names according to columns-selector and optional meta-field. 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

Datatype groups are:

(:V1 :V2 :V3 :V4)

()

- :type/numerical any numerical type
- :type/float floating point number (:float32 and :float64)
- :type/integer any integer
- :type/datetime any datetime type

If qualified keyword starts with :!type, complement set is used.

\_\_\_\_

To select all column names you can use column-names function.

```
(api/column-names DS)

(:V1 :V2 :V3 :V4)
or
(api/column-names DS :all)
```

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)
(api/column-names DS :type/integer)
(:V1:V2)
And finally we can use predicate to select names. Let's select double precision columns.
(api/column-names DS #{:float64} :datatype)
(:V3)
or
(api/column-names DS :type/float64)
(: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)
(api/column-names DS :!type/float64)
(:V2:V3:V4)
(:V1 :V2 :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 float 64 columns

```
(api/select-columns DS #(= :float64 %) :datatype)
_unnamed [9 1]:
                                                  :V3
                                                  0.5
                                                  1.0
                                                  1.5
                                                  0.5
                                                  1.0
                                                  1.5
                                                  0.5
                                                  1.0
                                                  1.5
or
(api/select-columns DS :type/float64)
_unnamed [9 1]:
                                                  :V3
                                                  0.5
                                                  1.0
                                                  1.5
                                                  0.5
                                                  1.0
                                                  1.5
                                                  0.5
                                                  1.0
                                                  1.5
Select all but :V1 columns
(api/select-columns DS (complement #{:V1}))
_unnamed [9 3]:
                                            :V2
                                                 :V3
                                                        :V4
                                            1
                                                  0.5
                                                        A
                                            2
                                                  1.0
                                                        В
                                            3
                                                  1.5
                                                        \mathbf{C}
                                                  0.5
                                            4
                                                        Α
                                            5
                                                  1.0
                                                        В
                                            6
                                                  1.5
                                                        \mathbf{C}
                                            7
                                                  0.5
                                                        Α
                                                        В
                                                  1.0
                                                        \mathbf{C}
                                                  1.5
```

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	0.5
3	1.5
5	1.0
7	0.5
9	1.5

, 2 Group: 2 [4 2]:

:V2	:V3
2	1.0
4	0.5
6	1.5
8	1.0

}

# Drop

 ${\tt drop\text{-}columns}$  creates dataset with removed columns.

Drop float64 columns

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

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

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

or

# (api/drop-columns DS :type/float64)

\_unnamed [9 3]:

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

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	С

, 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. You can also pass mapping function with optional columns-selector

\_unnamed [9 4]:

v1	v2	$[1 \ 2 \ 3]$	java.lang. Object@e59 faef
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	С

Map all names with function

```
(api/rename-columns DS (comp str second name))
```

1	2	3	4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	С
2	4	0.5	A
1	5	1.0	В
2	6	1.5	С
1	7	0.5	A
2	8	1.0	В
1	9	1.5	С

Map selected names with function

```
(api/rename-columns DS [:V1 :V3] (comp str second name))
```

\_unnamed [9 4]:

1	:V2	3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

Function works on grouped dataset

 $\{1 \text{ Group: } 1 \text{ [5 4]:}$ 

v1	v2	[1 2 3]	java.lang.Object@7981e710
1	1	0.5	A
1	3	1.5	$\mathbf{C}$
1	5	1.0	В
1	7	0.5	A
1	9	1.5	C

, 2 Group: 2 [4 4]:

v1	v2	[1 2 3]	java.lang.Object@7981e710
2	2	1.0	В
2	4	0.5	A
2	6	1.5	С
2	8	1.0	В

}

# Add or update

To add (or replace existing) column call add-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
- :strict throws an exception when sizes mismatch

Function works on grouped dataset.

Add single value as column

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

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

:V1	:V2	:V3	:V4	:V5
1	1	0.5	A	X
2	2	1.0	В	X
1	3	1.5	$\mathbf{C}$	X
2	4	0.5	$\mathbf{A}$	X
1	5	1.0	В	X
2	6	1.5	$\mathbf{C}$	X
1	7	0.5	$\mathbf{A}$	X
2	8	1.0	В	X
1	9	1.5	$\mathbf{C}$	X

Replace one column (column is trimmed)

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

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
0.1383	1	0.5	A
0.2082	2	1.0	В
0.3080	3	1.5	$\mathbf{C}$
0.4503	4	0.5	A
0.9489	5	1.0	В
0.1236	6	1.5	$\mathbf{C}$
0.1189	7	0.5	A
0.4253	8	1.0	В
0.3583	9	1.5	$\mathbf{C}$

Copy column

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

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:V5
1	1	0.5	A	1
2	2	1.0	В	2
1	3	1.5	$\mathbf{C}$	1
2	4	0.5	A	2
1	5	1.0	В	1
2	6	1.5	$\mathbf{C}$	2
1	7	0.5	A	1
2	8	1.0	В	2
1	9	1.5	$\mathbf{C}$	1

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

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:row-count
1	1	0.5	A	9
2	2	1.0	В	9
1	3	1.5	$\mathbf{C}$	9
2	4	0.5	A	9
1	5	1.0	В	9
2	6	1.5	$\mathbf{C}$	9
1	7	0.5	$\mathbf{A}$	9
2	8	1.0	В	9
1	9	1.5	$\mathbf{C}$	9

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

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

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

:V1	:V2	:V3	:V4	:row-count
1	1	0.5	A	5
1	3	1.5	$\mathbf{C}$	5
1	5	1.0	В	5
1	7	0.5	A	5
1	9	1.5	$\mathbf{C}$	5
2	2	1.0	В	4
2	4	0.5	A	4
2	6	1.5	$\mathbf{C}$	4

:V1	:V2	:V3	:V4	:row-count
2	8	1.0	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-column DS : V5 [:r :b])
```

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

:V1	:V2	:V3	:V4	:V5
1	1	0.5	A	:r
2	2	1.0	В	:b
1	3	1.5	$\mathbf{C}$	:r
2	4	0.5	A	:b
1	5	1.0	В	:r
2	6	1.5	$\mathbf{C}$	:b
1	7	0.5	A	:r
2	8	1.0	В	:b
1	9	1.5	С	:r

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

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:V5
1	1	0.5	A	:r
2	2	1.0	В	:b
1	3	1.5	$\mathbf{C}$	
2	4	0.5	A	
1	5	1.0	В	
2	6	1.5	$\mathbf{C}$	
1	7	0.5	A	
2	8	1.0	В	
1	9	1.5	С	

Exception is thrown when :strict strategy is used and column size is not equal row count

```
(try
  (api/add-column DS :V5 [:r :b] :strict)
  (catch Exception e (str "Exception caught: "(ex-message e))))
```

"Exception caught: Column size (2) should be exactly the same as dataset row count (9)"

The same applies for grouped dataset

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

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

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:V5
2	2	1.0	В	:r
1	5	1.0	В	:b
2	8	1.0	В	
1	1	0.5	A	:r
2	4	0.5	A	:b
1	7	0.5	A	
1	3	1.5	$\mathbf{C}$	:r
2	6	1.5	С	:b
1	9	1.5	С	

Let's use other column to fill groups

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

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

:V1	:V2	:V3	:V4	:V5
2	2	1.0	В	1
1	5	1.0	В	2
2	8	1.0	В	3
1	1	0.5	A	1
2	4	0.5	A	2
1	7	0.5	A	3
1	3	1.5	$\mathbf{C}$	1
2	6	1.5	С	2
1	9	1.5	С	3

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

```
(api/add-columns DS {:V1 #(map inc (% :V1))

:V5 #(map (comp keyword str) (% :V4))

:V6 11})
```

:V1	:V2	:V3	:V4	:V5	:V6
2	1	0.5	A	:A	11
3	2	1.0	В	:В	11
2	3	1.5	$\mathbf{C}$	:C	11
3	4	0.5	A	:A	11
2	5	1.0	В	:В	11

:V1	:V2	:V3	:V4	:V5	:V6
3	6	1.5	С	:C	11
2	7	0.5	A	:A	11
3	8	1.0	В	:В	11
2	9	1.5	$\mathbf{C}$	:С	11

## Update

If you want to modify specific column(s) you can call update-columns. Arguments:

- dataset
- one of:
  - columns-selector and function (or sequence of functions)
  - map where keys are column names and vals are function

Functions accept column and have to return column or sequence

Reverse of columns

```
(api/update-columns DS :all reverse)
```

\_unnamed [9 4]:

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

Apply dec/inc on numerical columns

:V1	:V2	:V3	:V4
0	2	-0.5000	A
1	3	0.000	В
0	4	0.5000	$\mathbf{C}$
1	5	-0.5000	A
0	6	0.000	В
1	7	0.5000	$\mathbf{C}$
0	8	-0.5000	A
1	9	0.000	В
0	10	0.5000	С

You can also assing function to a column by packing operations into the map.

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	3	0.5	A
2	2	1.0	В
1	6	1.5	$\mathbf{C}$
2	8	0.5	A
1	9	1.0	В
2	7	1.5	$\mathbf{C}$
1	5	0.5	A
2	1	1.0	В
1	4	1.5	$\mathbf{C}$

### Map

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

## Arguments:

- ds dataset
- $\bullet\,$  column-name target column name
- $\bullet$  columns-selector columns selected
- map-fn mapping function

Let's add numerical columns together

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:sum-of-numbers
1	1	0.5	A	2.5
2	2	1.0	В	5.0
1	3	1.5	$\mathbf{C}$	5.5
2	4	0.5	A	6.5
1	5	1.0	В	7.0
2	6	1.5	$\mathbf{C}$	9.5
1	7	0.5	A	8.5
2	8	1.0	В	11.0
1	9	1.5	$\mathbf{C}$	11.5

The same works on grouped dataset

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

:V1	:V2	:V3	:V4	:sum-of-numbers
1	1	0.5	A	2.5
2	4	0.5	A	6.5
1	7	0.5	A	8.5
2	2	1.0	В	5.0
1	5	1.0	В	7.0
2	8	1.0	В	11.0
1	3	1.5	$\mathbf{C}$	5.5
2	6	1.5	С	9.5
1	9	1.5	$\mathbf{C}$	11.5

## 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])
```

\_unnamed [9 4]:

:V4	:V3	:V2	:V1
A	0.5	1	1
В	1.0	2	2
$\mathbf{C}$	1.5	3	1
A	0.5	4	2
В	1.0	5	1
$\mathbf{C}$	1.5	6	2
A	0.5	7	1
В	1.0	8	2
$\mathbf{C}$	1.5	9	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))
```

:V3	:V4	:V1	:V2
0.5	A	1	1
1.0	В	2	2

:V3	:V4	:V1	:V2
1.5	С	1	3
0.5	A	2	4
1.0	В	1	5
1.5	С	2	6
0.5	A	1	7
1.0	В	2	8
1.5	$\mathbf{C}$	1	9

#### Type conversion

To convert column into given datatype can be done using convert-types 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
  - column-selector and new-types column name and new datatype (or datatypes as sequence)

#### new-types can be:

- a type like :int64 or :string or sequence of types
- or sequence of pair of datetype 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-types :V1 :float64)
     (api/info :columns))
```

\_unnamed :column info [4 6]:

:name	:datatype	:n-elems	:unparsed-indexes	:unparsed-data	:categorical?
:V1	:float64	9	{}		_
:V2	:int $64$	9			
:V3	:float64	9			
:V4	:string	9			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-types :V4 [[:int16 #(Integer/parseInt % 16)]]))
```

:V1	:V2	:V3	:V4
1	1	0.5	10
2	2	1.0	11
1	3	1.5	12
2	4	0.5	10
1	5	1.0	11
2	6	1.5	12
1	7	0.5	10
2	8	1.0	11
1	9	1.5	12

You can process several columns at once

\_unnamed :column info [4 5]:

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

Convert one type into another

```
(-> DS
     (api/convert-types :type/numerical :int16)
     (api/info :columns))
```

\_unnamed :column info [4 6]:

:name	:datatype	:n-elems	:unparsed-indexes	:unparsed-data	:categorical?
:V1	:int16	9	{}		
:V2	:int16	9	$\{\}$		
:V3	:int 16	9	{}		
:V4	:string	9			true

Function works on the grouped dataset

```
(-> DS
    (api/group-by :V1)
    (api/convert-types :V1 :float32)
    (api/ungroup)
```

```
(api/info :columns))
```

\_unnamed :column info [4 6]:

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

Double array conversion.

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

```
#object["[J" 0x619ce45e "[J@619ce45e"]
```

Function also works on grouped dataset

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

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;" 0x4e1b666f "[Ljava.lang.String;@4e1b666f"]
#object["[F" 0x551f6c24 "[F@551f6c24"]
```

#### 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 (select-keys 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-columns on a dataset. :pre is used as a column definitions.

#### Select

Select fifth row

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

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

:V1	:V2	:V3	:V4
1	5	1.0	В

Select 3 rows

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

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

:V1	:V2	:V3	:V4
2	2	1.0	В
1	5	1.0	В
2	6	1.5	$\mathbf{C}$

Select rows using sequence of true/false values

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

\_unnamed [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	4	0.5	A

Select rows using predicate

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

\_unnamed [3 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	4	0.5	A
1	7	0.5	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	0.5	A
2	2	1.0	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	0.5	A
2	4	0.5	A
2	2	1.0	В
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$

### 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.

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

:V1	:V2	:V3	:V4
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

### Other

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

All random functions accept :seed as an option if you want to fix returned result.

First row

(api/first DS)

\_unnamed [1 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A

Last row

(api/last DS)

\_unnamed [1 4]:

Random row (single)

(api/rand-nth DS)

\_unnamed [1 4]:

:V1	:V2	:V3	:V4
2	2	1.0	В

Random row (single) with seed

(api/rand-nth DS {:seed 42})

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

Random  ${\tt n}$  (default: row count) rows with repetition.

(api/random DS)

:V1	:V2	:V3	:V4
1	5	1.0	В
1	7	0.5	A
1	7	0.5	A
2	8	1.0	В

:V1	:V2	:V3	:V4
1	7	0.5	A
2	8	1.0	В
2	2	1.0	В
1	9	1.5	$\mathbf{C}$
1	7	0.5	A

Five random rows with repetition

(api/random DS 5)

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

:V1	:V2	:V3	:V4
2	2	1.0	В
2	2	1.0	В
1	7	0.5	A
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$

Five random, non-repeating rows

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

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

:V1	:V2	:V3	:V4
2	8	1.0	В
2	6	1.5	$^{\mathrm{C}}$
1	1	0.5	A
2	4	0.5	A
1	9	1.5	$\mathbf{C}$

Five random, with seed

(api/random DS 5 {:seed 42})

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

:V1	:V2	:V3	:V4
2	6	1.5	С
1	5	1.0	В
1	3	1.5	$^{\mathrm{C}}$
1	1	0.5	A
1	9	1.5	$\mathbf{C}$

## Shuffle dataset

# (api/shuffle DS)

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

:V1	:V2	:V3	:V4
1	5	1.0	В
2	8	1.0	В
1	3	1.5	$^{\mathrm{C}}$
2	6	1.5	$^{\mathrm{C}}$
1	1	0.5	A
2	4	0.5	A
1	7	0.5	A
1	9	1.5	$\mathbf{C}$
2	2	1.0	В

## Shuffle with seed

# (api/shuffle DS {:seed 42})

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	5	1.0	В
2	2	1.0	В
2	6	1.5	$\mathbf{C}$
2	4	0.5	A
2	8	1.0	В
1	3	1.5	$\mathbf{C}$
1	7	0.5	A
1	1	0.5	A
1	9	1.5	С

# First n rows (default 5)

# (api/head DS)

\_unnamed [5 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$^{\mathrm{C}}$
2	4	0.5	A
1	5	1.0	В

Last n rows (default 5)

## (api/tail DS)

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

:V1	:V2	:V3	:V4
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

by-rank calculates rank on column(s). It's base on R rank() with addition of :dense (default) tie strategy which give consecutive rank numbering.

:desc? options (default: true) sorts input with descending order, giving top values under 0 value.

rank is zero based and is defined at tablecloth.api.utils namespace.

```
(api/by-rank DS : V3 zero?) ;; most V3 values
```

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

:V1	:V2	:V3	:V4
1	3	1.5	С
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$

(api/by-rank DS : V3 zero? {:desc? false}) ;; least V3 values

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

:V1	:V2	:V3	:V4
1	1	0.5	A
2	4	0.5	A
1	7	0.5	A

Rank also works on multiple columns

```
(api/by-rank DS [:V1 :V3] zero? {:desc? false})
```

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

:V1	:V2	:V3	:V4
1	1	0.5	A
1	7	0.5	A

Select 5 random rows from each group

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

\_unnamed [15 4]:

:V1	:V2	:V3	:V4
2	4	0.5	A
1	1	0.5	A
2	4	0.5	A
2	4	0.5	A
1	1	0.5	A
2	8	1.0	В
2	2	1.0	В
2	2	1.0	В
2	8	1.0	В
2	2	1.0	В
2	6	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
1	9	1.5	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 off 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).

Let's calculate mean of some columns

(api/aggregate DS #(reduce + (% :V2)))

\_unnamed [1 1]:

\_summary

45

Let's give resulting column a name.

Sequential result is spread into separate columns

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

 $\underline{\phantom{a}}$ 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

 $\underline{\phantom{a}}$ unnamed [1 5]:

$: \!\! V2\text{-value-}0\text{-}0$	$: \!\! V2\text{-}value\text{-}0\text{-}1$	:V2-value-0-2	: V2-value-1-sum-v1	: V2-value-1-prod-v3
1	2	3	13	0.421875

Processing grouped dataset

\_unnamed [3 6]:

:V4	:V2-value-0-0	:V2-value-0-1	:V2-value-0-2	:V2-value-1-sum-v1	:V2-value-1-prod- v3
В	2	5	8	5	1.000
$\mathbf{C}$	3	6	9	4	3.375
A	1	4	7	4	0.125

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

\_unnamed [3 3]:

:name	:group-id	:data
{:V3 1.0}	0	_unnamed [1 5]:
$\{:V3\ 0.5\}$	1	$\underline{}$ unnamed [1 5]:
$\{:V3\ 1.5\}$	2	$\underline{}$ unnamed [1 5]:

### Column

You can perform columnar aggregation also. aggregate-columns selects columns and apply aggregating function (or sequence of functions) for each column separately.

```
(api/aggregate-columns DS [:V1 :V2 :V3] #(reduce + %))
```

unnamed  $[1 \ 3]$ :

 $\underline{\phantom{a}}$ unnamed [1 3]:

:V1	:V2	:V3
13	9	0.421875

```
(-> DS
    (api/group-by [:V4])
    (api/aggregate-columns [:V1 :V2 :V3] #(reduce + %)))
```

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

:V4	:V1	:V2	:V3
В	5	15	3.0
$\mathbf{C}$	4	18	4.5
A	4	12	1.5

## $\mathbf{Order}$

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

:select-keys 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	0.5	A
1	3	1.5	$\mathbf{C}$
1	5	1.0	В
1	7	0.5	A
1	9	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
2	4	0.5	A
2	8	1.0	В
2	2	1.0	В

Descending order

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

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
2	2	1.0	В
2	4	0.5	A
2	6	1.5	$\mathbf{C}$
2	8	1.0	В
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
1	7	0.5	A
1	1	0.5	A
1	9	1.5	С

Order by two columns

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

:V1	:V2	:V3	:V4
1	1	0.5	A
1	3	1.5	$\mathbf{C}$
1	5	1.0	В
1	7	0.5	A
1	9	1.5	$\mathbf{C}$
2	2	1.0	В
2	4	0.5	A
2	6	1.5	$\mathbf{C}$
2	8	1.0	В

Use different orders for columns

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

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	9	1.5	С
1	7	0.5	A
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
1	1	0.5	A
2	8	1.0	В
2	6	1.5	$\mathbf{C}$
2	4	0.5	A
2	2	1.0	В

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

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
2	8	1.0	В
2	6	1.5	$\mathbf{C}$
2	4	0.5	A
2	$^2$	1.0	В
1	9	1.5	$\mathbf{C}$
1	7	0.5	A
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
1	1	0.5	A

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

:V1	:V2	:V3	:V4
2	4	0.5	A
2	2	1.0	В
2	8	1.0	В
2	6	1.5	$\mathbf{C}$
1	1	0.5	A
1	7	0.5	A
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$

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

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	3	1.5	С
1	9	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
2	2	1.0	В
1	5	1.0	В
2	8	1.0	В
1	1	0.5	A
1	7	0.5	A
2	4	0.5	A

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 z else along x and y. We use first three columns for that.

#'user/dist

:V1	:V2	:V3	:V4
1	1	0.5	A
1	5	1.0	В
1	7	0.5	A
1	9	1.5	$\mathbf{C}$
2	$^2$	1.0	В
2	4	0.5	A
1	3	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
2	8	1.0	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	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	С

Remove duplicates from each group selected by column.

(api/unique-by DS : V1)

\_unnamed [2 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В

Pair of columns

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

\_unnamed [6 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$

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	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$

The same can be achived with group-by

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

:V1	:V2	:V3	:V4
1	3	1.5	С
1	1	0.5	A
2	2	1.0	В

Grouped dataset

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

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

:V1	:V2	:V3	:V4
1	1	0.5	A
2	4	0.5	A
2	2	1.0	В
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$

# ${\bf Strategies}$

There are 4 strategies defined:

- :first select first row (default)
- ullet :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})
```

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

:V1	:V2	:V3	:V4
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

Random

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

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

:V1	:V2	:V3	:V4
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

Pack columns into vector

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

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

:V1	:V2	:V3
[2 1 2]	[2 5 8]	[1.0 1.0 1.0]
[1 2 1]	[3 6 9]	[1.5 1.5 1.5]
[1 2 1]	[1 4 7]	[0.5 0.5 0.5]

Sum columns

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

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

:V1	:V2	:V3
5	15	3.0
4	18	4.5
4	12	1.5

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}))
```

 $\underline{\quad}$  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

```
:V3 (take 9 (cycle [0.5 1.0 nil 1.5]))
:V4 (take 9 (cycle ["A" "B" "C"]))}))
```

DSm

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
	1		
1	1	0.5	A
2	2	1.0	В
	3		$\mathbf{C}$
1	4	1.5	A
2	5	0.5	В
	6	1.0	$\mathbf{C}$
1	7		A
2	8	1.5	В
	9	0.5	$\mathbf{C}$

## Select

Select rows with missing values

(api/select-missing DSm)

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

:V1	:V2	:V3	:V4
	3		$^{\rm C}$
	6	1.0	$\mathbf{C}$
1	7		A
	9	0.5	$\mathbf{C}$

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

(api/select-missing DSm :V1)

\_unnamed [3 4]:

:V1	:V2	:V3	:V4
	3		С
	6	1.0	$\mathbf{C}$
	9	0.5	$\mathbf{C}$

The same with grouped dataset

 $\underline{\phantom{a}}$ 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	0.5	A
2	2	1.0	В
1	4	1.5	A
2	5	0.5	В
2	8	1.5	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	0.5	A
2	2	1.0	В
1	4	1.5	A
2	5	0.5	В
1	7		A
2	8	1.5	В

The same with grouped dataset

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

:V1	:V2	:V3	:V4
1	1	0.5	A
1	4	1.5	A
1	7		A
2	2	1.0	В

:V1	:V2	:V3	:V4
2	5	0.5	В
2	8	1.5	В

#### Replace

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

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

#### Strategies are:

- :value replace with given value
- :up copy values up and then down for missing values at the end
- :down copy values down and then up for missing values at the beginning
- :mid or :nearest copy values around known values
- :midpoint use average value from previous and next non-missing
- :lerp trying to lineary approximate values, works for numbers and datetime, otherwise applies :nearest. For numbers always results in float datatype.

Let's define special dataset here:

unnamed  $[15\ 2]$ :

:a	:1
	2
	2
	2
1.0	
2.0	

Replace missing with default strategy for all columns

# (api/replace-missing DSm2)

\_unnamed [15 2]:

:a	:b
1.0	2
1.0	2
1.0	2
1.0	$^2$
2.0	2
2.0	2
2.0	13
2.0	13
4.0	13
4.0	13
4.0	13
4.0	3
11.0	4
11.0	5
11.0	5

Replace missing with single value in whole dataset

```
(api/replace-missing DSm2 :all :value 999)
```

\_unnamed [15 2]:

:a	:b
999.0	2
999.0	2
999.0	2
1.0	999
2.0	999
999.0	999
999.0	999
999.0	999
999.0	999
999.0	13
4.0	999
999.0	3
11.0	4
999.0	5
999.0	5

Replace missing with single value in :a column

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

\_unnamed [15 2]:

 $:\!\!\mathrm{b}$ :a 2 999.0 999.0 2 999.0 1.0 2.0 999.0999.0999.0 999.0999.0 13 4.0 999.03 11.0 4 999.0 5 999.0 5

Replace missing with sequence in  $: \mathtt{a}$  column

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

\_unnamed [15 2]:

```
:a
        :b
        2
-999.0
-998.0
        2
-997.0
        2
1.0
2.0
-999.0
-998.0
-997.0
-999.0
-998.0
        13
4.0
-997.0
        3
11.0
        4
-999.0
        5
-998.0
```

Replace missing with a function (mean)

```
(api/replace-missing DSm2 :a :value tech.v3.datatype.functional/mean)
```

\_unnamed [15 2]:

```
    :a
    :b

    4.5
    2

    4.5
    2

    4.5
    2

    4.5
    2
```

:a	:b
1.0	
2.0	
4.5	
4.5	
4.5	
4.5	
4.5	13
4.0	
4.5	3
11.0	4
4.5	5
4.5	5

Using :down strategy, fills gaps with values from above. You can see that if missings are at the beginning, the are filled with first value

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

## \_unnamed [15 2]:

:a	:b
1.0	2
1.0	2
1.0	$^{2}$
1.0	$^{2}$
2.0	2
2.0	2
2.0	$^{2}$
2.0	$^{2}$
2.0	$^{2}$
2.0	13
4.0	13
4.0	3
11.0	4
11.0	5
11.0	5

To fix above issue you can provide value

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

 $\underline{\phantom{a}}$ unnamed [15 2]:

:a	:b
999.0	2
999.0	2
999.0	2
1.0	2
2.0	2

:a	:b
2.0	2
2.0	2
2.0	2
2.0	2
2.0	13
4.0	13
4.0	3
11.0	4
11.0	5
11.0	5

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

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

\_unnamed [15 2]:

:a	:b
1.0	2
1.0	2
1.0	2
1.0	13
2.0	13
4.0	13
4.0	13
4.0	13
4.0	13
4.0	13
4.0	3
11.0	3
11.0	4
11.0	5
11.0	5

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

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

\_unnamed [15 2]:

:a	:b
1.0	2.0
1.0	2.0
1.0	2.0
1.0	7.5
2.0	7.5
3.0	7.5
3.0	7.5
3.0	7.5

:a	:b
3.0	7.5
3.0	13.0
4.0	8.0
7.5	3.0
11.0	4.0
11.0	5.0
11.0	5.0

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

```
(api/replace-missing DSm2 [:a :b] :down tech.v3.datatype.functional/mean)
```

 $\_{\rm unnamed}$  [15 2]:

:a	:b
4.5	2
4.5	2
4.5	2
1.0	2
2.0	2
2.0	2
2.0	2
2.0	2
2.0	2
2.0	13
4.0	13
4.0	3
11.0	4
11.0	5
11.0	5

Lerp tries to apply linear interpolation of the values

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

 $\underline{\phantom{a}}$ unnamed [15 2]:

:a	:b
1.00000000	2.00000000
1.00000000	2.00000000
1.00000000	2.00000000
1.00000000	3.57142857
2.00000000	5.14285714
2.333333333	6.71428571
2.66666667	8.28571429
3.00000000	9.85714286
3.33333333	11.42857143
3.66666667	13.00000000
4.00000000	8.00000000

:a	:b
7.50000000	3.00000000
11.00000000	4.00000000
11.00000000	5.00000000
11.00000000	5.00000000

Lerp works also on dates

\_unnamed [9 1]:

:dt
2020-01-01T11:22:33
2020-02-04T16:04:51.500
2020-03-09T20:47:10
2020-04-13T01:29:28.500
2020-05-17T06:11:47
2020-06-20T10:54:05.500
2020-07-24T15:36:24
2020-08-27T20:18:42.500
2020-10-01T01:01:01

## Inject

When your column contains not continuous data range you can fill up with lacking values. Arguments:

- dataset
- column name
- expected step (max-span, milliseconds in case of datetime column)
- (optional) missing-strategy how to replace missing, default :down (set to nil if none)
- (optional) missing-value optional value for replace missing

:a	:b
1.0	:a
2.0	:b
3.0	:b
4.0	:b
5.0	:b
6.0	:b
7.0	:b

:a	:b
8.0	:b
9.0	:c

### 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 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])
```

\_unnamed [9 2]:

```
:V3
      :joined
0.5
      1-1-A
      2-2-B
1.0
      3-C
1.5
      1-4-A
0.5
      2 - 5 - B
1.0
      6-C
      1-7-A
1.5
      2-8-B
0.5
      9-C
```

Without dropping source columns.

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

unnamed [9 5]:

:V1	:V2	:V3	:V4	:joined
1	1	0.5	A	1-1-A
2	2	1.0	В	2-2-B
	3		С	3-C
1	4	1.5	$\mathbf{A}$	1-4-A
2	5	0.5	В	2-5-B
	6	1.0	$\mathbf{C}$	6-C
1	7		A	1-7-A
2	8	1.5	В	2 - 8 - B
	9	0.5	$\mathbf{C}$	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
0.5	1-1-A
1.0	2-2-B
	NA-3-C
1.5	1-4-A
0.5	2 - 5 - B
1.0	NA-6-C
	1-7-A
1.5	2 - 8 - B
0.5	NA-9-C

We can use custom separator.

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

:V3	:joined
0.5	1/1/A
1.0	2/2/B
	./3/C
1.5	1/4/A
0.5	2/5/B
1.0	./6/C
	1/7/A
1.5	2/8/B
0.5	./9/C

Or even sequence of separators.

\_unnamed [9 2]:

```
:V3
    :joined
0.5
     1-1/A
1.0
     2-2/B
     .-3/C
1.5
     1-4/A
     2-5/B
0.5
     .-6/C
1.0
     1-7/A
1.5
     2-8/B
     .-9/C
0.5
```

The other types of results, map:

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

\_unnamed [9 2]:

:V3	:joined
0.5	{:V1 1, :V2 1, :V4 "A"}
1.0	{:V1 2, :V2 2, :V4 "B"}
	{:V1 nil, :V2 3, :V4 "C"}
1.5	{:V1 1, :V2 4, :V4 "A"}
0.5	{:V1 2, :V2 5, :V4 "B"}
1.0	{:V1 nil, :V2 6, :V4 "C"}
	{:V1 1, :V2 7, :V4 "A"}
1.5	{:V1 2, :V2 8, :V4 "B"}
0.5	{:V1 nil, :V2 9, :V4 "C"}

Sequence

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

:V3	:joined
0.5	(1 1 "A")
1.0	(2 2 "B")
	(nil 3 "C")
1.5	(1 4 "A")
0.5	$(2\ 5\ "B")$
1.0	(nil 6 "C")
	(1 7 "A")
1.5	(2 8  "B")
0.5	(nil 9 "C")

## Custom function, calculate hash

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

\_unnamed [9 2]:

:V3	:joined
0.5	535226087
1.0	1128801549
	-1842240303
1.5	2022347171
0.5	1884312041
1.0	-1555412370
	1640237355
1.5	-967279152
0.5	1128367958

## ${\bf Grouped\ dataset}$

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

## \_unnamed [9 2]:

```
:V3:joined
      1-1-A
0.5
      1-4-A
1.5
      1-7-A
1.0
     2 - 2 - B
0.5
      2 - 5 - B
1.5
      2 - 8 - B
      3-C
      6-C
1.0
0.5
      9-C
```

## Tidyr examples

source

#'user/df

df

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

unnamed  $[4\ 3]$ :

:x	<b>:</b> y	Z
a	b	a_b
$\mathbf{a}$		$a\_NA$
	b	$NA_b$
		NA_NA

\_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 can be nil or :infer if separator returns map
- 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(s) or not (default: true or :all in case of empty target-columns). When set to :all keeps only separation result.
  - :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	0.5	A
2	2	1	0.0	В
1	3	1	0.5	$\mathbf{C}$
2	4	0	0.5	A
1	5	1	0.0	В
2	6	1	0.5	$^{\mathrm{C}}$
1	7	0	0.5	A
2	8	1	0.0	В
1	9	1	0.5	$\mathbf{C}$

Source column can be kept

## \_unnamed [9 6]:

:V1	:V2	:V3	:int-part	:frac-part	:V4
1	1	0.5	0	0.5	A
2	2	1.0	1	0.0	В
1	3	1.5	1	0.5	$\mathbf{C}$
2	4	0.5	0	0.5	A
1	5	1.0	1	0.0	В
2	6	1.5	1	0.5	$\mathbf{C}$
1	7	0.5	0	0.5	A
2	8	1.0	1	0.0	В
1	9	1.5	1	0.5	С

We can treat 0 or 0.0 as missing value

## \_unnamed [9 5]:

:V1	:V2	:int-part	:frac-part	:V4
1	1		0.5	A
2	2	1		В
1	3	1	0.5	$\mathbf{C}$
2	4		0.5	A

:V1	:V2	:int-part	:frac-part	:V4
1	5	1		В
2	6	1	0.5	$\mathbf{C}$
1	7		0.5	A
2	8	1		В
1	9	1	0.5	$\mathbf{C}$

Works on grouped dataset

\_unnamed [9 5]:

:V1	:V2	:int-part	:fract-part	:V4
1	1	0	0.5	A
2	4	0	0.5	A
1	7	0	0.5	A
2	2	1	0.0	В
1	5	1	0.0	В
2	8	1	0.0	В
1	3	1	0.5	$\mathbf{C}$
2	6	1	0.5	$\mathbf{C}$
1	9	1	0.5	$\mathbf{C}$

Separate using separator returning sequence of maps, in this case we drop all other columns.

\_unnamed [9 2]:

:int-part	:fract-part
0	0.5
1	0.0
1	0.5
0	0.5
1	0.0
1	0.5
0	0.5
1	0.0
1	0.5

Keeping all columns

## $\underline{\text{unnamed } [9 \ 6]}$ :

:V1	:V2	:V3	:int-part	:fract-part	:V4
1	1	0.5	0	0.5	A
2	2	1.0	1	0.0	В
1	3	1.5	1	0.5	$\mathbf{C}$
2	4	0.5	0	0.5	A
1	5	1.0	1	0.0	В
2	6	1.5	1	0.5	$\mathbf{C}$
1	7	0.5	0	0.5	A
2	8	1.0	1	0.0	В
1	9	1.5	1	0.5	$\mathbf{C}$

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)))
```

## \_unnamed [9 4]:

:V3	:v1	:v2	:v4
0.5	1	1	A
1.0	2	2	В
		3	$\mathbf{C}$
1.5	1	4	A
0.5	2	5	В
1.0		6	$\mathbf{C}$
	1	7	Α
1.5	2	8	В
0.5		9	С

```
(-> 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
0.5	1	1	A
1.0	2	2	В
		3	$\mathbf{C}$
1.5	1	4	A
0.5	2	5	В
1.0		6	$\mathbf{C}$
	1	7	A

:V3	:v1	:v2	:v4
1.5	2	8	В
0.5		9	$\mathbf{C}$

#### Tidyr examples

df-extract

\_unnamed [5 1]:

```
separate source extract source
```

```
(\texttt{def df-separate (api/dataset } \{ :x \ [\texttt{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"]}))
(\texttt{def df-extract (api/dataset } \{ : x \ [\texttt{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]:
                                                  :x
                                                  a.b
                                                  a.d
                                                  b.c
df-separate2
_unnamed [4 1]:
                                                 :x
                                                 a b
                                                 a b c
df-separate3
_unnamed [4 1]:
                                                  :x
                                                  a?b
                                                  a.b
                                                  b:c
```

d-e(api/separate-column df-separate :x [:A :B] "\\.") \_unnamed [4 2]: :A :В b d  $^{\mathrm{c}}$ 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] "\\.") \_unnamed [4 1]: :В b d Extra data is dropped (api/separate-column df-separate2 :x ["a" "b"] " ")  $\underline{\quad}$  unnamed [4 2]: b b a b Split with regular expression (api/separate-column df-separate3 :x ["a" "b"] "[?\\.:]")  $\underline{\phantom{a}}$ unnamed [4 2]:

:x

a-b a-d b-c

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

b

a b

Only a,b,c,d strings

```
(api/separate-column df-extract :x ["A" "B"] #"([a-d]+)-([a-d]+)")
```

 $\underline{\quad}$  unnamed [5 2]:

A	В
a	b
a	d
b	$\mathbf{c}$

## Fold/Unroll Rows

To pack or unpack the data into single value you can use fold-by and unroll functions.

fold-by groups dataset and packs columns data from each group separately into desired datastructure (like vector or sequence). unroll does the opposite.

## Fold-by

Group-by and pack columns into vector

```
(api/fold-by DS [:V3 :V4 :V1])
```

\_unnamed [6 4]:

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

You can pack several columns at once.

```
(api/fold-by DS [:V4])
```

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

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

You can use custom packing function

```
(api/fold-by DS [:V4] seq)
```

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

:V4	:V1	:V2	:V3
В	$(2\ 1\ 2)$	$(2\ 5\ 8)$	$(1.0\ 1.0\ 1.0)$
$\mathbf{C}$	$(1\ 2\ 1)$	(3 6 9)	$(1.5 \ 1.5 \ 1.5)$
A	$(1\ 2\ 1)$	$(1 \ 4 \ 7)$	$(0.5 \ 0.5 \ 0.5)$

or

```
(api/fold-by DS [:V4] set)
```

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

:V4	:V1	:V2	:V3
В	#{1 2}	#{2 5 8}	#{1.0}
$\mathbf{C}$	$\#\{1\ 2\}$	$\#\{6\ 3\ 9\}$	$\#\{1.5\}$
A	$\#\{1\ 2\}$	$\#\{7\ 1\ 4\}$	$\#\{0.5\}$

This works also on grouped dataset

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

\_unnamed [6 4]:

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

#### Unroll

unroll unfolds sequences stored in data, multiplying other ones when necessary. You can unroll more than one column at once (folded data should have the same size!).

Options:

- :indexes? if true (or column name), information about index of unrolled sequence is added.
- :datatypes list of datatypes which should be applied to restored columns, a map

Unroll one column

```
(api/unroll (api/fold-by DS [:V4]) [:V1])
```

\_unnamed [9 4]:

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

Unroll all folded columns

```
(api/unroll (api/fold-by DS [:V4]) [:V1 :V2 :V3])
```

\_unnamed [9 4]:

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

Unroll one by one leads to cartesian product

```
(-> DS
          (api/fold-by [:V4 :V1])
          (api/unroll [:V2])
          (api/unroll [:V3]))
```

\_unnamed [15 4]:

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

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

You can add indexes

```
(api/unroll (api/fold-by DS [:V1]) [:V4 :V2 :V3] {:indexes? true})
```

\_unnamed [9 5]:

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

```
(api/unroll (api/fold-by DS [:V1]) [:V4 :V2 :V3] {:indexes? "vector idx"})
```

\_unnamed [9 5]:

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

You can also force datatypes

\_unnamed :column info [4 4]:

:name	:datatype	$: \! n\text{-}elems$	:categorical?
:V1	:int64	9	
:V4	:string	9	true
:V2	:int16	9	
:V3	:float32	9	

This works also on grouped dataset

```
(-> DS
    (api/group-by :V1)
    (api/fold-by [:V1 :V4])
    (api/unroll :V3 {:indexes? true})
    (api/ungroup))
```

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

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

#### Reshape

Reshaping data provides two types of operations:

- pivot->longer converting columns to rows
- pivot->wider converting rows to columns

Both functions are inspired on tidyr R package and provide almost the same functionality.

All examples are taken from mentioned above documentation.

Both functions work only on regular dataset.

#### Longer

pivot->longer converts columns to rows. Column names are treated as data.

Arguments:

- dataset
- columns selector
- options:
  - :target-columns names of the columns created or columns pattern (see below) (default: :\$column)
  - :value-column-name name of the column for values (default: :\$value)
  - :splitter string, regular expression or function which splits source column names into data
  - :drop-missing? remove rows with missing? (default: :true)

- :datatypes - map of target columns data types

:target-columns - can be:

- $\bullet\,$  column name source columns names are put there as a data
- column names as sequence source columns names after split are put separately into :target-columns as data
- pattern is a sequence of names, where some of the names are nil. nil is replaced by a name taken from splitter and such column is used for values.

Create rows from all columns but "religion".

(def relig-income (api/dataset "data/relig\_income.csv"))

#### relig-income

data/relig\_income.csv [18 11]:

religion	<\$10k	\$10- 20k	\$20- 30k	\$30- 40k	\$40- 50k	\$50- 75k	\$75- 100k	\$100- 150k	>150k	Don't know/refused
Agnostic	27	34	60	81	76	137	122	109	84	96
Atheist	12	27	37	52	35	70	73	59	74	76
Buddhist	27	21	30	34	33	58	62	39	53	54
Catholic	418	617	732	670	638	1116	949	792	633	1489
Don't	15	14	15	11	10	35	21	17	18	116
know/refused										
Evangelical	575	869	1064	982	881	1486	949	723	414	1529
Prot										
Hindu	1	9	7	9	11	34	47	48	54	37
Historically	228	244	236	238	197	223	131	81	78	339
Black Prot										
Jehovah's	20	27	24	24	21	30	15	11	6	37
Witness										
Jewish	19	19	25	25	30	95	69	87	151	162
Mainline Prot	289	495	619	655	651	1107	939	753	634	1328
Mormon	29	40	48	51	56	112	85	49	42	69
Muslim	6	7	9	10	9	23	16	8	6	22
Orthodox	13	17	23	32	32	47	38	42	46	73
Other	9	7	11	13	13	14	18	14	12	18
Christian										
Other Faiths	20	33	40	46	49	63	46	40	41	71
Other World	5	2	3	4	2	7	3	4	4	8
Religions										
Unaffiliated	217	299	374	365	341	528	407	321	258	597

(api/pivot->longer relig-income (complement #{"religion"}))

data/relig\_income.csv [180 3]:

religion	:column  :value	e
Agnostic	<\$10k	27
Atheist	<\$10k	12
Buddhist	< \$10k	27

religion	:column :value	
Catholic	<\$10k	418
Don't know/refused	<\$10k	15
Evangelical Prot	<\$10k	575
Hindu	<\$10k	1
Historically Black Prot	<\$10k	228
Jehovah's Witness	<\$10k	20
Jewish	<\$10k	19
Mainline Prot	<\$10k	289
Mormon	<\$10k	29
Muslim	<\$10k	6
Orthodox	<\$10k	13
Other Christian	<\$10k	9
Other Faiths	<\$10k	20
Other World Religions	<\$10k	5
Unaffiliated	<\$10k	217
Agnostic	Don't know/refused	96
Atheist	Don't know/refused	76
Buddhist	Don't know/refused	54
Catholic	Don't know/refused	1489
Don't know/refused	Don't know/refused	116
Evangelical Prot	Don't know/refused	1529
Hindu	Don't know/refused	37

 $Convert\ only\ columns\ starting\ with\ \verb"wk"\ and\ pack\ them\ into\ \verb:week\ column,\ values\ go\ to\ \verb:rank\ column$ 

(api/column-names)
(take 13)
(api/select-columns bilboard))

data/billboard.csv.gz [317 13]:

artist	track	date.entered	wk1	wk2	wk3	wk4	wk5	wk6	wk7	wk8	-
2 Pac	Baby Don't Cry (Keep	2000-02-26	87	82	72	77	87	94	99		
2Ge+her	The Hardest Part Of	2000-09-02	91	87	92						
3 Doors Down	Kryptonite	2000-04-08	81	70	68	67	66	57	54	53	ļ
3 Doors Down	Loser	2000-10-21	76	76	72	69	67	65	55	59	(
504 Boyz	Wobble Wobble	2000-04-15	57	34	25	17	17	31	36	49	ļ
98^0	Give Me Just One Nig	2000-08-19	51	39	34	26	26	19	2	2	;
A*Teens	Dancing Queen	2000-07-08	97	97	96	95	100				
Aaliyah	I Don't Wanna	2000-01-29	84	62	51	41	38	35	35	38	;
Aaliyah	Try Again	2000-03-18	59	53	38	28	21	18	16	14	
Adams, Yolanda	Open My Heart	2000-08-26	76	76	74	69	68	67	61	58	ļ
Adkins, Trace	More	2000-04-29	84	84	75	73	73	69	68	65	,
Aguilera, Christina	Come On Over Baby (A	2000-08-05	57	47	45	29	23	18	11	9	9
Aguilera, Christina	I Turn To You	2000-04-15	50	39	30	28	21	19	20	17	
Aguilera, Christina	What A Girl Wants	1999-11-27	71	51	28	18	13	13	11	1	
Alice Deejay	Better Off Alone	2000-04-08	79	65	53	48	45	36	34	29	

artist	track	date.entered	wk1	wk2	wk3	wk4	wk5	wk6	wk7	wk8	7
Allan, Gary	Smoke Rings In The D	2000-01-22	80	78	76	77	92				
Amber	Sexual	1999-07-17	99	99	96	96	100	93	93	96	
Anastacia	I'm Outta Love	2000-04-01	92			95					
Anthony, Marc	My Baby You	2000-09-16	82	76	76	70	82	81	74	80	7
Anthony, Marc	You Sang To Me	2000-02-26	77	54	50	43	30	27	21	18	1
Avant	My First Love	2000-11-04	70	62	56	43	39	33	26	26	4
Avant	Separated	2000-04-29	62	32	30	23	26	30	35	32	;
BBMak	Back Here	2000-04-29	99	86	60	52	38	34	28	21	1
Backstreet Boys, The	Shape Of My Heart	2000-10-14	39	25	24	15	12	12	10	9	1
Backstreet Boys, The	Show Me The Meaning	2000-01-01	74	62	55	25	16	14	12	10	-

data/billboard.csv.gz [5307 5]:

artist	track	date.entered	:week	:rank
3 Doors Down	Kryptonite	2000-04-08	wk35	4
Braxton, Toni	He Wasn't Man Enough	2000-03-18	wk35	34
Creed	Higher	1999-09-11	wk35	22
Creed	With Arms Wide Open	2000-05-13	wk35	5
Hill, Faith	Breathe	1999-11-06	wk35	8
Joe	I Wanna Know	2000-01-01	wk35	5
Lonestar	Amazed	1999-06-05	wk35	14
Vertical Horizon	Everything You Want	2000-01-22	wk35	27
matchbox twenty	Bent	2000-04-29	wk35	33
Creed	Higher	1999-09-11	wk55	21
Lonestar	Amazed	1999-06-05	wk55	22
3 Doors Down	Kryptonite	2000-04-08	wk19	18
3 Doors Down	Loser	2000-10-21	wk19	73
98^0	Give Me Just One Nig	2000-08-19	wk19	93
Aaliyah	I Don't Wanna	2000-01-29	wk19	83
Aaliyah	Try Again	2000-03-18	wk19	3
Adams, Yolanda	Open My Heart	2000-08-26	wk19	79
Aguilera, Christina	Come On Over Baby (A	2000-08-05	wk19	23
Aguilera, Christina	I Turn To You	2000-04-15	wk19	29
Aguilera, Christina	What A Girl Wants	1999-11-27	wk19	18
Alice Deejay	Better Off Alone	2000-04-08	wk19	79
Amber	Sexual	1999-07-17	wk19	95
Anthony, Marc	My Baby You	2000-09-16	wk19	91
Anthony, Marc	You Sang To Me	2000-02-26	wk19	9
Avant	My First Love	2000-11-04	wk19	81

We can create numerical column out of column names

# data/billboard.csv.gz [5307 5]:

artist	track	date.entered	:week	:rank
3 Doors Down	Kryptonite	2000-04-08	46	21
Creed	Higher	1999-09-11	46	7
Creed	With Arms Wide Open	2000-05-13	46	37
Hill, Faith	Breathe	1999-11-06	46	31
Lonestar	Amazed	1999-06-05	46	5
3 Doors Down	Kryptonite	2000-04-08	51	42
Creed	Higher	1999-09-11	51	14
Hill, Faith	Breathe	1999-11-06	51	49
Lonestar	Amazed	1999-06-05	51	12
2 Pac	Baby Don't Cry (Keep	2000-02-26	6	94
3 Doors Down	Kryptonite	2000-04-08	6	57
3 Doors Down	Loser	2000-10-21	6	65
504 Boyz	Wobble Wobble	2000-04-15	6	31
98^0	Give Me Just One Nig	2000-08-19	6	19
Aaliyah	I Don't Wanna	2000-01-29	6	35
Aaliyah	Try Again	2000-03-18	6	18
Adams, Yolanda	Open My Heart	2000-08-26	6	67
Adkins, Trace	More	2000-04-29	6	69
Aguilera, Christina	Come On Over Baby (A	2000-08-05	6	18
Aguilera, Christina	I Turn To You	2000-04-15	6	19
Aguilera, Christina	What A Girl Wants	1999-11-27	6	13
Alice Deejay	Better Off Alone	2000-04-08	6	36
Amber	Sexual	1999-07-17	6	93
Anthony, Marc	My Baby You	2000-09-16	6	81
Anthony, Marc	You Sang To Me	2000-02-26	6	27

When column names contain observation data, such column names can be splitted and data can be restored into separate columns.

data/who.csv.gz [7240 10]:

country	iso2	iso3	year	new_	sp_	m <b>01e4</b> v	_sp_	_m <b>h52</b> 4_	_sp_	_m <b>253</b> 4_	_sp_	_m <b>3\54</b> 4_	_sp_	_m <b>455</b> 4_	_sp_	_m5564
Afghanista	anAF	AFG	1980													
Afghanista	anAF	AFG	1981													
Afghanista	anAF	AFG	1982													
Afghanista	anAF	AFG	1983													
Afghanista	anAF	AFG	1984													
Afghanista	anAF	AFG	1985													
Afghanista	anAF	AFG	1986													
Afghanista	anAF	AFG	1987													
Afghanista	anAF	AFG	1988													
Afghanista	anAF	AFG	1989													

					01.4	1504	0504	0511	1221	
country	iso2	iso3	year	new_	sp_m <b>u.e</b> 4v_	_sp_m <b>h524</b> _sp_	_m <b>x</b> asv4_sp_	_m <b>x</b> xxx4v4_sp	_m4A&5v4_sp_	_m
Afghanista	nAF	AFG	1990							
Afghanista	nAF	AFG	1991							
Afghanista	nAF	AFG	1992							
Afghanista	nAF	AFG	1993							
Afghanista	ınAF	AFG	1994							
Afghanista	nAF	AFG	1995							
Afghanista	nAF	AFG	1996							
Afghanista	nAF	AFG	1997	0	10	6	3	5	2	
Afghanista	nAF	AFG	1998	30	129	128	90	89	64	
Afghanista	ınAF	AFG	1999	8	55	55	47	34	21	
Afghanista	ınAF	AFG	2000	52	228	183	149	129	94	
Afghanista	ınAF	AFG	2001	129	379	349	274	204	139	
Afghanista	nAF	AFG	2002	90	476	481	368	246	241	
Afghanista	ınAF	AFG	2003	127	511	436	284	256	288	
Afghanista	nAF	AFG	2004	139	537	568	360	358	386	

data/who.csv.gz [76046 8]:

country	iso2	iso3	year	:diagnosis	:gender	:age	:count
Albania	AL	ALB	2013	rel	m	1524	60
Algeria	DZ	DZA	2013	$_{\mathrm{rel}}$	m	1524	1021
Andorra	AD	AND	2013	rel	m	1524	0
Angola	AO	AGO	2013	rel	m	1524	2992
Anguilla	AI	AIA	2013	rel	m	1524	0
Antigua and Barbuda	$\overline{AG}$	ATG	2013	rel	m	1524	1
Argentina	AR	ARG	2013	rel	m	1524	1124
Armenia	AM	ARM	2013	rel	m	1524	116
Australia	$\mathrm{AU}$	AUS	2013	rel	m	1524	105
Austria	AT	AUT	2013	rel	m	1524	44
Azerbaijan	AZ	AZE	2013	rel	m	1524	958
Bahamas	$_{\mathrm{BS}}$	BHS	2013	$_{\mathrm{rel}}$	m	1524	2
Bahrain	BH	BHR	2013	$_{\mathrm{rel}}$	m	1524	13
Bangladesh	BD	BGD	2013	rel	m	1524	14705
Barbados	BB	BRB	2013	$_{\mathrm{rel}}$	m	1524	0
Belarus	BY	BLR	2013	$_{\mathrm{rel}}$	m	1524	162
Belgium	BE	$\operatorname{BEL}$	2013	$_{\mathrm{rel}}$	m	1524	63
Belize	BZ	BLZ	2013	$_{\mathrm{rel}}$	m	1524	8
Benin	$_{\mathrm{BJ}}$	BEN	2013	$_{\mathrm{rel}}$	m	1524	301
Bermuda	BM	BMU	2013	$_{\mathrm{rel}}$	m	1524	0
Bhutan	$\operatorname{BT}$	BTN	2013	rel	m	1524	180
Bolivia (Plurinational State of)	ВО	BOL	2013	rel	m	1524	1470
Bonaire, Saint Eustatius and Saba	$_{\mathrm{BQ}}$	BES	2013	rel	m	1524	0
Bosnia and Herzegovina	BA	BIH	2013	rel	m	1524	57
Botswana	BW	BWA	2013	$\operatorname{rel}$	m	1524	423

When data contains multiple observations per row, we can use splitter and pattern for target columns to create new columns and put values there. In following dataset we have two observations dob and gender for two childs. We want to put child infomation into the column and leave dob and gender for values.

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

#### family

data/family.csv [5 5]:

family	dob_child1	dob_child2	gender_child1	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

data/family.csv [9 4]:

family	:child	dob	gender
1	child1	1998-11-26	1
2	child1	1996-06-22	2
3	child1	2002-07-11	2
4	child1	2004-10-10	1
5	child1	2000 - 12 - 05	2
1	child2	2000 - 01 - 29	2
3	child2	2004-04-05	2
4	child2	2009-08-27	1
5	child2	2005-02-28	1

Similar here, we have two observations:  ${\tt x}$  and  ${\tt y}$  in four groups.

```
(def anscombe (api/dataset "data/anscombe.csv"))
```

#### anscombe

data/anscombe.csv [11 8]:

x1	x2	х3	x4	y1	y2	у3	y4
10	10	10	8	8.04	9.14	7.46	6.58
8	8	8	8	6.95	8.14	6.77	5.76
13	13	13	8	7.58	8.74	12.74	7.71
9	9	9	8	8.81	8.77	7.11	8.84
11	11	11	8	8.33	9.26	7.81	8.47
14	14	14	8	9.96	8.10	8.84	7.04
6	6	6	8	7.24	6.13	6.08	5.25
4	4	4	19	4.26	3.10	5.39	12.50
12	12	12	8	10.84	9.13	8.15	5.56

x1	x2	x3	x4	y1	y2	у3	y4
7	7	7	8	4.82	7.26	6.42	7.91
5	5	5	8	5.68	4.74	5.73	6.89

data/anscombe.csv [44 3]:

:set	X	У
1	10	8.04
1	8	6.95
1	13	7.58
1	9	8.81
1	11	8.33
1	14	9.96
1	6	7.24
1	4	4.26
1	12	10.84
1	7	4.82
1	5	5.68
2	10	9.14
2	8	8.14
2	13	8.74
2	9	8.77
2	11	9.26
2	14	8.10
2	6	6.13
2	4	3.10
2	12	9.13
2	7	7.26
2	5	4.74
3	10	7.46
3	8	6.77
3	13	12.74

\_unnamed [4 7]:

:x	:a	:b	:y1	:y2	:z1	:z2
1	1	0	0.16392200	0.94276927	3	-2

:x	:a	:b	:y1	:y2	:z1	:z2
2	1	1	0.00266031	0.45980959	3	-2
3	0	1	0.30751714	0.46815845	3	-2
4	0	1	0.15802876	0.36985121	3	-2

 $\underline{\text{unnamed } [8 6]}$ :

:x	:a	:b	:times	У	Z
1	1	0	1	0.16392200	3
2	1	1	1	0.00266031	3
3	0	1	1	0.30751714	3
4	0	1	1	0.15802876	3
1	1	0	2	0.94276927	-2
2	1	1	2	0.45980959	-2
3	0	1	2	0.46815845	-2
4	0	1	2	0.36985121	-2

#### Wider

pivot->wider converts rows to columns.

Arguments:

- dataset
- columns-selector values from selected columns are converted to new columns
- value-columns what are values

When multiple columns are used as columns selector, names are joined using :concat-columns-with option. :concat-columns-with can be a string or function (default: "\_\_"). Function accepts sequence of names.

When columns-selector creates non unique set of values, they are folded using :fold-fn (default: vec) option.

When value-columns is a sequence, multiple observations as columns are created appending value column names into new columns. Column names are joined using :concat-value-with option. :concat-value-with can be a string or function (default: "-"). Function accepts current column name and value.

Use station as a name source for columns and seen for values

```
(def fish (api/dataset "data/fish_encounters.csv"))
```

#### fish

data/fish\_encounters.csv [114 3]:

fish	station	seen
4842	Release	1
4842	I80_1	1
4842	Lisbon	1
4842	Rstr	1
4842	$Base\_TD$	1

fish	station	seen
4842	BCE	1
4842	BCW	1
4842	BCE2	1
4842	BCW2	1
4842	MAE	1
4842	MAW	1
4843	Release	1
4843	I80_1	1
4843	Lisbon	1
4843	Rstr	1
4843	$Base\_TD$	1
4843	BCE	1
4843	BCW	1
4843	BCE2	1
4843	BCW2	1
4843	MAE	1
4843	MAW	1
4844	Release	1
4844	I80_1	1
4844	Lisbon	1

(api/pivot->wider fish "station" "seen" {:drop-missing? false})

data/fish\_encounters.csv [19 12]:

fish	Rstr	$Base\_TD$	I80_1	Release	MAE	BCE2	MAW	BCW2	BCE	Lisbon	BCW
4842	1	1	1	1	1	1	1	1	1	1	1
4843	1	1	1	1	1	1	1	1	1	1	1
4844	1	1	1	1	1	1	1	1	1	1	1
4850	1	1	1	1					1		1
4857	1	1	1	1		1		1	1	1	1
4858	1	1	1	1	1	1	1	1	1	1	1
4861	1	1	1	1	1	1	1	1	1	1	1
4862	1	1	1	1		1		1	1	1	1
4864			1	1							
4865			1	1						1	
4845	1	1	1	1						1	
4847			1	1						1	
4848	1		1	1						1	
4849			1	1							
4851			1	1							
4854			1	1							
4855	1	1	1	1						1	
4859	1	1	1	1						1	
4863			1	1							

If selected columns contain multiple values, such values should be folded.

```
(def warpbreaks (api/dataset "data/warpbreaks.csv"))
warpbreaks
```

data/warpbreaks.csv [54 3]:

breaks	wool	tension
26	A	
30	A	L
54	A	L
25	A	L
70	A	L
52	A	${ m L}$
51	A	${ m L}$
26	A	L
67	A	L
18	A	${ m M}$
21	A	${ m M}$
29	A	${ m M}$
17	A	${ m M}$
12	A	${ m M}$
18	A	${ m M}$
35	A	${ m M}$
30	A	${ m M}$
36	A	M
36	A	Η
21	A	Η
24	A	H
18	A	H
10	A	Η
43	A	Η
28	A	Η

Let's see how many values are for each type of wool and tension groups

```
(-> warpbreaks
   (api/group-by ["wool" "tension"])
   (api/aggregate {:n api/row-count}))
```

 $\underline{\phantom{a}}$ unnamed [6 3]:

```
wool
       tension
                 :n
Α
       Η
                 9
В
       Η
                 9
                 9
A
       \mathbf{L}
Α
       Μ
                 9
                 9
В
       L
В
       Μ
                 9
```

```
(-> warpbreaks
   (api/reorder-columns ["wool" "tension" "breaks"])
   (api/pivot->wider "wool" "breaks" {:fold-fn vec}))
```

data/warpbreaks.csv [3 3]:

tension	В	A
M	[42 26 19 16 39 28 21 39 29]	[18 21 29 17 12 18 35 30 36]
H	[20 21 24 17 13 15 15 16 28]	[36 21 24 18 10 43 28 15 26]
L	$[27\ 14\ 29\ 19\ 29\ 31\ 41\ 20\ 44]$	$[26\ 30\ 54\ 25\ 70\ 52\ 51\ 26\ 67]$

We can also calculate mean (aggreate values)

```
(-> warpbreaks
    (api/reorder-columns ["wool" "tension" "breaks"])
    (api/pivot->wider "wool" "breaks" {:fold-fn tech.v3.datatype.functional/mean}))
```

data/warpbreaks.csv [3 3]:

tension	В	A
H	18.7777778	24.5555556
${ m M}$	28.77777778	24.00000000
L	28.2222222	44.55555556

Multiple source columns, joined with default separator.

```
(def production (api/dataset "data/production.csv"))
```

#### production

data/production.csv [45 4]:

product	country	year	production
A	AI	2000	1.63727158
A	AI	2001	0.15870784
A	AI	2002	-1.56797745
A	AI	2003	-0.44455509
A	AI	2004	-0.07133701
A	AI	2005	1.61183090
A	AI	2006	-0.70434682
A	AI	2007	-1.53550542
A	AI	2008	0.83907155
A	AI	2009	-0.37424110
A	AI	2010	-0.71158926
A	AI	2011	1.12805634
A	AI	2012	1.45718247
A	AI	2013	-1.55934101
A	AI	2014	-0.11695838
В	AI	2000	-0.02617661
В	AI	2001	-0.68863576
В	AI	2002	0.06248741
В	AI	2003	-0.72339686
В	AI	2004	0.47248952
В	AI	2005	-0.94173861

product	country	year	production
В	AI	2006	-0.34782108
В	AI	2007	0.52425284
В	AI	2008	1.83230937
В	AI	2009	0.10706491

# (api/pivot->wider production ["product" "country"] "production")

data/production.csv [15 4]:

year	A_AI	B_EI	B_AI
2000	1.63727158	1.40470848	-0.02617661
2001	0.15870784	-0.59618369	-0.68863576
2002	-1.56797745	-0.26568579	0.06248741
2003	-0.44455509	0.65257808	-0.72339686
2004	-0.07133701	0.62564999	0.47248952
2005	1.61183090	-1.34530299	-0.94173861
2006	-0.70434682	-0.97184975	-0.34782108
2007	-1.53550542	-1.69715821	0.52425284
2008	0.83907155	0.04556128	1.83230937
2009	-0.37424110	1.19315043	0.10706491
2010	-0.71158926	-1.60557503	-0.32903664
2011	1.12805634	-0.77235497	-1.78319121
2012	1.45718247	-2.50262738	0.61125798
2013	-1.55934101	-1.62753769	-0.78526092
2014	-0.11695838	0.03329645	0.97843635

Joined with custom function

(api/pivot->wider production ["product" "country"] "production" {:concat-columns-with vec})

data/production.csv [15 4]:

year	["A" "AI"]	["B" "EI"]	["B" "AI"]
2000	1.63727158	1.40470848	-0.02617661
2001	0.15870784	-0.59618369	-0.68863576
2002	-1.56797745	-0.26568579	0.06248741
2003	-0.44455509	0.65257808	-0.72339686
2004	-0.07133701	0.62564999	0.47248952
2005	1.61183090	-1.34530299	-0.94173861
2006	-0.70434682	-0.97184975	-0.34782108
2007	-1.53550542	-1.69715821	0.52425284
2008	0.83907155	0.04556128	1.83230937
2009	-0.37424110	1.19315043	0.10706491
2010	-0.71158926	-1.60557503	-0.32903664
2011	1.12805634	-0.77235497	-1.78319121
2012	1.45718247	-2.50262738	0.61125798
2013	-1.55934101	-1.62753769	-0.78526092
2014	-0.11695838	0.03329645	0.97843635

# Multiple value columns

(def income (api/dataset "data/us\_rent\_income.csv"))

## income

data/us\_rent\_income.csv [104 5]:

GEOID	NAME	variable	estimate	moe
1	Alabama	income	24476	136
1	Alabama	rent	747	3
2	Alaska	income	32940	508
2	Alaska	rent	1200	13
4	Arizona	income	27517	148
4	Arizona	rent	972	4
5	Arkansas	income	23789	165
5	Arkansas	rent	709	5
6	California	income	29454	109
6	California	rent	1358	3
8	Colorado	income	32401	109
8	Colorado	rent	1125	5
9	Connecticut	income	35326	195
9	Connecticut	rent	1123	5
10	Delaware	income	31560	247
10	Delaware	rent	1076	10
11	District of Columbia	income	43198	681
11	District of Columbia	rent	1424	17
12	Florida	income	25952	70
12	Florida	rent	1077	3
13	Georgia	income	27024	106
13	Georgia	rent	927	3
15	Hawaii	income	32453	218
15	Hawaii	rent	1507	18
16	Idaho	income	25298	208

(api/pivot->wider income "variable" ["estimate" "moe"] {:drop-missing? false})

data/us\_rent\_income.csv [52 6]:

GEOID	NAME	rent-estimate	rent-moe	income-estimate	income-moe
1	Alabama	747	3	24476	136
2	Alaska	1200	13	32940	508
4	Arizona	972	4	27517	148
5	Arkansas	709	5	23789	165
6	California	1358	3	29454	109
8	Colorado	1125	5	32401	109
9	Connecticut	1123	5	35326	195
10	Delaware	1076	10	31560	247
11	District of Columbia	1424	17	43198	681
12	Florida	1077	3	25952	70
13	Georgia	927	3	27024	106
15	Hawaii	1507	18	32453	218
16	Idaho	792	7	25298	208
17	Illinois	952	3	30684	83

GEOID	NAME	rent-estimate	rent-moe	income-estimate	income-moe
18	Indiana	782	3	27247	117
19	Iowa	740	4	30002	143
20	Kansas	801	5	29126	208
21	Kentucky	713	4	24702	159
22	Louisiana	825	4	25086	155
23	Maine	808	7	26841	187
24	Maryland	1311	5	37147	152
25	Massachusetts	1173	5	34498	199
26	Michigan	824	3	26987	82
27	Minnesota	906	4	32734	189
28	Mississippi	740	5	22766	194

Value concatenated by custom function

data/us\_rent\_income.csv [52 6]:

		["rent"	["rent"	["income"	["income"
GEOID	NAME	"estimate"]	"moe"]	"estimate"]	"moe"]
1	Alabama	747	3	24476	136
2	Alaska	1200	13	32940	508
4	Arizona	972	4	27517	148
5	Arkansas	709	5	23789	165
6	California	1358	3	29454	109
8	Colorado	1125	5	32401	109
9	Connecticut	1123	5	35326	195
10	Delaware	1076	10	31560	247
11	District of	1424	17	43198	681
	Columbia				
12	Florida	1077	3	25952	70
13	Georgia	927	3	27024	106
15	Hawaii	1507	18	32453	218
16	Idaho	792	7	25298	208
17	Illinois	952	3	30684	83
18	Indiana	782	3	27247	117
19	Iowa	740	4	30002	143
20	Kansas	801	5	29126	208
21	Kentucky	713	4	24702	159
22	Louisiana	825	4	25086	155
23	Maine	808	7	26841	187
24	Maryland	1311	5	37147	152
25	Massachusetts	1173	5	34498	199
26	Michigan	824	3	26987	82
27	Minnesota	906	4	32734	189
28	Mississippi	740	5	22766	194

Reshape contact data

# (def contacts (api/dataset "data/contacts.csv")) contacts

data/contacts.csv [6 3]:

field	value	person_id
name	Jiena McLellan	1
company	Toyota	1
name	John Smith	2
company	google	2
email	john@google.com	2
name	Huxley Ratcliffe	3

```
(api/pivot->wider contacts "field" "value" {:drop-missing? false})
```

data/contacts.csv [3 4]:

person_id	email	name	company
1 2 3	john@google.com	Jiena McLellan John Smith Huxley Ratcliffe	Toyota google

## Reshaping

A couple of tidyr examples of more complex reshaping.

World bank

```
(def world-bank-pop (api/dataset "data/world_bank_pop.csv.gz"))
(->> world-bank-pop
```

```
(api/column-names)
  (take 8)
  (api/select-columns world-bank-pop))
```

data/world\_bank\_pop.csv.gz [1056 8]:

country	indicator	2000	2001	2002	2003	2004	2005
$\overline{\mathrm{ABW}}$	SP.URB.7	COLT214440000	E+0430480000	DE+043670000	0E+0442460000	E+0446690000	E+0448890000E+0
ABW	SP.URB.0	GRO <b>W</b> 263237	E+01041302122	2E+0104345595	3E+01031036044	4E+090.51477684	E- 4.91302715E-
						01	01
ABW	SP.POP.T	O <b>9TU</b> 8530000	E+0428980000	0E+0944992000	0E+09470170000	DE+09487370000	E+0400031000E+0
ABW	SP.POP.C	R2000/502678	E+02022593013	BE+0202290560	5E+0 <b>2</b> 010935434	4E + 0075735287	E+01030203884E+0
AFG	SP.URB.7	COFTAB629900	E+04664805500	0E+0468929510	0E+05615568600	0E+05642677000	E+0669182300E+0
AFG	SP.URB.C	GR <b>XXXI</b> 222846	E+0066283822	2E+0 <b>5</b> 01346745	4E+0502304585	3E+05012439302	E + 0.076864700E + 0
AFG	SP.POP.T	CO2TU0937560	E+027.09664630	0E+0 <b>27</b> .1979923	0E+0 <b>27</b> .30648510	)E+0 <b>27</b> .41189790	E+027.50707980E+0
AFG	SP.POP.C	R <b>XX</b> 9465874	E+0025150411	1E + 007205284	6E+04081804112	2E+01046891840	E+03087047016E+0
AGO	SP.URB.7	T <b>C8T2B</b> 476600	E+0670800000	0E+062187870	0E+0676519700	0E+0603435060	E+017.09494240E+0
AGO	SP.URB.0	G <b>ROW</b> 749411	E+0505877195	4E+0507001323	7E+007581271	1E+0 <b>5</b> 075341450	E + 0069279690E + 0
AGO	SP.POP.T	ОП 64409240	E+017.69832660	)E+017.7572649	0E+0 <b>17</b> .82033690	E+017.88657160	E+017.95525420E+0

country	indicator	2000	2001	2002	2003	2004	2005
AGO	SP.POP.G	R30039294342E+	-00024549139E+	0 <b>3</b> 041151529E+	0 <b>3</b> 052630277E+	03057396197E+	08057589970E+00
ALB	SP.URB.T	COTESS 939100E+	01629858400E+	01632722000E+	01635484800E+	01638182800E+	01640729800E+06
ALB	SP.URB.C	F7012478629E-	7.10442618E-	2.18120890E +	O2006027418E+	01097179894E+	01082642936E+00
		01	01				
ALB	SP.POP.T	ONU8902700E+	-0606017300E+	0605101000E+	0603961600E+	0602693900E+	06601148700E+06
ALB	SP.POP.G	ROW	-	-	-	-	-
		6.37356834E-	9.38470428E-	2.99876697E-	3.74149169E-	4.17931378E-	5.11790116E-
		01	01	01	01	01	01
AND	SP.URB.T	CONTINUA 170000E+	-06419910000E+	-06441940000E+	0467470000E+	-06491920000E+	07412050000E+04
AND	SP.URB.C	R <b>○779</b> 31383E+	O2057186909E+	00049205352E+	<b>®</b> 089996041E+	00059758966E+	O2086777917E+00
AND	SP.POP.T	OT13900000E+	-06473410000E+	-0 <b>7</b> 400490000E+	0 <b>4</b> 31820000E+	07462440000E+	07488670000E+04
AND	SP.POP.G	RIOW216555E+	(M)93999221E+	00094257335E+	0 <b>1</b> 037544919E+	0009892348E+	OB)38241655E+00
ARB	SP.URB.T	COTAI9981223E+	-0853924351E+	0857985738E+	0862267754E +	0866820459E+	01871813698E+08
ARB	SP.URB.C	FROM 956290E+	-02062908111E+	OD 63856042E+	O2071038136E+	0080567450E+	O2099318143E+00
ARB	SP.POP.T	O2T83832016E+	-02889850357E+	0 <b>8</b> 96026575E+	<b>08</b> 02434519E+	0809162029E+	<b>08</b> 16264728E+08
ARB	SP.POP.G	R2OM/148598E+	(M)12038835E+	OD 13082988E+	O2016465160E+	(M)22445177E+	O2029740341E+00
ARE	SP.URB.T	<b>C2T5B</b> 138600E+	-0668261100E+	0684320800E+	0604862700E+	0634683000E+	03676723900E+06

Step 1 - convert years column into values

pop2

data/world\_bank\_pop.csv.gz [19008 4]:

country	indicator	year	value
ABW	SP.URB.TOTL	2013	4.43600000E+04
ABW	SP.URB.GROW	2013	6.69503994 E-01
ABW	SP.POP.TOTL	2013	1.03187000E + 05
ABW	SP.POP.GROW	2013	5.92914005 E-01
AFG	SP.URB.TOTL	2013	7.73396400E+06
AFG	SP.URB.GROW	2013	4.19297967E+00
AFG	SP.POP.TOTL	2013	3.17316880E + 07
AFG	SP.POP.GROW	2013	3.31522413E+00
AGO	SP.URB.TOTL	2013	1.61194910E+07
AGO	SP.URB.GROW	2013	4.72272270E+00
AGO	SP.POP.TOTL	2013	2.59983400E+07
AGO	SP.POP.GROW	2013	3.53182419E+00
ALB	SP.URB.TOTL	2013	1.60350500E+06
ALB	SP.URB.GROW	2013	1.74363937E+00
ALB	SP.POP.TOTL	2013	2.89509200E+06
ALB	SP.POP.GROW	2013	-1.83211385E-01
AND	SP.URB.TOTL	2013	7.15270000E+04
AND	SP.URB.GROW	2013	-2.11923331E+00
AND	SP.POP.TOTL	2013	8.07880000E+04
AND	SP.POP.GROW	2013	-2.01331401E+00
ARB	SP.URB.TOTL	2013	2.18605128E+08
ARB	SP.URB.GROW	2013	2.78289395E+00
ARB	SP.POP.TOTL	2013	3.81702086E + 08

country	indicator	year	value
ARB	SP.POP.GROW		2.24884429E+00
ARE	SP.URB.TOTL		7.66126800E+06

Step 2 - separate "indicate" column

## pop3

data/world\_bank\_pop.csv.gz [19008 5]:

area	variable	year	value
URB	TOTL	2013	4.43600000E+04
URB	GROW	2013	6.69503994E-01
POP	TOTL	2013	1.03187000E + 05
POP	GROW	2013	5.92914005E- $01$
URB	TOTL	2013	7.73396400E+06
URB	GROW	2013	4.19297967E+00
POP	TOTL	2013	3.17316880E + 07
POP	GROW	2013	3.31522413E+00
URB	TOTL	2013	1.61194910E + 07
URB	GROW	2013	4.72272270E+00
POP	TOTL	2013	2.59983400E+07
POP	GROW	2013	3.53182419E+00
URB	TOTL	2013	1.60350500E + 06
URB	GROW	2013	1.74363937E+00
POP	TOTL	2013	2.89509200E+06
POP	GROW	2013	-1.83211385E-01
URB	TOTL	2013	7.15270000E+04
URB	GROW	2013	-2.11923331E+00
POP	TOTL	2013	8.07880000E+04
POP	GROW	2013	-2.01331401E+00
URB	TOTL	2013	2.18605128E + 08
URB	GROW	2013	2.78289395E+00
POP	TOTL	2013	3.81702086E + 08
POP	GROW	2013	2.24884429E+00
URB	TOTL	2013	7.66126800E + 06
	URB URB POP POP URB URB POP POP URB URB POP POP URB URB	URB GROW POP TOTL POP GROW URB TOTL POP GROW URB TOTL POP GROW URB TOTL URB GROW URB TOTL URB GROW URB TOTL URB GROW POP TOTL POP GROW URB TOTL URB GROW URB TOTL URB GROW POP TOTL POP GROW URB TOTL URB GROW POP TOTL POP GROW URB TOTL POP GROW URB TOTL POP GROW URB TOTL POP GROW URB TOTL	URB         TOTL         2013           URB         GROW         2013           POP         TOTL         2013           POP         GROW         2013           URB         TOTL         2013           URB         GROW         2013           POP         TOTL         2013           POP         GROW         2013           URB         TOTL         2013           URB         GROW         2013           POP         TOTL         2013           URB         TOTL         2013           URB         GROW         2013           POP         TOTL         2013           POP         GROW         2013           URB         TOTL         2013           URB         GROW         2013           POP         TOTL         2013           POP         TOTL         2013           POP         GROW         2013           POP         GROW         2013           URB         TOTL         2013           URB         TOTL         2013           URB         GROW         2013           URB

Step 3 - Make columns based on "variable" values.

```
(api/pivot->wider pop3 "variable" "value" {:drop-missing? false})
```

data/world\_bank\_pop.csv.gz [9504 5]:

country	area	year	GROW	TOTL
ABW	URB	2013	0.66950399	4.43600000E+04
ABW	POP	2013	0.59291401	1.03187000E + 05
AFG	URB	2013	4.19297967	7.73396400E+06
AFG	POP	2013	3.31522413	3.17316880E+07

country	area	year	GROW	TOTL
AGO	URB	2013	4.72272270	1.61194910E+07
AGO	POP	2013	3.53182419	$2.59983400\mathrm{E}{+07}$
ALB	URB	2013	1.74363937	$1.60350500\mathrm{E}{+06}$
ALB	POP	2013	-0.18321138	$2.89509200\mathrm{E}{+06}$
AND	URB	2013	-2.11923331	$7.15270000\mathrm{E}{+04}$
AND	POP	2013	-2.01331401	8.07880000E+04
ARB	URB	2013	2.78289395	2.18605128E + 08
ARB	POP	2013	2.24884429	3.81702086E + 08
ARE	URB	2013	1.55515587	$7.66126800\mathrm{E}{+06}$
ARE	POP	2013	1.18180499	9.00626300E + 06
ARG	URB	2013	1.18764913	3.88172560E + 07
ARG	POP	2013	1.04727675	$4.25399250\mathrm{E}{+07}$
ARM	URB	2013	0.28102719	$1.82765600\mathrm{E}{+06}$
ARM	POP	2013	0.40125198	2.89350900E + 06
ASM	URB	2013	0.05797582	4.83100000E+04
ASM	POP	2013	0.13931989	5.53070000E+04
ATG	URB	2013	0.38383110	$2.47980000\mathrm{E}{+04}$
ATG	POP	2013	1.07605830	9.78240000E+04
AUS	URB	2013	1.87536404	1.97902080E+07
AUS	POP	2013	1.75833808	2.31459010E+07
AUT	URB	2013	0.91956020	4.86199100E+06

#### Multi-choice

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

:id	:choice1	:choice2	:choice3
1	A	В	$\overline{C}$
2	$\mathbf{C}$	В	
3	D		
4	В	D	

Step 1 - convert all choices into rows and add artificial column to all values which are not missing.

 $\underline{\text{unnamed } [8\ 4]}$ :

:id	:column :value	:checked	
1	:choice1	A	true
2	:choice1	$\mathbf{C}$	true
3	:choice1	D	true
4	:choice1	В	true
1	:choice2	В	true
2	:choice2	В	$\operatorname{true}$
4	:choice2	D	$\operatorname{true}$
1	:choice3	$\mathbf{C}$	$\operatorname{true}$

Step 2 - Convert back to wide form with actual choices as columns

```
(-> multi2
    (api/drop-columns :$column)
    (api/pivot->wider :$value :checked {:drop-missing? false})
    (api/order-by :id))
```

#### \_unnamed [4 5]:

:id	A	В	С	D
1	true	true	true	
2		${\it true}$	${\it true}$	
3				true
4		true		true

#### Construction

#### construction

data/construction.csv [9 9]:

Year	Month	1 unit	2 to 4 units	5 units or more	Northeast	Midwest	South	West
2018	January	859		348	114	169	596	339
2018	February	882		400	138	160	655	336
2018	March	862		356	150	154	595	330
2018	April	797		447	144	196	613	304
2018	May	875		364	90	169	673	319
2018	June	867		342	76	170	610	360
2018	July	829		360	108	183	594	310
2018	August	939		286	90	205	649	286
2018	September	835		304	117	175	560	296

Conversion 1 - Group two column types

data/construction.csv [63 5]:

Year	Month	:units	region:	:n
2018	January	1		859
2018	February	1		882
2018	March	1		862
2018	April	1		797
2018	May	1		875
2018	June	1		867
2018	July	1		829
2018	August	1		939
2018	September	1		835
2018	January	2-4		
2018	February	2-4		
2018	March	2-4		
2018	April	2-4		
2018	May	2-4		
2018	June	2-4		
2018	July	2-4		
2018	August	2-4		
2018	September	2-4		
2018	January	5+		348
2018	February	5+		400
2018	March	5+		356
2018	April	5+		447
2018	May	5+		364
2018	June	5+		342
2018	July	5+		360

Conversion 2 - Convert to longer form and back and rename columns

data/construction.csv [9 9]:

Year	Month	Midwest	5 units or more	2 to 4 units	Northeast	South	1 unit	West
2018	January	169	348		114	596	859	339
2018	February	160	400		138	655	882	336
2018	March	154	356		150	595	862	330
2018	April	196	447		144	613	797	304
2018	May	169	364		90	673	875	319
2018	June	170	342		76	610	867	360
2018	July	183	360		108	594	829	310
2018	August	205	286		90	649	939	286
2018	September	175	304		117	560	835	296

Various operations on stocks, examples taken from gather and spread manuals.

(def stocks-tidyr (api/dataset "data/stockstidyr.csv"))

## stocks-tidyr

data/stockstidyr.csv [10 4]:

time	X	Y	Z
2009-01-01	1.30989806	-1.89040193	-1.77946880
2009-01-02	-0.29993804	-1.82473090	2.39892513
2009-01-03	0.53647501	-1.03606860	-3.98697977
2009-01-04	-1.88390802	-0.52178390	-2.83065490
2009-01-05	-0.96052361	-2.21683349	1.43715171
2009-01-06	-1.18528966	-2.89350924	3.39784140
2009-01-07	-0.85207056	-2.16794818	-1.20108258
2009-01-08	0.25234172	-0.32854117	-1.53160473
2009-01-09	0.40257136	1.96407898	-6.80878830
2009-01-10	-0.64383500	2.68618382	-2.55909321

## Convert to longer form

#### stocks-long

data/stockstidyr.csv [30 3]:

time	:stocks	:price
2009-01-01	X	1.30989806
2009-01-02	X	-0.29993804
2009-01-03	X	0.53647501
2009-01-04	X	-1.88390802
2009-01-05	X	-0.96052361
2009-01-06	X	-1.18528966
2009-01-07	X	-0.85207056
2009-01-08	X	0.25234172
2009-01-09	X	0.40257136
2009-01-10	X	-0.64383500
2009-01-01	Y	-1.89040193

time	:stocks	:price
2009-01-02	Y	-1.82473090
2009-01-03	Y	-1.03606860
2009-01-04	Y	-0.52178390
2009-01-05	Y	-2.21683349
2009-01-06	Y	-2.89350924
2009-01-07	Y	-2.16794818
2009-01-08	Y	-0.32854117
2009-01-09	Y	1.96407898
2009-01-10	Y	2.68618382
2009-01-01	$\mathbf{Z}$	-1.77946880
2009-01-02	$\mathbf{Z}$	2.39892513
2009-01-03	$\mathbf{Z}$	-3.98697977
2009-01-04	Z	-2.83065490
2009-01-05	Z	1.43715171

Convert back to wide form

```
(api/pivot->wider stocks-long :stocks :price)
```

data/stockstidyr.csv [10 4]:

time	Z	X	Y
2009-01-01	-1.77946880	1.30989806	-1.89040193
2009-01-02	2.39892513	-0.29993804	-1.82473090
2009-01-03	-3.98697977	0.53647501	-1.03606860
2009-01-04	-2.83065490	-1.88390802	-0.52178390
2009-01-05	1.43715171	-0.96052361	-2.21683349
2009-01-06	3.39784140	-1.18528966	-2.89350924
2009-01-07	-1.20108258	-0.85207056	-2.16794818
2009-01-08	-1.53160473	0.25234172	-0.32854117
2009-01-09	-6.80878830	0.40257136	1.96407898
2009-01-10	-2.55909321	-0.64383500	2.68618382

Convert to wide form on time column (let's limit values to a couple of rows)

```
(-> stocks-long
    (api/select-rows (range 0 30 4))
    (api/pivot->wider "time" :price {:drop-missing? false}))
```

data/stockstidyr.csv [3 6]:

:stocks	2009-01-05	2009-01-07	2009-01-01	2009-01-03	2009-01-09
X	-0.96052361		1.30989806		0.40257136
$\mathbf{Z}$	1.43715171		-1.77946880		-6.80878830
Y		-2.16794818		-1.0360686	

## Join/Concat Datasets

Dataset join and concatenation functions.

Joins accept left-side and right-side datasets and columns selector. Options are the same as in

#### tech.ml.dataset functions.

The difference between tech.ml.dataset join functions are: arguments order (first datasets) and possibility to join on multiple columns.

Additionally set operations are defined: intersect and difference.

To concat two datasets rowwise you can choose:

- concat concats rows for matching columns, the number of columns should be equal.
- union like concat but returns unique values
- bind concats rows add missing, empty columns

To add two datasets columnwise use bind. The number of rows should be equal.

Datasets used in examples:

 $\underline{\phantom{a}}$ unnamed [9 3]:

:a	:b	:c
1	101	a
2	102	b
1	103	$\mathbf{s}$
2	104	
3	105	$\mathbf{t}$
4	106	r
	107	a
	108	$\mathbf{c}$
4	109	$\mathbf{t}$
_		

\_unnamed [9 4]:

:a	:b	:c	:d
	110	d	X
1	109	$\mathbf{a}$	Χ
2	108	$\mathbf{t}$	X
5	107	$\mathbf{a}$	X
4	106	$\mathbf{t}$	Χ
3	105	a	X
2	104	b	X
1	103	1	X
	102	e	Χ

Left

## (api/left-join ds1 ds2 :b)

left-outer-join [9 7]:

:b	:a	:c	:right.b	:right.a	:right.c	:d
109	4	t	109	1	a	X
108		$\mathbf{c}$	108	2	t	X
107		a	107	5	a	X
106	4	$\mathbf{r}$	106	4	t	X
105	3	$\mathbf{t}$	105	3	a	X
104	2		104	2	b	X
103	1	$\mathbf{S}$	103	1	1	X
102	2	b	102		e	X
101	1	$\mathbf{a}$				

# (api/left-join ds2 ds1 :b)

left-outer-join [9 7]:

:b	:a	:c	:d	:right.b	:right.a	:right.c
102		е	X	102	2	b
103	1	1	X	103	1	S
104	2	b	X	104	2	
105	3	$\mathbf{a}$	X	105	3	$\mathbf{t}$
106	4	$\mathbf{t}$	X	106	4	r
107	5	$\mathbf{a}$	X	107		a
108	2	$\mathbf{t}$	X	108		$\mathbf{c}$
109	1	$\mathbf{a}$	X	109	4	$\mathbf{t}$
110		d	X			

# (api/left-join ds1 ds2 [:a :b])

left-outer-join [9 7]:

:a	:b	:с	:right.a	:right.b	:right.c	:d
4	106	r	4	106	t	X
3	105	$\mathbf{t}$	3	105	a	X
2	104		2	104	b	X
1	103	$\mathbf{s}$	1	103	1	X
2	102	b				
	108	$\mathbf{c}$				
	107	a				
1	101	a				
4	109	$\mathbf{t}$				

# (api/left-join ds2 ds1 [:a :b])

left-outer-join [9 7]:

:a	:b	:c	:d	:right.a	:right.b	:right.c
1	103	1	X	1	103	s
2	104	b	X	2	104	
3	105	a	X	3	105	t
4	106	$\mathbf{t}$	X	4	106	r
2	108	$\mathbf{t}$	X			
1	109	$\mathbf{a}$	X			
5	107	a	X			
	110	d	X			
	102	$\mathbf{e}$	X			

# Right

(api/right-join ds1 ds2 :b)

right-outer-join [9 7]:

:b	:a	:c	:right.b	:right.a	: right.c	:d
109	4	t	109	1	a	X
108		$\mathbf{c}$	108	2	t	X
107		$\mathbf{a}$	107	5	a	X
106	4	$\mathbf{r}$	106	4	$\mathbf{t}$	X
105	3	$\mathbf{t}$	105	3	a	X
104	2		104	2	b	X
103	1	$\mathbf{S}$	103	1	1	X
102	2	b	102		e	X
			110		d	X

# (api/right-join ds2 ds1 :b)

right-outer-join [9 7]:

:b	:a	:c	:d	$: \!\! right.b$	:right.a	:right.c
102		е	X	102	2	b
103	1	1	X	103	1	S
104	2	b	X	104	2	
105	3	a	X	105	3	$\mathbf{t}$
106	4	$\mathbf{t}$	X	106	4	r
107	5	a	X	107		a
108	2	$\mathbf{t}$	X	108		$^{\mathrm{c}}$
109	1	a	X	109	4	$\mathbf{t}$
				101	1	a

# (api/right-join ds1 ds2 [:a :b])

right-outer-join [9 7]:

4 106					:d
	r	4	106	t	X
3 105	$\mathbf{t}$	3	105	a	X
2 104		2	104	b	X
1 103	$\mathbf{s}$	1	103	1	X
			110	d	X
		1	109	a	Χ
		2	108	t	Χ
		5	107	a	Χ
			102	e	X

# (api/right-join ds2 ds1 [:a :b])

right-outer-join [9 7]:

:a	:b	:c	:d	:right.a	:right.b	:right.c
1	103	1	X	1	103	s
2	104	b	X	2	104	
3	105	$\mathbf{a}$	X	3	105	$\mathbf{t}$
4	106	$\mathbf{t}$	X	4	106	r
				1	101	a
				2	102	b
					107	a
					108	$\mathbf{c}$
				4	109	t

### Inner

(api/inner-join ds1 ds2 :b)

inner-join [8 6]:

:b	:a	:c	:right.a	:right.c	:d
109	4	t	1	a	X
108		$\mathbf{c}$	2	t	X
107		a	5	a	X
106	4	$\mathbf{r}$	4	$\mathbf{t}$	X
105	3	$\mathbf{t}$	3	a	X
104	2		2	b	X
103	1	$\mathbf{s}$	1	1	X
102	2	b		e	X

# (api/inner-join ds2 ds1 :b)

inner-join [8 6]:

:b	:a	:c	:d	:right.a	:right.c
102		е	X	2	b
103	1	1	X	1	S
104	2	b	X	2	
105	3	$\mathbf{a}$	X	3	t
106	4	$\mathbf{t}$	X	4	r
107	5	$\mathbf{a}$	X		a
108	2	$\mathbf{t}$	X		$\mathbf{c}$
109	1	a	X	4	$\mathbf{t}$

(api/inner-join ds1 ds2 [:a :b])

inner-join [4 7]:

:a	:b	:c	:right.a	:right.b	:right.c	:d
4	106	r	4	106	t	X
3	105	$\mathbf{t}$	3	105	a	X
2	104		2	104	b	X
1	103	$\mathbf{s}$	1	103	1	X

(api/inner-join ds2 ds1 [:a :b])

inner-join [4 7]:

:a	:b	:c	:d	:right.a	:right.b	:right.c
1	103	1	X	1	103	s
2	104	b	X	2	104	
3	105	a	X	3	105	$\mathbf{t}$
4	106	$\mathbf{t}$	X	4	106	$\mathbf{r}$

# Full

Join keeping all rows

(api/full-join ds1 ds2 :b)

full-join [10 7]:

:b	:a	:c	:right.b	:right.a	:right.c	:d
109	4	t	109	1	a	X
108		$\mathbf{c}$	108	2	$\mathbf{t}$	X
107		a	107	5	a	X
106	4	r	106	4	$\mathbf{t}$	X
105	3	$\mathbf{t}$	105	3	a	X
104	2		104	2	b	X
103	1	$\mathbf{s}$	103	1	1	X

:b	:a	:c	:right.b	:right.a	:right.c	:d
-			102		e	X
101	1	a				
			110		d	X

# (api/full-join ds2 ds1 :b)

full-join [10 7]:

:b	:a	:c	:d	:right.b	:right.a	:right.c
102		e	X	102	2	b
103	1	1	X	103	1	S
104	2	b	X	104	2	
105	3	a	X	105	3	t
106	4	$\mathbf{t}$	X	106	4	r
107	5	$\mathbf{a}$	X	107		a
108	2	$\mathbf{t}$	X	108		$\mathbf{c}$
109	1	a	X	109	4	t
110		d	X			
				101	1	a

# (api/full-join ds1 ds2 [:a :b])

full-join [14 7]:

:a	:b	:c	:right.a	:right.b	:right.c	:d
4	106	r	4	106	t	X
3	105	$\mathbf{t}$	3	105	a	X
2	104		2	104	b	X
1	103	$\mathbf{s}$	1	103	1	X
2	102	b				
	108	$^{\mathrm{c}}$				
	107	$\mathbf{a}$				
1	101	$\mathbf{a}$				
4	109	$\mathbf{t}$				
				110	d	X
			1	109	a	X
			2	108	$\mathbf{t}$	X
			5	107	a	X
				102	e	X

# (api/full-join ds2 ds1 [:a :b])

full-join [14 7]:

:a	:b	:c	:d	:right.a	:right.b	:right.c
1	103	1	X	1	103	s
2	104	b	X	2	104	
3	105	$\mathbf{a}$	X	3	105	$\mathbf{t}$
4	106	$\mathbf{t}$	X	4	106	r
2	108	$\mathbf{t}$	X			
1	109	a	X			
5	107	a	X			
	110	d	X			
	102	e	X			
				1	101	a
				2	102	b
					107	a
					108	$\mathbf{c}$
				4	109	t

## $\mathbf{Semi}$

Return rows from ds1 matching ds2  $\,$ 

(api/semi-join ds1 ds2 :b)

semi-join [5 3]:

:b	:a	:с
109	4	t
106	4	r
105	3	$\mathbf{t}$
104	2	
103	1	$\mathbf{s}$

(api/semi-join ds2 ds1 :b)

semi-join [5 4]:

:b	:a	:c	:d
103	1	1	X
104	2	b	X
105	3	a	X
106	4	$\mathbf{t}$	X
109	1	a	X

(api/semi-join ds1 ds2 [:a :b])

semi-join [4 3]:

:a	:b	:0
4	106	r
3	105	$\mathbf{t}$

:a	:b	:с
2	104	
1	103	$\mathbf{s}$

(api/semi-join ds2 ds1 [:a :b])

semi-join  $[4 \ 4]$ :

:a	:b	:c	:d
1	103	1	X
2	104	b	Χ
3	105	a	Χ
4	106	$\mathbf{t}$	X

## Anti

Return rows from ds1 not matching ds2  $\,$ 

(api/anti-join ds1 ds2 :b)

anti-join  $[4\ 3]$ :

:b	:a	:с
108		c
107		a
102	2	b
101	1	a

(api/anti-join ds2 ds1 :b)

anti-join [4 4]:

:b	:a	:c	:d
102		е	X
107	5	a	X
108	2	$\mathbf{t}$	X
110		d	X

(api/anti-join ds1 ds2 [:a :b])

anti-join  $[5 \ 3]$ :

:a	:b	:0
2	102	b
	108	$\mathbf{c}$
	107	a

```
:a :b :c

1 101 a

4 109 t
```

```
(api/anti-join ds2 ds1 [:a :b])
```

anti-join [5 4]:

:a	:b	:c	:d
2	108	t	X
1	109	a	X
5	107	a	Χ
	110	d	Χ
	102	e	X

asof

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

:a	:left-val
1	a
5	b
10	$\mathbf{c}$

 $\underline{\phantom{a}}$ unnamed [5 2]:

:a	:right-val
1	:a
2	:b
3	:c
6	:d
7	<b>:</b> e

# (api/asof-join left-ds right-ds :a)

 $asof-<=[3\ 4]:$ 

:a	: left-val	: right. a	:right-val
1	a	1	:a
5	b	6	:d

:a	:left-val	:right.a	:right-val
10	c		

(api/asof-join left-ds right-ds :a {:asof-op :nearest})

asof-nearest [3 4]:

:a	:left-val	:right.a	:right-val
1	a	1	:a
5	b	6	:d
10	$\mathbf{c}$	7	<b>:</b> e

(api/asof-join left-ds right-ds :a {:asof-op :>=})

 $asof->= [3 \ 4]:$ 

:a	:left-val	:right.a	:right-val
1	a	1	:a
5	b	3	:c
10	$\mathbf{c}$	7	<b>:</b> e

### Concat

 ${\tt contact}$  joins rows from other datasets

(api/concat ds1)

\_unnamed [9 3]:

:a	:b	:0
1	101	a
2	102	b
1	103	$\mathbf{s}$
2	104	
3	105	$\mathbf{t}$
4	106	r
	107	a
	108	$\mathbf{c}$
4	109	$\mathbf{t}$

 ${\tt concat-copying} \ {\rm ensures} \ {\rm all} \ {\rm readers} \ {\rm are} \ {\rm evaluated}.$ 

(api/concat-copying ds1)

 $\underline{\phantom{a}}$ unnamed [9 3]:

:a	:b	:0
1	101	a
2	102	b

:a	:b	:с
1	103	s
2	104	
3	105	$\mathbf{t}$
4	106	r
	107	a
	108	$\mathbf{c}$
4	109	$\mathbf{t}$

# (api/concat ds1 (api/drop-columns ds2 :d))

null [18 3]:

:a	:b	:с
1	101	a
2	102	b
1	103	$\mathbf{s}$
2	104	
3	105	$\mathbf{t}$
4	106	r
	107	a
	108	$\mathbf{c}$
4	109	$\mathbf{t}$
	110	d
1	109	a
2	108	$\mathbf{t}$
5	107	a
4	106	$\mathbf{t}$
3	105	a
2	104	b
1	103	1
	102	е

# (apply api/concat (repeatedly 3 #(api/random DS)))

null [27 4]:

:V1	:V2	:V3	:V4
1	5	1.0	В
1	1	0.5	A
2	6	1.5	$\mathbf{C}$
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
2	8	1.0	В
1	5	1.0	В
2	2	1.0	В
1	9	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
2	2	1.0	В

:V1	:V2	:V3	:V4
1	3	1.5	С
1	3	1.5	$\mathbf{C}$
2	8	1.0	В
2	4	0.5	A
1	1	0.5	A
1	5	1.0	В
2	4	0.5	A
2	4	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$
1	7	0.5	A
1	3	1.5	$\mathbf{C}$
1	5	1.0	В
1	7	0.5	A

# Concat grouped dataset

Concatenation of grouped datasets results also in grouped dataset.

null [6 3]:

:name	:group-id	:data
{:V3 1.0}	0	Group: {:V3 1.0} [3 4]:
$\{: V3 \ 0.5\}$	1	Group: {:V3 0.5} [3 4]:
$\{: V3 \ 1.5\}$	2	Group: {:V3 1.5} [3 4]:
{:V4 "B"}	3	Group: {:V4 "B"} [3 4]:
{:V4 "C"}	4	Group: {:V4 "C"} [3 4]:
{:V4 "A"}	5	Group: {:V4 "A"} [3 4]:

## Union

The same as concat but returns unique rows

```
(apply api/union (api/drop-columns ds2 :d) (repeat 10 ds1))
```

union [18 3]:

:a	:b	:0
	110	d
1	109	a
2	108	$\mathbf{t}$
5	107	a
4	106	$\mathbf{t}$
3	105	a
2	104	b
1	103	1
	102	e
1	101	a
2	102	b

:a	:b	:c
1	103	s
2	104	
3	105	$\mathbf{t}$
4	106	r
	107	a
	108	$\mathbf{c}$
4	109	$\mathbf{t}$

(apply api/union (repeatedly 10 #(api/random DS)))

union [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	6	1.5	$^{\mathrm{C}}$
1	3	1.5	$\mathbf{C}$
2	8	1.0	В
1	7	0.5	A
1	9	1.5	$\mathbf{C}$
1	5	1.0	В
2	2	1.0	В
2	4	0.5	A

# Bind

 $\verb|bind| adds| empty| columns| during| concat|$ 

(api/bind ds1 ds2)

null [18 4]:

:a	:b	:c	:d
1	101	a	
2	102	b	
1	103	$\mathbf{s}$	
2	104		
3	105	$\mathbf{t}$	
4	106	$\mathbf{r}$	
	107	a	
	108	$\mathbf{c}$	
4	109	$\mathbf{t}$	
	110	d	Χ
1	109	a	X
2	108	$\mathbf{t}$	X
5	107	a	X
4	106	$\mathbf{t}$	X
3	105	a	X
2	104	b	X
1	103	1	X
	102	e	X

# (api/bind ds2 ds1)

null [18 4]:

	1		
:a	:b	:c	:d
	110	d	X
1	109	a	X
2	108	$\mathbf{t}$	X
5	107	a	X
4	106	$\mathbf{t}$	X
3	105	a	X
2	104	b	Χ
1	103	1	X
	102	e	X
1	101	a	
2	102	b	
1	103	$\mathbf{S}$	
2	104		
3	105	$\mathbf{t}$	
4	106	$\mathbf{r}$	
	107	a	
	108	$\mathbf{c}$	
4	109	t	

# Append

append concats columns

```
(api/append ds1 ds2)
```

\_unnamed [9 7]:

:a	:b	:c	:a	:b	:c	:d
1	101	a		110	d	X
2	102	b	1	109	$\mathbf{a}$	X
1	103	$\mathbf{s}$	2	108	$\mathbf{t}$	X
2	104		5	107	$\mathbf{a}$	X
3	105	$\mathbf{t}$	4	106	$\mathbf{t}$	X
4	106	$\mathbf{r}$	3	105	$\mathbf{a}$	X
	107	a	2	104	b	X
	108	$\mathbf{c}$	1	103	1	X
4	109	t		102	e	X

# Intersection

intersection [8 1]:

#### Difference

```
(api/difference (api/select-columns ds1 :b)
(api/select-columns ds2 :b))

difference [1 1]:

:b
101

(api/difference (api/select-columns ds2 :b)
(api/select-columns ds1 :b))

difference [1 1]:

:b
110
```

### Split into train/test

In ML world very often you need to test given model and prepare collection of train and test datasets. split creates new dataset with two additional columns:

- :\$split-name with :train, :test, :split-2, ... values
- :\$split-id id of splitted group (for k-fold and repeating)

split-type can be one of the following:

- :kfold (default) k-fold strategy, :k defines number of folds (defaults to 5), produces k splits
- :bootstrap :ratio defines ratio of observations put into result (defaults to 1.0), produces 1 split
- :holdout split into two parts with given ratio (defaults to 2/3), produces 1 split
- :loo leave one out, produces the same number of splits as number of observations

:holdout can accept also probabilites or ratios and can split to more than 2 subdatasets

Additionally you can provide:

- :seed for random number generator
- :repeats repeat procedure :repeats times
- :partition-selector same as in group-by for stratified splitting to reflect dataset structure in splits.
- :split-names names of subdatasets different than default, ie. [:train :test :split-2 ...]

- :split-col-name a column where name of split is stored, either :train or :test values (default: :\$split-name)
- :split-id-col-name a column where id of the train/test pair is stored (default: :\$split-id)

In case of grouped dataset each group is processed separately.

See more

### for-splitting

\_unnamed [25 3]:

:id	:partition	:group
0	:a	:g2
1	:a	:g3
2	:a	:g3
3	:a	:g1
4	:a	:g3
5	:a	:g3
6	:a	:g1
7	:a	:g2
8	:a	:g2
9	:a	:g2
10	:a	:g2
11	:a	:g2
12	:a	:g1
13	:a	:g2
14	:a	:g2
15	:a	:g3
16	:a	:g2
17	:a	:g2
18	:a	:g3
19	:a	:g3
20	:b	:g1
21	:b	:g3
22	:b	:g1
23	:b	:g2
24	:b	:g1

## k-Fold

Returns  $k=5~\mathrm{maps}$ 

```
(-> for-splitting
    (api/split)
    (api/head 30))
```

\_unnamed, (splitted) [30 5]:

:id	:partition	:group	:split-name :split-id	_
21	:b	:g3	:train	0
11	:a	:g2	:train	0
13	:a	:g2	:train	0
23	:b	:g2	:train	0
19	:a	:g3	:train	0
16	:a	:g2	:train	0
6	:a	:g1	:train	0
24	:b	:g1	:train	0
7	:a	:g2	:train	0
1	:a	:g3	:train	0
22	:b	:g1	:train	0
4	:a	:g3	:train	0
15	:a	:g3	:train	0
8	:a	:g2	:train	0
14	:a	:g2	:train	0
12	:a	:g1	:train	0
18	:a	:g3	:train	0
10	:a	:g2	:train	0
3	:a	:g1	:train	0
20	:b	:g1	:train	0
2	:a	:g3	:test	0
17	:a	:g2	:test	0
9	:a	:g2	:test	0
5	:a	:g3	:test	0
0	:a	:g2	:test	0
2	:a	:g3	:train	1
17	:a	:g2	:train	1
9	:a	:g2	:train	1
5	:a	:g3	:train	1
0	:a	:g2	:train	1

Partition according to : k column to reflect it's distribution

```
(-> for-splitting
   (api/split :kfold {:partition-selector :partition})
   (api/head 30))
```

\_unnamed, (splitted) [30 5]:

:id	:partition	:group	:split-name :split-id	
$\overline{24}$	:b	:g1	:train	0
21	:b	:g3	:train	0
22	:b	:g1	:train	0
20	:b	:g1	:train	0
23	:b	:g2	:test	0
23	:b	:g2	:train	1
21	:b	:g3	:train	1
22	:b	:g1	:train	1
20	:b	:g1	:train	1
24	:b	:g1	:test	1
23	:b	:g2	:train	2
24	:b	:g1	:train	2

:id	:partition	:group	:split-name :split-id	
22	:b	:g1	:train	2
20	:b	:g1	:train	2
21	:b	:g3	:test	2
23	:b	:g2	:train	3
24	:b	:g1	:train	3
21	:b	:g3	:train	3
20	:b	:g1	:train	3
22	:b	:g1	:test	3
23	:b	:g2	:train	4
24	:b	:g1	:train	4
21	:b	:g3	:train	4
22	:b	:g1	:train	4
20	:b	:g1	:test	4
6	:a	:g1	:train	0
0	:a	:g2	:train	0
8	:a	:g2	:train	0
2	:a	:g3	:train	0
19	:a	:g3	:train	0

# Bootstrap

(api/split for-splitting :bootstrap)

 $\_$ unnamed, (splitted) [35 5]:

:id	:partition	:group	:split-name :split-id	
9	:a	:g2	:train	0
16	:a	:g2	:train	0
6	:a	:g1	:train	0
2	:a	:g3	:train	0
19	:a	:g3	:train	0
8	:a	:g2	:train	0
22	:b	:g1	:train	0
24	:b	:g1	:train	0
24	:b	:g1	:train	0
22	:b	:g1	:train	0
24	:b	:g1	:train	0
1	:a	:g3	:train	0
8	:a	:g2	:train	0
21	:b	:g3	:train	0
5	:a	:g3	:train	0
3	:a	:g1	:train	0
24	:b	:g1	:train	0
20	:b	:g1	:train	0
3	:a	:g1	:train	0
15	:a	:g3	:train	0
21	:b	:g3	:train	0
22	:b	:g1	:train	0
16	:a	:g2	:train	0
11	:a	:g2	:train	0
5	:a	:g3	:train	0

with repeats, to get 100 splits

```
(-> for-splitting
   (api/split :bootstrap {:repeats 100})
   (:$split-id)
   (distinct)
   (count))
```

100

### Holdout

with small ratio

```
(api/split for-splitting :holdout {:ratio 0.2})
```

\_unnamed, (splitted) [25 5]:

$:\! \mathrm{id}$	: partition	:group	:split-name :split-id	
24	:b	:g1	:train	0
1	:a	:g3	:train	0
6	:a	:g1	:train	0
3	:a	:g1	:train	0
9	:a	:g2	:train	0
23	:b	:g2	:test	0
12	:a	:g1	:test	0
21	:b	:g3	:test	0
11	:a	:g2	:test	0
22	:b	:g1	:test	0
4	:a	:g3	:test	0
19	:a	:g3	:test	0
15	:a	:g3	:test	0
8	:a	:g2	:test	0
7	:a	:g2	:test	0
17	:a	:g2	:test	0
10	:a	:g2	:test	0
16	:a	:g2	:test	0
5	:a	:g3	:test	0
18	:a	:g3	:test	0
2	:a	:g3	:test	0
0	:a	:g2	:test	0
14	:a	:g2	:test	0
13	:a	:g2	:test	0
20	:b	:g1	:test	0

you can split to more than two subdatasets with holdout

```
(api/split for-splitting :holdout {:ratio [0.1 0.2 0.3 0.15 0.25]})
```

\_unnamed, (splitted) [25 5]:

:id	:partition	:group	:split-name :split-id	
15	:a	:g3	:train	0
13	:a	:g2	:train	0
9	:a	:g2	:test	0

:id	:partition	:group	:split-name :split-id	_
0	:a	:g2	:test	0
21	:b	:g3	:test	0
17	:a	:g2	:test	0
5	:a	:g3	:test	0
4	:a	:g3	:split-2	0
12	:a	:g1	:split-2	0
22	:b	:g1	:split-2	0
19	:a	:g3	:split-2	0
7	:a	:g2	:split-2	0
24	:b	:g1	:split-2	0
14	:a	:g2	:split-2	0
18	:a	:g3	:split-3	0
2	:a	:g3	:split-3	0
23	:b	:g2	:split-3	0
11	:a	:g2	:split-4	0
3	:a	:g1	:split-4	0
1	:a	:g3	:split-4	0
6	:a	:g1	:split-4	0
10	:a	:g2	:split-4	0
16	:a	:g2	:split-4	0
20	:b	:g1	:split-4	0
8	:a	:g2	:split-4	0

you can use also proportions with custom names

\_unnamed, (splitted) [25 5]:

:id	: partition	:group	:split-name :split-id	
3	:a	:g1	small	0
2	:a	:g3	$\operatorname{small}$	0
0	:a	:g2	small	0
15	:a	:g3	small	0
13	:a	:g2	$\operatorname{small}$	0
20	:b	:g1	$\operatorname{smaller}$	0
7	:a	:g2	smaller	0
12	:a	:g1	smaller	0
4	:a	:g3	big	0
18	:a	:g3	big	0
21	:b	:g3	big	0
5	:a	:g3	big	0
8	:a	:g2	big	0
9	:a	:g2	big	0
11	:a	:g2	big	0
10	:a	:g2	big	0
22	:b	:g1	big	0
19	:a	:g3	big	0
16	:a	:g2	big	0
23	:b	:g2	big	0
1	:a	:g3	big	0

:id	:partition	:group	:split-name :split-id	
14	:a	:g2	the rest	0
6	:a	:g1	the rest	0
17	:a	:g2	the rest	0
24	:b	:g1	the rest	0

# Leave One Out

```
(-> for-splitting
   (api/split :loo)
   (api/head 30))
```

\_unnamed, (splitted) [30 5]:

:id	:partition	:group	:split-name :split-id	
9	:a	:g2	:train	0
14	:a	:g2	:train	0
21	:b	:g3	:train	0
19	:a	:g3	:train	0
1	:a	:g3	:train	0
23	:b	:g2	:train	0
20	:b	:g1	:train	0
16	:a	:g2	:train	0
4	:a	:g3	:train	0
6	:a	:g1	:train	0
17	:a	:g2	:train	0
3	:a	:g1	:train	0
13	:a	:g2	:train	0
18	:a	:g3	:train	0
15	:a	:g3	:train	0
7	:a	:g2	:train	0
10	:a	:g2	:train	0
12	:a	:g1	:train	0
24	:b	:g1	:train	0
5	:a	:g3	:train	0
0	:a	:g2	:train	0
2	:a	:g3	:train	0
8	:a	:g2	:train	0
22	:b	:g1	:train	0
11	:a	:g2	:test	0
11	:a	:g2	:train	1
14	:a	:g2	:train	1
21	:b	:g3	:train	1
19	:a	:g3	:train	1
1	:a	:g3	:train	1

```
(-> for-splitting
   (api/split :loo)
   (api/row-count))
```

625

# Grouped dataset with partitioning

```
(-> for-splitting
   (api/group-by :group)
   (api/split :bootstrap {:partition-selector :partition :seed 11 :ratio 0.8}))
```

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

:name	:group-id	:data
:g1	0	Group: :g1, (splitted) [9 5]:
:g3	1	Group: :g3, (splitted) [9 5]:
:g2	2	Group: :g2, (splitted) [14 5]:

# Split as a sequence

To get a sequence of pairs, use split->seq function

```
(-> for-splitting
   (api/split->seq :kfold {:partition-selector :partition})
   (first))
```

{:train Group: 0 [20 3]:

:id	:partition	:group
23	:b	:g2
22	:b	:g1
24	:b	:g1
20	:b	:g1
17	:a	:g2
11	:a	:g2
5	:a	:g3
12	:a	:g1
2	:a	:g3
15	:a	:g3
14	:a	:g2
9	:a	:g2
19	:a	:g3
13	:a	:g2
4	:a	:g3
18	:a	:g3
6	:a	:g1
7	:a	:g2
3	:a	:g1
1	:a	:g3

, :test Group: 0 [5 3]:

:id	: partition	:group
21	:b	:g3
8	:a	:g2
0	:a	:g2
16	:a	:g2

:id	:partition	:group
10	:a	:g2

:id	:partition	:group
24	:b	:g1
24	:b	:g1
24	:b	:g1
3	:a	:g1
12	:a	:g1
6	:a	:g1

, :test Group: 0 [2 3]:

:id	: partition	:group
20	:b	:g1
22	:b	:g1

} {:train Group: 1 [6 3]:

:id	:partition	:group
20	:b	:g1
22	:b	:g1
20	:b	:g1
3	:a	:g1
6	:a	:g1
12	:a	:g1

, :test Group: 1 [1 3]:

:id	:partition	:group
24	:b	:g1

})]

# Pipeline

tablecloth.pipeline exports special versions of API which create functions operating only on dataset. This creates the possibility to chain operations and compose them easily.

There are two ways to create pipelines:

- functional, as a composition of functions
- declarative, separating task declarations and concrete parametrization.

Pipeline operations are prepared to work with metamorph library. That means that result of the pipeline is wrapped into a map and dataset is stored under :metamorph/data key.

```
(require '[tablecloth.pipeline :as pip])
```

#### **Functional**

To create composable function, call API function but defined in tablecloth.pipeline namespace and without ds argument.

```
(pip/select-columns :type/numerical)
```

#function[tablecloth.pipeline/select-columns/fn-39721]

Calling such function on a dataset gives a requested result.

```
((pip/select-columns :type/numerical) DS)
```

#:metamorph{:data \_unnamed [9 3]:

:V1	:V2	:V3
1	1	0.5
2	2	1.0
1	3	1.5
2	4	0.5
1	5	1.0
2	6	1.5
1	7	0.5
2	8	1.0
1	9	1.5

}

Pipeline functions can be composed using a comp function or pipeline. The latter is just reversed comp to create the order from first operation to last.

#:metamorph{:data \_unnamed [6 4]:

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

}

#### Declarative

To create a pipeline declarative way you can use ->pipeline function and apply sequence of definition. Definition is simple a sequence where on the first position name of the operation is passed as a keyword, a symbol or var. The rest are operation parameters which are mostly the same as when calling a function.

To help keeping pipeline declaration as a pure data structure you can use namespaced keywords to create reference to defined symbols. The special ctx namespace is used to refer to a values stored in an optional map passed as an argument to ->pipeline.

You can't used type and !type namespaces for a symbol reference, because they are reserved for column type selectors.

#:metamorph $\{:$ data \_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"]

```
{
(pipeline-2 DS)
```

#:metamorph{:data \_unnamed [6 5]:

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

}

#### Custom operator

Custom pipeline operator is just function which returns another function operating on a dataset. If you want it to work with metamorph you have to use metamorph/lift function.

\_unnamed [9 7]:

:V1	:V2	:V3	:V4	:V1-copy	:V2-copy	:V3-copy
1	1	0.5	A	1	1	0.5
2	2	1.0	В	2	2	1.0
1	3	1.5	$\mathbf{C}$	1	3	1.5
2	4	0.5	A	2	4	0.5
1	5	1.0	В	1	5	1.0
2	6	1.5	$\mathbf{C}$	2	6	1.5
1	7	0.5	A	1	7	0.5
2	8	1.0	В	2	8	1.0
1	9	1.5	$\mathbf{C}$	1	9	1.5

# **Functions**

This API doesn't provide any statistical, numerical or date/time functions. Use below namespaces:

Namespace	functions
tech.v3.datatype.functional tech.v3.datatype.datetime	primitive oprations, reducers, statistics date/time converters and operations

# Other examples

## Stocks

```
(defonce stocks (api/dataset "https://raw.githubusercontent.com/techascent/tech.ml.dataset/master/test/stocks
```

https://raw.githubusercontent.com/techascent/tech.ml.dataset/master/test/data/stocks.csv [560 3]:

:symbol	:date	:price
MSFT	2000-01-01	39.81
MSFT	2000-02-01	36.35
MSFT	2000-03-01	43.22
MSFT	2000-04-01	28.37
MSFT	2000-05-01	25.45

:symbol	:date	:price
MSFT	2000-06-01	32.54
MSFT	2000-07-01	28.40
MSFT	2000-08-01	28.40
MSFT	2000-09-01	24.53
MSFT	2000-10-01	28.02
MSFT	2000-11-01	23.34
MSFT	2000-12-01	17.65
MSFT	2001-01-01	24.84
MSFT	2001-02-01	24.00
MSFT	2001-03-01	22.25
MSFT	2001-04-01	27.56
MSFT	2001-05-01	28.14
MSFT	2001-06-01	29.70
MSFT	2001-07-01	26.93
MSFT	2001-08-01	23.21
MSFT	2001-09-01	20.82
MSFT	2001-10-01	23.65
MSFT	2001-11-01	26.12
MSFT	2001-12-01	26.95
MSFT	2002-01-01	25.92

# \_unnamed [51 3]:

:symbol	:year	:summary
AAPL	2000	21.74833333
AAPL	2001	10.17583333
AAPL	2002	9.40833333
AAPL	2003	9.34750000
AAPL	2004	18.723333333
AAPL	2005	48.17166667
AAPL	2006	72.043333333
AAPL	2007	133.35333333
AAPL	2008	138.48083333
AAPL	2009	150.393333333
AAPL	2010	206.56666667
AMZN	2000	43.93083333
AMZN	2001	11.73916667
AMZN	2002	16.723333333
AMZN	2003	39.01666667
AMZN	2004	43.26750000
AMZN	2005	40.18750000
AMZN	2006	36.25166667
AMZN	2007	69.95250000
AMZN	2008	69.01500000

:symbol	:year	:summary
AMZN	2009	90.73083333
AMZN	2010	124.21000000
GOOG	2004	159.47600000
GOOG	2005	286.47250000
GOOG	2006	415.25666667

\_unnamed [51 3]:

:symbol	:year	:summary
AMZN	2007	69.95250000
AMZN	2008	69.01500000
AMZN	2009	90.73083333
AMZN	2010	124.21000000
AMZN	2000	43.93083333
AMZN	2001	11.73916667
AMZN	2002	16.723333333
AMZN	2003	39.01666667
AMZN	2004	43.26750000
AMZN	2005	40.18750000
AMZN	2006	36.25166667
$_{\rm IBM}$	2001	96.96833333
$_{\rm IBM}$	2002	75.12500000
IBM	2000	96.91416667
MSFT	2006	24.758333333
MSFT	2005	23.84583333
MSFT	2004	22.67416667
MSFT	2003	20.93416667
AAPL	2001	10.17583333
MSFT	2010	28.50666667
AAPL	2002	9.40833333
MSFT	2009	22.87250000
MSFT	2008	25.20833333
AAPL	2000	21.74833333
MSFT	2007	29.28416667

## data.table

Below you can find comparizon between functionality of data.table and Clojure dataset API. I leave it without comments, please refer original document explaining details:

Introduction to data.table

 $\mathbf{R}$ 

```
library(data.table)
library(knitr)
```

flights <- fread("https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.cs
kable(head(flights))</pre>

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
2014	1	1	14	13	AA	JFK	LAX	359	2475	9
2014	1	1	-3	13	AA	$_{ m JFK}$	LAX	363	2475	11
2014	1	1	2	9	AA	$_{ m JFK}$	LAX	351	2475	19
2014	1	1	-8	-26	AA	LGA	PBI	157	1035	7
2014	1	1	2	1	AA	$_{ m JFK}$	LAX	350	2475	13
2014	1	1	4	0	AA	EWR	LAX	339	2454	18

(api/head flights 6)

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [6 11]:

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
${2014}$	1	1	14	13	AA	JFK	LAX	359	2475	9
2014	1	1	-3	13	AA	$_{ m JFK}$	LAX	363	2475	11
2014	1	1	2	9	AA	$_{ m JFK}$	LAX	351	2475	19
2014	1	1	-8	-26	AA	LGA	PBI	157	1035	7
2014	1	1	2	1	AA	JFK	LAX	350	2475	13
2014	1	1	4	0	AA	EWR	LAX	339	2454	18

### Basics

### Shape of loaded data

 $\mathbf{R}$ 

```
dim(flights)
```

[1] 253316 11

Clojure

(api/shape flights)

[253316 11]

What is data.table?

 $\mathbf{R}$ 

```
DT = data.table(
   ID = c("b","b","b","a","a","c"),
   a = 1:6,
   b = 7:12,
   c = 13:18
)
kable(DT)
```

```
ID a
        b
             ^{\mathrm{c}}
b
    1
        7 13
    2
b
        8 14
    3
        9 15
b
    4 \quad 10 \quad 16
a
    5
       11 17
    6 12 18
```

```
class(DT$ID)
```

[1] "character"

Clojure

DT

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

:ID	:a	:b	:с
b	1	7	13
b	2	8	14
b	3	9	15
a	4	10	16
a	5	11	17
$\mathbf{c}$	6	12	18

```
(-> :ID DT meta :datatype)
```

:string

Get all the flights with "JFK" as the origin airport in the month of June.

R

```
ans <- flights[origin == "JFK" & month == 6L]
kable(head(ans))</pre>
```

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
2014	6	1	-9	-5	AA	JFK	LAX	324	2475	8
2014	6	1	-10	-13	AA	$_{ m JFK}$	LAX	329	2475	12
2014	6	1	18	-1	AA	JFK	LAX	326	2475	7
2014	6	1	-6	-16	AA	$_{ m JFK}$	LAX	320	2475	10
2014	6	1	-4	-45	AA	$_{ m JFK}$	LAX	326	2475	18
2014	6	1	-6	-23	AA	JFK	LAX	329	2475	14

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [6 11]:

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
${2014}$	6	1	-9	-5	AA	JFK	LAX	324	2475	8
2014	6	1	-10	-13	AA	$_{ m JFK}$	LAX	329	2475	12
2014	6	1	18	-1	AA	$_{ m JFK}$	LAX	326	2475	7
2014	6	1	-6	-16	AA	$_{ m JFK}$	LAX	320	2475	10
2014	6	1	-4	-45	AA	$_{ m JFK}$	LAX	326	2475	18
2014	6	1	-6	-23	AA	$_{ m JFK}$	LAX	329	2475	14

### Get the first two rows from flights.

R

```
ans <- flights[1:2]
kable(ans)</pre>
```

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
2014	1	1	14	13	AA	JFK	LAX	359	2475	9
2014	1	1	-3	13	AA	$_{ m JFK}$	LAX	363	2475	11

### Clojure

```
(api/select-rows flights (range 2))
```

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [2 11]:

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
2014	1	1	14	13	AA	JFK	LAX	359	2475	9
2014	1	1	-3	13	AA	$_{ m JFK}$	LAX	363	2475	11

Sort flights first by column origin in ascending order, and then by dest in descending order

```
\mathbf{R}
```

```
ans <- flights[order(origin, -dest)]
kable(head(ans))</pre>
```

year	month	day	${\rm dep\_delay}$	$arr\_delay$	carrier	origin	dest	$air\_time$	distance	hour
2014	1	5	6	49	EV	EWR	XNA	195	1131	8
2014	1	6	7	13	$\mathrm{EV}$	EWR	XNA	190	1131	8
2014	1	7	-6	-13	$\mathrm{EV}$	EWR	XNA	179	1131	8
2014	1	8	-7	-12	$\mathrm{EV}$	EWR	XNA	184	1131	8
2014	1	9	16	7	$\mathrm{EV}$	EWR	XNA	181	1131	8
2014	1	13	66	66	$\mathrm{EV}$	EWR	XNA	188	1131	9

```
(-> flights
   (api/order-by ["origin" "dest"] [:asc :desc])
   (api/head 6))
```

 $https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv\ [6\ 11]:$ 

year	month	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
${2014}$	6	3	-6	-38	EV	EWR	XNA	154	1131	6
2014	1	20	-9	-17	$\mathrm{EV}$	EWR	XNA	177	1131	8
2014	3	19	-6	10	$\mathrm{EV}$	EWR	XNA	201	1131	6
2014	2	3	231	268	$\mathrm{EV}$	EWR	XNA	184	1131	12
2014	4	25	-8	-32	$\mathrm{EV}$	EWR	XNA	159	1131	6
2014	2	19	21	10	$\mathrm{EV}$	EWR	XNA	176	1131	8

## Select arr\_delay column, but return it as a vector

 $\mathbf{R}$ 

```
ans <- flights[, arr_delay]
head(ans)</pre>
```

```
[1] 13 13 9 -26 1 0
```

Clojure

```
(take 6 (flights "arr_delay"))
```

```
(13 13 9 -26 1 0)
```

## Select arr\_delay column, but return as a data.table instead

 $\mathbf{R}$ 

```
ans <- flights[, list(arr_delay)]
kable(head(ans))</pre>
```

```
(-> flights
    (api/select-columns "arr_delay")
    (api/head 6))
```

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [6 1]:

arr_	_delay
13	
13	
9	
-26	
1	
0	

## Select both arr\_delay and dep\_delay columns

R

```
ans <- flights[, .(arr_delay, dep_delay)]
kable(head(ans))</pre>
```

arr_delay	dep_delay
13	14
13	-3
9	2
-26	-8
1	2
0	4

Clojure

```
(-> flights
    (api/select-columns ["arr_delay" "dep_delay"])
    (api/head 6))
```

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [6 2]:

$$\frac{\text{arr\_delay} \quad \text{dep\_delay}}{13} \quad 14$$

arr_delay	dep_delay
13	-3
9	2
-26	-8
1	2
0	4

Select both arr\_delay and dep\_delay columns and rename them to delay\_arr and delay\_dep  $\ensuremath{\mathrm{R}}$ 

```
ans <- flights[, .(delay_arr = arr_delay, delay_dep = dep_delay)]
kable(head(ans))</pre>
```

delay_arr	delay_dep
13	14
13	-3
9	2
-26	-8
1	2
0	4

Clojure

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [6 2]:

delay_arr	delay_arr
13	14
13	-3
9	2
-26	-8
1	2
0	4

How many trips have had total delay < 0?

```
\mathbf{R}
```

```
ans <- flights[, sum( (arr_delay + dep_delay) < 0 )]
ans</pre>
```

[1] 141814

Clojure

```
(->> (dfn/+ (flights "arr_delay") (flights "dep_delay"))
        (aops/argfilter #(< % 0.0))
        (dtype/ecount))

141814
or pure Clojure functions (much, much slower)
(->> (map + (flights "arr_delay") (flights "dep_delay"))
        (filter neg?)
        (count))
```

141814

Calculate the average arrival and departure delay for all flights with "JFK" as the origin airport in the month of June

 $\mathbf{R}$ 

m_arr	m_dep
5.839349	9.807884

Clojure

 $\underline{\phantom{a}}$ unnamed [1 2]:

:m_arr	$:\!\!\mathrm{m\_dep}$
5.83934932	9.80788411

How many trips have been made in 2014 from "JFK" airport in the month of June?

```
R
```

```
ans <- flights[origin == "JFK" & month == 6L, length(dest)]
ans

[1] 8422
or
ans <- flights[origin == "JFK" & month == 6L, .N]
ans</pre>
```

[1] 8422

8422

### deselect columns using - or !

 $\mathbf{R}$ 

```
ans <- flights[, !c("arr_delay", "dep_delay")]
kable(head(ans))</pre>
```

year	month	day	carrier	origin	dest	air_time	distance	hour
2014	1	1	AA	JFK	LAX	359	2475	9
2014	1	1	AA	$_{ m JFK}$	LAX	363	2475	11
2014	1	1	AA	$_{ m JFK}$	LAX	351	2475	19
2014	1	1	AA	LGA	PBI	157	1035	7
2014	1	1	AA	$_{ m JFK}$	LAX	350	2475	13
2014	1	1	AA	EWR	LAX	339	2454	18

or

```
ans <- flights[, -c("arr_delay", "dep_delay")]
kable(head(ans))</pre>
```

year	month	day	carrier	origin	dest	$air\_time$	distance	hour
2014	1	1	AA	JFK	LAX	359	2475	9
2014	1	1	AA	$_{ m JFK}$	LAX	363	2475	11
2014	1	1	AA	$_{ m JFK}$	LAX	351	2475	19
2014	1	1	AA	LGA	PBI	157	1035	7
2014	1	1	AA	$_{ m JFK}$	LAX	350	2475	13
2014	1	1	AA	EWR	LAX	339	2454	18

Clojure

```
(-> flights
   (api/select-columns (complement #{"arr_delay" "dep_delay"}))
   (api/head 6))
```

https://raw.githubusercontent.com/Rdatatable/data.table/master/vignettes/flights14.csv [6 9]:

year	month	day	carrier	origin	$\operatorname{dest}$	$air\_time$	distance	hour
2014	1	1	AA	JFK	LAX	359	2475	9
2014	1	1	AA	$_{ m JFK}$	LAX	363	2475	11
2014	1	1	AA	JFK	LAX	351	2475	19
2014	1	1	AA	LGA	PBI	157	1035	7
2014	1	1	AA	JFK	LAX	350	2475	13
2014	1	1	AA	EWR	LAX	339	2454	18

### Aggregations

How can we get the number of trips corresponding to each origin airport?

R

```
ans <- flights[, .(.N), by = .(origin)]
kable(ans)</pre>
```

origin	N
JFK	81483
LGA	84433
EWR	87400

Clojure

```
(-> flights
   (api/group-by ["origin"])
   (api/aggregate {:N api/row-count}))
```

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

origin	:N
LGA	84433
EWR	87400
JFK	81483

How can we calculate the number of trips for each origin airport for carrier code "AA"?

 $\mathbf{R}$ 

```
ans <- flights[carrier == "AA", .N, by = origin]
kable(ans)</pre>
```

origin	N
JFK	11923
LGA	11730
EWR	2649

Clojure

```
(-> flights
   (api/select-rows #(= (get % "carrier") "AA"))
   (api/group-by ["origin"])
   (api/aggregate {:N api/row-count}))
```

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

origin	:N
LGA	11730
EWR	2649
JFK	11923

How can we get the total number of trips for each origin, dest pair for carrier code "AA"?

 $\mathbf{R}$ 

```
ans <- flights[carrier == "AA", .N, by = .(origin, dest)]
kable(head(ans))</pre>
```

origin	dest	N
JFK	LAX	3387
LGA	PBI	245
EWR	LAX	62
JFK	MIA	1876
JFK	SEA	298
EWR	MIA	848

## Clojure

```
(-> flights
   (api/select-rows #(= (get % "carrier") "AA"))
   (api/group-by ["origin" "dest"])
   (api/aggregate {:N api/row-count})
   (api/head 6))
```

\_unnamed [6 3]:

origin	dest	:N
JFK	MIA	1876
LGA	PBI	245
$_{ m JFK}$	SEA	298
LGA	DFW	3785
$_{ m JFK}$	AUS	297
JFK	STT	229

How can we get the average arrival and departure delay for each orig,dest pair for each month for carrier code "AA"?

 ${\bf R}$ 

origin	$\operatorname{dest}$	month	V1	V2
JFK	LAX	1	6.590361	14.2289157
LGA	PBI	1	-7.758621	0.3103448

origin	dest	month	V1	V2
EWR	LAX	1	1.366667	7.5000000
JFK	MIA	1	15.720670	18.7430168
JFK	SEA	1	14.357143	30.7500000
EWR	MIA	1	11.011236	12.1235955
JFK	SFO	1	19.252252	28.6396396
JFK	BOS	1	12.919643	15.2142857
JFK	ORD	1	31.586207	40.1724138
JFK	IAH	1	28.857143	14.2857143

## \_unnamed [10 5]:

41-				
month	origin	dest	$: \!\! \text{summary-} 0$	$: \! \! \text{summary-1} \\$
9	LGA	DFW	-8.78772379	-0.25575448
10	LGA	DFW	3.50000000	4.55276382
1	JFK	AUS	25.20000000	27.60000000
4	JFK	AUS	4.36666667	-0.13333333
5	JFK	AUS	6.76666667	14.733333333
2	JFK	AUS	26.26923077	21.50000000
3	JFK	AUS	8.19354839	2.70967742
8	JFK	AUS	20.41935484	20.77419355
1	EWR	LAX	1.36666667	7.50000000
9	JFK	AUS	16.26666667	14.36666667

## So how can we directly order by all the grouping variables?

 $\mathbf{R}$ 

origin	dest	month	V1	V2
EWR	DFW	1	6.427673	10.012579
EWR	DFW	2	10.536765	11.345588
EWR	DFW	3	12.865031	8.079755
EWR	DFW	4	17.792683	12.920732
EWR	DFW	5	18.487805	18.682927
EWR	DFW	6	37.005952	38.744048
EWR	DFW	7	20.250000	21.154762
EWR	DFW	8	16.936046	22.069767

origin	dest	month	V1	V2
EWR	DFW	9	5.865031	13.055215
EWR	DFW	10	18.813665	18.894410

## Clojure

\_unnamed [10 5]:

month	origin	dest	:summary-0	:summary-1
1	EWR	DFW	6.42767296	10.01257862
2	EWR	DFW	10.53676471	11.34558824
3	EWR	DFW	12.86503067	8.07975460
4	EWR	DFW	17.79268293	12.92073171
5	EWR	DFW	18.48780488	18.68292683
6	EWR	DFW	37.00595238	38.74404762
7	EWR	DFW	20.25000000	21.15476190
8	EWR	DFW	16.93604651	22.06976744
9	EWR	DFW	5.86503067	13.05521472
10	EWR	DFW	18.81366460	18.89440994

Can by accept expressions as well or does it just take columns?

 $\mathbf{R}$ 

```
ans <- flights[, .N, .(dep_delay>0, arr_delay>0)]
kable(ans)
```

dep_delay	$\operatorname{arr\_delay}$	N
TRUE	TRUE	72836
FALSE	TRUE	34583
FALSE	FALSE	119304
TRUE	FALSE	26593

## Clojure

\_unnamed [4 3]:

:dep_delay	:arr_delay	:N
true	false	26593
false	true	34583
false	false	119304
true	true	72836

# Do we have to compute mean() for each column individually?

R

kable(DT)

ID	a	b	c
b	1	7	13
b	2	8	14
b	3	9	15
a	4	10	16
a	5	11	17
$\mathbf{c}$	6	12	18

# DT[, print(.SD), by = ID]

```
a b c

1: 1 7 13

2: 2 8 14

3: 3 9 15

a b c

1: 4 10 16

2: 5 11 17

a b c
```

1: 6 12 18

Empty data.table (0 rows and 1 cols): ID

kable(DT[, lapply(.SD, mean), by = ID])

ID	a	b	c
b	2.0	8.0	14.0
a	4.5	10.5	16.5
c	6.0	12.0	18.0

Clojure

DT

```
(api/group-by DT :ID {:result-type :as-map})
```

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

:ID	:a	:b	:с
b	1	7	13
b	2	8	14
b	3	9	15
a	4	10	16
a	5	11	17
c	6	12	18

 ${\text{``a''} Group: a [2 4]:}$ 

:ID	:a	:b	:c
a	4	10	16
a	5	11	17

, "b" Group: b [3 4]:

:ID	:a	:b	:c
b	1	7	13
b	2	8	14
b	3	9	15

, "c" Group: c [1 4]:

:ID	:a	:b	:c
c	6	12	18

```
(-> DT
     (api/group-by [:ID])
     (api/aggregate-columns (complement #{:ID}) dfn/mean))
```

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

:ID	:a	:b	:c
a	4.5	10.5	16.5
b	2.0	8.0	14.0
$\mathbf{c}$	6.0	12.0	18.0

How can we specify just the columns we would like to compute the mean() on?

R

origin	$\operatorname{dest}$	month	$\operatorname{arr\_delay}$	dep_delay
JFK	LAX	1	6.590361	14.2289157
LGA	PBI	1	-7.758621	0.3103448
EWR	LAX	1	1.366667	7.5000000
$_{ m JFK}$	MIA	1	15.720670	18.7430168
$_{ m JFK}$	SEA	1	14.357143	30.7500000
EWR	MIA	1	11.011236	12.1235955

## Clojure

```
(-> flights
   (api/select-rows #(= (get % "carrier") "AA"))
   (api/group-by ["origin" "dest" "month"])
   (api/aggregate-columns ["arr_delay" "dep_delay"] dfn/mean)
   (api/head 6))
```

#### \_unnamed [6 5]:

month	origin	dest	arr_delay	dep_delay
9	LGA	DFW	-8.78772379	-0.25575448
10	LGA	DFW	3.50000000	4.55276382
1	JFK	AUS	25.20000000	27.60000000
4	JFK	AUS	4.36666667	-0.13333333
5	JFK	AUS	6.76666667	14.733333333
2	JFK	AUS	26.26923077	21.50000000

## How can we return the first two rows for each month?

R

```
ans <- flights[, head(.SD, 2), by = month]
kable(head(ans))</pre>
```

month	year	day	dep_delay	arr_delay	carrier	origin	dest	air_time	distance	hour
1	2014	1	14	13	AA	JFK	LAX	359	2475	9
1	2014	1	-3	13	AA	$_{ m JFK}$	LAX	363	2475	11
2	2014	1	-1	1	AA	$_{ m JFK}$	LAX	358	2475	8
2	2014	1	-5	3	AA	$_{ m JFK}$	LAX	358	2475	11
3	2014	1	-11	36	AA	$_{ m JFK}$	LAX	375	2475	8
3	2014	1	-3	14	AA	JFK	LAX	368	2475	11

# Clojure

```
(-> flights
   (api/group-by ["month"])
   (api/head 2) ;; head applied on each group
   (api/ungroup)
   (api/head 6))
```

\_unnamed [6 11]:

dep_delay	origin	air_time	hour	arr_delay	dest	distance	year	month	day	carrier
-8	LGA	113	18	-23	BNA	764	2014	4	1	$\overline{MQ}$
-8	LGA	71	18	-11	RDU	431	2014	4	1	MQ
43	JFK	288	17	5	LAS	2248	2014	5	1	AA
-1	JFK	330	7	-38	SFO	2586	2014	5	1	AA
-9	JFK	324	8	-5	LAX	2475	2014	6	1	AA
-10	$_{ m JFK}$	329	12	-13	LAX	2475	2014	6	1	AA

How can we concatenate columns a and b for each group in ID?

R

```
kable(DT[, .(val = c(a,b)), by = ID])
```

ID	val
b	1
b	2
b	3
b	7
b	8
b	9
a	4
a	5
a	10
a	11
$\mathbf{c}$	6
$\mathbf{c}$	12

Clojure

```
(-> DT
    (api/pivot->longer [:a :b] {:value-column-name :val})
    (api/drop-columns [:$column :c]))
```

 $\underline{\phantom{a}}$ unnamed [12 2]:

:ID	:va
b	1
b	2
b	3
a	4
a	5
$\mathbf{c}$	6
b	7
b	8
b	9
$\mathbf{a}$	10
$\mathbf{a}$	11
$\mathbf{c}$	12

What if we would like to have all the values of column a and b concatenated, but returned as a list column?

 $\mathbf{R}$ 

```
kable(DT[, .(val = list(c(a,b))), by = ID])
                                          ID
                                               val
                                          b
                                               1, 2, 3, 7, 8, 9
                                               4, 5, 10, 11
                                          a
                                               6, 12
```

Clojure

```
(-> DT
    (api/pivot->longer [:a :b] {:value-column-name :val})
    (api/drop-columns [:$column :c])
    (api/fold-by :ID))
```

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

:ID	:val
a b	[4 5 10 11] [1 2 3 7 8 9]
$\mathbf{c}$	[6 12]

#### **API** tour

Below snippets are taken from A data.table and dplyr tour written by Atrebas (permission granted).

I keep structure and subtitles but I skip data.table and dplyr examples.

Example data

```
(def DS (api/dataset {:V1 (take 9 (cycle [1 2]))
                      :V2 (range 1 10)
                      :V3 (take 9 (cycle [0.5 1.0 1.5]))
                      :V4 (take 9 (cycle ["A" "B" "C"]))}))
(api/dataset? DS)
(class DS)
true
tech.v3.dataset.impl.dataset.Dataset
```

DS

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В

:V1	:V2	:V3	:V4
2	6	1.5	С
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

# **Basic Operations**

#### Filter rows

Filter rows using indices

(api/select-rows DS [2 3])

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

:V1	:V2	:V3	:V4
1	3	1.5	С
2	4	0.5	A

Discard rows using negative indices

In Clojure API we have separate function for that: drop-rows.

(api/drop-rows DS (range 2 7))

\_unnamed [4 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

Filter rows using a logical expression

(api/select-rows DS (comp #(> % 5) :V2))

\_unnamed [4 4]:

:V1	:V2	:V3	:V4
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

(api/select-rows DS (comp #{"A" "C"} :V4))

\_unnamed [6 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
1	9	1.5	$\mathbf{C}$

Filter rows using multiple conditions

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

:V1	:V2	:V3	:V4
1	1	0.5	A
1	7	0.5	A

Filter unique rows

```
(api/unique-by DS)
```

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$^{\mathrm{C}}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

```
(api/unique-by DS [:V1 :V4])
```

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

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$

# Discard rows with missing values

# (api/drop-missing DS)

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	С

# Other filters

(api/random DS 3) ;; 3 random rows

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

:V1	:V2	:V3	:V4
2	4	0.5	A
1	1	0.5	A
2	8	1.0	В

(api/random DS (/ (api/row-count DS) 2)) ;; fraction of random rows

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

:V1	:V2	:V3	:V4
1	9	1.5	С
2	2	1.0	В
1	9	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$
2	8	1.0	В

(api/by-rank DS :V1 zero?) ;; take top n entries

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

:V1	:V2	:V3	:V4
2	2	1.0	В
2	4	0.5	A
2	6	1.5	$\mathbf{C}$
2	8	1.0	В

Convenience functions

(api/select-rows DS (comp (partial re-matches #"^B") str :V4))

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

:V1	:V2	:V3	:V4
2	2	1.0	В
1	5	1.0	В
2	8	1.0	В

(api/select-rows DS (comp #(<= 3 % 5) :V2))

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

:V1	:V2	:V3	:V4
1	3	1.5	С
2	4	0.5	A
1	5	1.0	В

(api/select-rows DS (comp #(<3%5):V2))

\_unnamed [1 4]:

:V1	:V2	:V3	:V4
2	4	0.5	A

(api/select-rows DS (comp #(<= 3 % 5) :V2))

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

:V1	:V2	:V3	:V4
1	3	1.5	С
2	4	0.5	A
1	5	1.0	В

Last example skipped.

Sort rows

Sort rows by column

(api/order-by DS :V3)

\_unnamed [9 4]:

:V1	:V2	:V3	:V4
1	1	0.5	A
2	4	0.5	A
1	7	0.5	A
2	2	1.0	В
1	5	1.0	В
2	8	1.0	В
1	3	1.5	$\mathbf{C}$
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$

Sort rows in decreasing order

```
(api/order-by DS : V3 :desc)
```

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

:V1	:V2	:V3	:V4
1	3	1.5	С
2	6	1.5	$\mathbf{C}$
1	9	1.5	$\mathbf{C}$
1	5	1.0	В
2	2	1.0	В
2	8	1.0	В
1	7	0.5	A
2	4	0.5	A
1	1	0.5	A

Sort rows based on several columns

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

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

:V1	:V2	:V3	:V4
1	9	1.5	С
1	7	0.5	A
1	5	1.0	В
1	3	1.5	$\mathbf{C}$
1	1	0.5	A
2	8	1.0	В
2	6	1.5	$\mathbf{C}$
2	4	0.5	A
2	2	1.0	В

# Select columns

Select one column using an index (not recommended)

```
(nth (api/columns DS :as-seq) 2) ;; as column (iterable)
#tech.v3.dataset.column<float64>[9]
: ٧3
[0.5000, 1.000, 1.500, 0.5000, 1.000, 1.500, 0.5000, 1.000, 1.500]
(api/dataset [(nth (api/columns DS :as-seq) 2)])
_unnamed [9 1]:
                                              :V3
                                             0.5
                                             1.0
                                             1.5
                                             0.5
                                             1.0
                                             1.5
                                             0.5
                                             1.0
                                             1.5
Select one column using column name
(api/select-columns DS :V2) ;; as dataset
_unnamed [9 1]:
                                             :V2
                                             2
                                             3
                                             4
                                             5
                                             6
                                             7
                                             8
                                             9
(api/select-columns DS [:V2]) ;; as dataset
_unnamed [9 1]:
                                              :V2
                                             1
                                             2
                                             3
                                             4
                                             5
                                             6
                                             7
```

```
<u>:V2</u>
9
```

```
(DS:V2);; as column (iterable)
```

#tech.v3.dataset.column<int64>[9]

:V2 [1, 2, 3, 4, 5, 6, 7, 8, 9]

\_\_\_\_\_

Select several columns

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

\_unnamed [9 3]:

:V2 :V3	:V4
1 0.5	A
2   1.0	В
$3 \qquad 1.5$	$\mathbf{C}$
4   0.5	A
5 $1.0$	В
6   1.5	$\mathbf{C}$
7   0.5	A
8 1.0	В
9 1.5	С

Exclude columns

```
(api/select-columns DS (complement #{:V2 :V3 :V4}))
```

\_unnamed [9 1]:

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

\_unnamed [9 1]:

Other seletions

```
(->> (range 1 3)
      (map (comp keyword (partial format "V%d")))
      (api/select-columns DS))
```

\_unnamed [9 2]:

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

(api/reorder-columns DS : V4)

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

:V4	:V1	:V2	:V3
A	1	1	0.5
В	2	2	1.0
$\mathbf{C}$	1	3	1.5
A	2	4	0.5
В	1	5	1.0
$\mathbf{C}$	2	6	1.5
A	1	7	0.5
В	2	8	1.0
C	1	9	1.5

```
(api/select-columns DS #(clojure.string/starts-with? (name %) "V"))
```

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

:V1	:V2	:V3	:V4
1	1	0.5	A
2	2	1.0	В
1	3	1.5	$\mathbf{C}$
2	4	0.5	A
1	5	1.0	В
2	6	1.5	$\mathbf{C}$
1	7	0.5	A
2	8	1.0	В
1	9	1.5	$\mathbf{C}$

```
(api/select-columns DS #(clojure.string/ends-with? (name %) "3"))
_unnamed [9 1]:
                                                 :V3
                                                 0.5
                                                 1.0
                                                 1.5
                                                 0.5
                                                 1.0
                                                 1.5
                                                 0.5
                                                 1.0
                                                 1.5
(api/select-columns DS #"..2") ;; regex converts to string using `str` function
_unnamed [9 1]:
                                                 :V2
                                                 1
                                                 2
3
4
                                                 5
6
                                                 7
                                                 8
(api/select-columns DS #{:V1 "X"})
_unnamed [9 1]:
                                                \frac{\overline{:V1}}{1}
                                                 2
```

1 2

```
\frac{\overline{V1}}{1}

\frac{1}{2}

\frac{1}{2}

\frac{1}{2}

\frac{1}{2}
```

```
(api/select-columns DS #(not (clojure.string/starts-with? (name %) "V2")))
```

\_unnamed [9 3]:

:V1	:V3	:V4
1	0.5	A
2	1.0	В
1	1.5	$\mathbf{C}$
2	0.5	A
1	1.0	В
2	1.5	$\mathbf{C}$
1	0.5	A
2	1.0	В
1	1.5	$\mathbf{C}$

#### Summarise data

```
Summarise one column
```

Summarize several columns

\_unnamed [1 2]:

:summary-0	:summary-1
13.0	0.4330127

```
(api/aggregate-columns DS [:V1 :V3] [dfn/sum dfn/standard-deviation])
_unnamed [1 2]:
```

```
:V1 :V3
13.0 0.4330127
```

Summarise several columns and assign column names

 $\underline{\quad}$  unnamed [1 2]:

:sumv1	:sdv3
13.0	0.4330127

Summarise a subset of rows

```
(-> DS
     (api/select-rows (range 4))
     (api/aggregate-columns :V1 dfn/sum))
```

\_unnamed [1 1]:

 $\frac{\overline{\text{V1}}}{6.0}$ 

#### Additional helpers

```
(-> DS
     (api/first)
     (api/select-columns :V3)) ;; select first row from `:V3` column
```

\_unnamed [1 1]:

 $\frac{:\text{V3}}{0.5}$ 

```
(-> DS
    (api/last)
    (api/select-columns :V3)) ;; select last row from `:V3` column
```

```
_unnamed [1 1]:
                                              :V3
                                              1.5
(-> DS
    (api/select-rows 4)
    (api/select-columns :V3)) ;; select forth row from `:V3` column
_unnamed [1 1]:
(-> DS
    (api/select :V3 4)) ;; select forth row from `:V3` column
_unnamed [1 1]:
                                              :V3
                                              1.0
(-> DS
    (api/unique-by: V4)
    (api/aggregate api/row-count)) ;; number of unique rows in `:V4` column, as dataset
_unnamed [1 1]:
                                            :summary
(-> DS
    (api/unique-by: V4)
    (api/row-count)) ;; number of unique rows in `:V4` column, as value
3
(-> DS
    (api/unique-by)
    (api/row-count)) ;; number of unique rows in dataset, as value
9
Add/update/delete columns
Modify a column
(api/map-columns DS :V1 [:V1] #(dfn/pow % 2))
\underline{\hspace{0.1cm}}unnamed [9 4]:
```

:V1	:V2	:V3	:V4
1	1	0.5	A
4	2	1.0	В
1	3	1.5	$\mathbf{C}$
4	4	0.5	A
1	5	1.0	В
4	6	1.5	$\mathbf{C}$
1	7	0.5	A
4	8	1.0	В
1	9	1.5	$\mathbf{C}$

```
(def DS (api/add-column DS :V1 (dfn/pow (DS :V1) 2)))
DS
```

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

:V1	:V2	:V3	:V4
1.0	1	0.5	A
4.0	2	1.0	В
1.0	3	1.5	$\mathbf{C}$
4.0	4	0.5	A
1.0	5	1.0	В
4.0	6	1.5	$\mathbf{C}$
1.0	7	0.5	A
4.0	8	1.0	В
1.0	9	1.5	$\mathbf{C}$

## Add one column

```
(api/map-columns DS :v5 [:V1] dfn/log)
```

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

:V1	:V2	:V3	:V4	:v5
1.0	1	0.5	A	0.00000000
4.0	2	1.0	В	1.38629436
1.0	3	1.5	$\mathbf{C}$	0.00000000
4.0	4	0.5	A	1.38629436
1.0	5	1.0	В	0.00000000
4.0	6	1.5	$\mathbf{C}$	1.38629436
1.0	7	0.5	A	0.00000000
4.0	8	1.0	В	1.38629436
1.0	9	1.5	С	0.00000000

```
(def DS (api/add-column DS :v5 (dfn/log (DS :V1))))
DS
```

\_unnamed [9 5]:

:V1	:V2	:V3	:V4	:v5
1.0	1	0.5	A	0.00000000
4.0	2	1.0	В	1.38629436
1.0	3	1.5	$\mathbf{C}$	0.00000000
4.0	4	0.5	A	1.38629436
1.0	5	1.0	В	0.00000000
4.0	6	1.5	$\mathbf{C}$	1.38629436
1.0	7	0.5	A	0.00000000
4.0	8	1.0	В	1.38629436
1.0	9	1.5	$\mathbf{C}$	0.00000000

# Add several columns

# $\underline{\phantom{a}}$ unnamed [9 7]:

:V1	:V2	:V3	:V4	:v5	:v6	:v7
1.0	1	0.5	A	0.00000000	1.0	X
4.0	2	1.0	В	1.38629436	2.0	X
1.0	3	1.5	$\mathbf{C}$	0.00000000	1.0	X
4.0	4	0.5	A	1.38629436	2.0	X
1.0	5	1.0	В	0.00000000	1.0	X
4.0	6	1.5	$\mathbf{C}$	1.38629436	2.0	X
1.0	7	0.5	A	0.00000000	1.0	X
4.0	8	1.0	В	1.38629436	2.0	X
1.0	9	1.5	$\mathbf{C}$	0.00000000	1.0	X

# Create one column and remove the others

```
(api/dataset {:v8 (dfn/+ (DS :V3) 1)})
```

# $\underline{\phantom{a}}$ unnamed [9 1]:

:v8 1.5 2.0 2.5 1.5 2.0 2.5 1.5 2.0 2.5 2.5 Remove one column

```
(def DS (api/drop-columns DS :v5))
```

DS

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

:V1	:V2	:V3	:V4	:v6	:v7
1.0	1	0.5	A	1.0	X
4.0	2	1.0	В	2.0	X
1.0	3	1.5	$\mathbf{C}$	1.0	X
4.0	4	0.5	A	2.0	X
1.0	5	1.0	В	1.0	X
4.0	6	1.5	$\mathbf{C}$	2.0	X
1.0	7	0.5	A	1.0	X
4.0	8	1.0	В	2.0	X
1.0	9	1.5	$\mathbf{C}$	1.0	X

Remove several columns

```
(def DS (api/drop-columns DS [:v6 :v7]))
```

DS

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

:V1	:V2	:V3	:V4
1.0	1	0.5	A
4.0	2	1.0	В
1.0	3	1.5	$\mathbf{C}$
4.0	4	0.5	A
1.0	5	1.0	В
4.0	6	1.5	$\mathbf{C}$
1.0	7	0.5	A
4.0	8	1.0	В
1.0	9	1.5	$\mathbf{C}$

Remove columns using a vector of colnames

We use set here.

```
(def DS (api/select-columns DS (complement #{:V3})))
```

DS

 $\underline{\phantom{a}}$ unnamed [9 3]:

:V1	:V2	:V4
1.0	1	A
4.0	2	В
1.0	3	$\mathbf{C}$
4.0	4	A

:V1	:V2	:V4
1.0	5	В
4.0	6	$\mathbf{C}$
1.0	7	A
4.0	8	В
1.0	9	$\mathbf{C}$

Replace values for rows matching a condition

```
(def DS (api/map-columns DS :V2 [:V2] #(if (< % 4.0) 0.0 %)))
DS
```

\_unnamed [9 3]:

:V1	:V2	:V4
1.0	0	A
4.0	0	В
1.0	0	$\mathbf{C}$
4.0	4	A
1.0	5	В
4.0	6	$\mathbf{C}$
1.0	7	A
4.0	8	В
1.0	9	С

 $\mathbf{b}\mathbf{y}$ 

By group

```
(-> DS
    (api/group-by [:V4])
    (api/aggregate {:sumV2 #(dfn/sum (% :V2))}))
```

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

:V4	:sumV2
В	13.0
$\mathbf{C}$	15.0
A	11.0

By several groups

 $\underline{\phantom{a}}$ unnamed [6 3]:

:V4	:V1	:sumV2
A	4.0	4.0
A	1.0	7.0
В	1.0	5.0
В	4.0	8.0
$\mathbf{C}$	4.0	6.0
$\mathbf{C}$	1.0	9.0

Calling function in by

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

:\$group-name	:sumV1
a	6.0
b	9.0
$\mathbf{c}$	6.0

Assigning column name in by

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

```
a 6.0
b 9.0
c 6.0
```

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

:abc	:sumV1
a	6.0
b	9.0
$\mathbf{c}$	6.0

Using a condition in by

```
(-> DS
          (api/group-by #(= (:V4 %) "A"))
          (api/aggregate #(dfn/sum (% :V1))))
```

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

:\$group-name	:summary
false	15.0
true	6.0

By on a subset of rows

```
(-> DS
    (api/select-rows (range 5))
    (api/group-by : V4)
    (api/aggregate {:sumV1 #(dfn/sum (% :V1))}))
```

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

:\$group-name	:sumV1
A	5.0
В	5.0
$\mathbf{C}$	1.0

Count number of observations for each group

```
(-> DS
    (api/group-by :V4)
    (api/aggregate api/row-count))
```

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

:\$group-name	:summary
A	3
В	3
С	3

Add a column with number of observations for each group

```
(-> DS
    (api/group-by [:V1])
    (api/add-column :n api/row-count)
    (api/ungroup))
```

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

:V1	:V2	:V4	:n
4.0	0	В	4
4.0	4	A	4
4.0	6	$\mathbf{C}$	4
4.0	8	В	4
1.0	0	$\mathbf{A}$	5
1.0	0	$\mathbf{C}$	5
1.0	5	В	5
1.0	7	A	5
1.0	9	$\mathbf{C}$	5

Retrieve the first/last/nth observation for each group

```
(-> DS
    (api/group-by [:V4])
    (api/aggregate-columns :V2 first))
```

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

:V2
0
0
0

```
(-> DS
     (api/group-by [:V4])
     (api/aggregate-columns :V2 last))
```

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

8
9
7

```
(-> DS
    (api/group-by [:V4])
    (api/aggregate-columns :V2 #(nth % 1)))
```

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

:V4	:V2
В	5
$\mathbf{C}$	6
A	4

Going further

#### Advanced columns manipulation

Summarise all the columns

```
;; custom max function which works on every type
(api/aggregate-columns DS :all (fn [col] (first (sort #(compare %2 %1) col))))
```

 $\underline{\quad}$  unnamed [1 3]:

```
:V1 :V2 :V4
4.0 9 C
```

Summarise several columns

```
(api/aggregate-columns DS [:V1 :V2] dfn/mean)
```

 $\underline{\quad}$  unnamed [1 2]:

:V1	:V2
2.33333333	4.33333333

Summarise several columns by group

```
(-> DS
    (api/group-by [:V4])
    (api/aggregate-columns [:V1 :V2] dfn/mean))
```

\_unnamed [3 3]:

:V4	:V1	:V2
В	3.0	4.33333333
$\mathbf{C}$	2.0	5.00000000
A	2.0	3.66666667

Summarise with more than one function by group

 $\underline{\phantom{a}}$ unnamed [3 5]:

:V4	:V1-sum	:V1-mean	:V2-sum	:V2-mean
В	9.0	3.0	13.0	4.33333333
$\mathbf{C}$	6.0	2.0	15.0	5.00000000
A	6.0	2.0	11.0	3.66666667

Summarise using a condition

```
(-> DS
    (api/select-columns :type/numerical)
    (api/aggregate-columns :all dfn/mean))
```

\_unnamed [1 2]:

```
:V1 :V2
2.33333333 4.33333333
```

Modify all the columns

```
(api/update-columns DS :all reverse)
```

\_unnamed [9 3]:

:V1	:V2	:V4
1.000	9	С
4.000	8	В
1.000	7	A
4.000	6	$\mathbf{C}$
1.000	5	В
4.000	4	A
1.000	0	$\mathbf{C}$
4.000	0	В
1.000	0	A

Modify several columns (dropping the others)

```
(-> DS
    (api/select-columns [:V1 :V2])
    (api/update-columns :all dfn/sqrt))
```

\_unnamed [9 2]:

```
:V1
     :V2
1.0
      0.00000000
2.0
      0.00000000
1.0
      0.00000000
2.0
      2.00000000
1.0
      2.23606798
2.0
      2.44948974
1.0
      2.64575131
2.0
      2.82842712
      3.00000000
1.0
```

```
(-> DS
    (api/select-columns (complement #{:V4}))
```

# (api/update-columns :all dfn/exp))

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

:V1	:V2
2.71828183	1.00000000
54.59815003	1.00000000
2.71828183	1.00000000
54.59815003	54.59815003
2.71828183	148.41315910
54.59815003	403.42879349
2.71828183	1096.63315843
54.59815003	2980.95798704
2.71828183	8103.08392758

Modify several columns (keeping the others)

```
(def DS (api/update-columns DS [:V1 :V2] dfn/sqrt))
```

DS

 $\underline{\phantom{a}}$ unnamed [9 3]:

:V1	:V2	:V4
1.0	0.00000000	A
2.0	0.00000000	В
1.0	0.00000000	$\mathbf{C}$
2.0	2.00000000	A
1.0	2.23606798	В
2.0	2.44948974	$\mathbf{C}$
1.0	2.64575131	A
2.0	2.82842712	В
1.0	3.00000000	$\mathbf{C}$

```
(def DS (api/update-columns DS (complement #{:V4}) #(dfn/pow % 2)))
DS
```

 $\underline{\phantom{a}}$ unnamed [9 3]:

:V1	:V2	:V4
1.0	0.0	A
4.0	0.0	В
1.0	0.0	$\mathbf{C}$
4.0	4.0	A
1.0	5.0	В
4.0	6.0	$\mathbf{C}$
1.0	7.0	A
4.0	8.0	В
1.0	9.0	$\mathbf{C}$

Modify columns using a condition (dropping the others)

```
(-> DS
     (api/select-columns :type/numerical)
     (api/update-columns :all #(dfn/- % 1)))
```

\_unnamed [9 2]:

:V1	:V2
0.0	-1.0
3.0	-1.0
0.0	-1.0
3.0	3.0
0.0	4.0
3.0	5.0
0.0	6.0
3.0	7.0
0.0	8.0

Modify columns using a condition (keeping the others)

```
(def DS (api/convert-types DS :type/numerical :int32))
```

DS

\_unnamed [9 3]:

:V1	:V2	:V4
1	0	A
4	0	В
1	0	$\mathbf{C}$
4	4	A
1	5	В
4	5	$\mathbf{C}$
1	7	A
4	8	В
1	9	С

Use a complex expression

```
(-> DS
    (api/group-by [:V4])
    (api/head 2)
    (api/add-column :V2 "X")
    (api/ungroup))
```

\_unnamed [6 3]:

:V1	:V2	:V4
4	X	В
1	X	В
1	X	$\mathbf{C}$
4	X	$\mathbf{C}$
1	X	A
4	X	A

Use multiple expressions

# (1 4 1 4 1 4 1 4 1) \_unnamed: descriptive-stats [1 11]:

:col-	:datatyp	:n- vevalid	:n- missing	:min	:mean	:max	:standard- deviation	:skew	:first	:last
:V1	:int32	9	0	1.0	2.333333		1.58113883	0.271055		1

## \_unnamed [9 2]:

:A	:В
1	39.0
2	42.0
3	39.0
4	42.0
5	39.0
6	42.0
7	39.0
8	42.0
9	39.0

# Chain expressions

Expression chaining using >

```
(-> DS
     (api/group-by [:V4])
     (api/aggregate {:V1sum #(dfn/sum (% :V1))})
     (api/select-rows #(>= (:V1sum %) 5)))
```

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

:V4	:V1sum
В	9.0
$\mathbf{C}$	6.0
A	6.0

```
(-> DS
    (api/group-by [:V4])
    (api/aggregate {:V1sum #(dfn/sum (% :V1))})
    (api/order-by :V1sum :desc))
```

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

:V4	:V1sum
В	9.0
$\mathbf{C}$	6.0
A	6.0

# Indexing and Keys

Set the key/index (order)

```
(def DS (api/order-by DS :V4))
```

DS

\_unnamed [9 3]:

:V1	:V2	:V4
1	0	A
4	4	A
1	7	A
4	0	В
1	5	В
4	8	В
1	0	$\mathbf{C}$
4	5	$\mathbf{C}$
1	9	С

Select the matching rows

```
(api/select-rows DS #(= (:V4 %) "A"))
```

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

:V1	:V2	:V4
1	0	Α
4	4	A
1	7	A

```
(api/select-rows DS (comp #{"A" "C"} :V4))
```

 $\underline{\phantom{a}}$ unnamed [6 3]:

:V1	:V2	:V4
1	0	A
4	4	Α

:V1	:V2	:V4
1	7	A
1	0	$\mathbf{C}$
4	5	$\mathbf{C}$
1	9	$\mathbf{C}$

Select the first matching row

```
(-> DS
          (api/select-rows #(= (:V4 %) "B"))
          (api/first))
```

\_unnamed [1 3]:

$$\frac{\text{:V1 :V2 :V4}}{4 \quad 0 \quad \text{B}}$$

```
(-> DS
    (api/unique-by :V4)
    (api/select-rows (comp #{"B" "C"} :V4)))
```

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

:V1	:V2	:V4
4	0	В
1	0	С

Select the last matching row  $\,$ 

```
(-> DS
     (api/select-rows #(= (:V4 %) "A"))
     (api/last))
```

 $\underline{\phantom{a}}$ unnamed [1 3]:

$$\begin{array}{c|cccc}
\hline
:V1 & :V2 & :V4 \\
\hline
1 & 7 & A
\end{array}$$

 $No match \ argument$ 

```
(api/select-rows DS (comp #{"A" "D"} :V4))
```

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

```
\frac{\text{:V1 :V2 :V4}}{1 \quad 7 \quad A}
```

Apply a function on the matching rows

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

 $\frac{\text{:V1-sum}}{12.0}$ 

Modify values for matching rows

DS

\_unnamed [9 3]:

0 0 A 0 4 A 0 7 A 4 0 B			
0 4 A 0 7 A 4 0 B	:V1	:V2	:V4
0 7 A 4 0 B	0	0	A
4 0 B	0	4	A
	0	7	Α
	4	0	В
1 5 B	1	5	В
4 8 B	4	8	В
1 0 C	1	0	$\mathbf{C}$
4 5 C	4	5	$\mathbf{C}$
1 9 C	1	9	С

Use keys in by

```
(-> DS
    (api/select-rows (comp (complement #{"B"}) :V4))
    (api/group-by [:V4])
    (api/aggregate-columns :V1 dfn/sum))
```

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

```
:V4 :V1
C 6.0
A 0.0
```

Set keys/indices for multiple columns (ordered)

```
(api/order-by DS [:V4 :V1])
```

\_unnamed [9 3]:

:V1	:V2	:V4
0	0	A
0	4	A
0	7	A
1	5	В
4	0	В
4	8	В
1	0	$\mathbf{C}$
1	9	$\mathbf{C}$
4	5	$\mathbf{C}$

Subset using multiple keys/indices

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

:V1	:V2	:V4
1	0	С
1	9	$\mathbf{C}$

\_unnamed [3 3]:

:V2	:V4
5	В
0	$\mathbf{C}$
9	$\mathbf{C}$
	5 0

(4 6 8)

# set\*() modifications

Replace values

There is no mutating operations  ${\tt tech.ml.dataset}$  or easy way to set value.

DS

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

:V1	:V2	:V4
0	3	A
0	4	A
0	7	A
4	0	В
1	5	В
4	8	В
1	0	$\mathbf{C}$
4	5	$\mathbf{C}$
1	9	С

 ${\bf Reorder\ rows}$ 

```
(def DS (api/order-by DS [:V4 :V1] [:asc :desc]))
```

DS

 $\underline{\phantom{a}}$ unnamed [9 3]:

:V1	:V2	:V4
0	3	A
0	4	A
0	7	A
4	0	В
4	8	В
1	5	В
4	5	$\mathbf{C}$
1	0	$\mathbf{C}$
1	9	С

 ${\bf Modify\ colnames}$ 

```
(def DS (api/rename-columns DS {:V2 "v2"}))
```

DS

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

:V1	v2	:V4
0	3	A
0	4	A
0	7	A
4	0	В

:V1	v2	:V4
4	8	В
1	5	В
4	5	$\mathbf{C}$
1	0	$\mathbf{C}$
1	9	$\mathbf{C}$

```
(def DS (api/rename-columns DS {"v2" :V2})) ;; revert back
```

Reorder columns

```
(def DS (api/reorder-columns DS : V4 : V1 : V2))
```

DS

 $\underline{\phantom{a}}$ unnamed [9 3]:

:V4	:V1	:V2
A	0	3
A	0	4
A	0	7
В	4	0
В	4	8
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	0
$\mathbf{C}$	1	9

## Advanced use of by

Select first/last/... row by group

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

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

```
    :V4
    :V1
    :V2

    A
    0
    3

    B
    4
    0

    C
    4
    5
```

```
(-> DS
    (api/group-by :V4)
    (api/select-rows [0 2])
    (api/ungroup))
```

 $\underline{\phantom{a}}$ unnamed [6 3]:

:V4	:V1	:V2
A	0	3
A	0	7
В	4	0
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	9

```
(-> DS
    (api/group-by :V4)
    (api/tail 2)
    (api/ungroup))
```

 $\underline{\phantom{a}}$ unnamed [6 3]:

:V4	:V1	:V2
A	0	4
A	0	7
В	4	8
В	1	5
$\mathbf{C}$	1	0
$\mathbf{C}$	1	9

Select rows using a nested query

```
(-> DS
    (api/group-by : V4)
    (api/order-by : V2)
    (api/first)
    (api/ungroup))
```

\_unnamed [3 3]:

:V4	:V1	:V2
A	0	3
В	4	0
$\mathbf{C}$	1	0

Add a group counter column

```
(-> DS
    (api/group-by [:V4 :V1])
    (api/ungroup {:add-group-id-as-column :Grp}))
```

\_unnamed [9 4]:

:Grp	:V4	:V1	:V2
0	A	0	3
0	A	0	4

:Grp	:V4	:V1	:V2
0	A	0	7
1	В	1	5
2	$\mathbf{C}$	1	0
2	$\mathbf{C}$	1	9
3	В	4	0
3	В	4	8
4	С	4	5

Get row number of first (and last) observation by group

```
(-> DS
    (api/add-column :row-id (range))
    (api/select-columns [:V4 :row-id])
    (api/group-by :V4)
    (api/ungroup))
```

\_unnamed [9 2]:

```
:V4
        :row-id
        0
Α
Α
        1
        2
Α
В
        3
В
        4
В
        5
\mathbf{C}
        6
\mathbf{C}
        7
\mathbf{C}
        8
```

```
(-> DS
    (api/add-column :row-id (range))
    (api/select-columns [:V4 :row-id])
    (api/group-by :V4)
    (api/first)
    (api/ungroup))
```

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

```
:V4 :row-id

A 0

B 3

C 6
```

```
(-> DS
    (api/add-column :row-id (range))
    (api/select-columns [:V4 :row-id])
    (api/group-by :V4)
    (api/select-rows [0 2])
```

# (api/ungroup))

 $\underline{\phantom{a}}$ unnamed [6 2]:

:V4	:row-id
A	0
A	2
В	3
В	5
$\mathbf{C}$	6
$\mathbf{C}$	8

Handle list-columns by group

```
(-> DS
    (api/select-columns [:V1 :V4])
    (api/fold-by :V4))
```

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

:V4	:V1
В	$[4 \ 4 \ 1]$
$\mathbf{C}$	$[4 \ 1 \ 1]$
A	$[0 \ 0 \ 0]$

```
(-> DS
    (api/group-by :V4)
    (api/unmark-group))
```

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

:name	:group-id	:data
A	0	Group: A [3 3]:
В	1	Group: B [3 3]:
С	2	Group: C [3 3]:

Grouping sets (multiple by at once)

Not available.

# Miscellaneous

## Read / Write data

Write data to a csv file

```
(api/write! DS "DF.csv")
```

nil

Write data to a tab-delimited file

```
(api/write! DS "DF.txt" {:separator \tab})
```

nil

or

```
(api/write! DS "DF.tsv")
```

nil

Read a csv / tab-delimited file

```
(api/dataset "DF.csv" {:key-fn keyword})
```

DF.csv [9 3]:

:V4	:V1	:V2
A	0	3
A	0	4
A	0	7
В	4	0
В	4	8
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	0
С	1	9

```
(api/dataset "DF.txt" {:key-fn keyword})
```

DF.txt [9 3]:

:V4	:V1	:V2
A	0	3
A	0	4
A	0	7
В	4	0
В	4	8
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	0
$\mathbf{C}$	1	9

```
(api/dataset "DF.tsv" {:key-fn keyword})
```

DF.tsv [9 3]:

:V4	:V1	:V2
A	0	3
A	0	4

:V4	:V1	:V2
A	0	7
В	4	0
В	4	8
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	0
$\mathbf{C}$	1	9

Read a csv file selecting / droping columns

DF.csv [9 2]:

:V1	:V4
0	A
0	A
0	A
4	В
4	В
1	В
4	$\mathbf{C}$
1	$\mathbf{C}$
1	$\mathbf{C}$

DF.csv [9 2]:

:V1	:V2
0	3
0	4
0	7
4	0
4	8
1	5
4	5
1	0
1	9

Read and rbind several files

```
(apply api/concat (map api/dataset ["DF.csv" "DF.csv"]))
```

null [18 3]:

V4	V1	$V_2$
A	0	3
A	0	4
A	0	7
В	4	0
В	4	8
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	0
$\mathbf{C}$	1	9
A	0	3
A	0	4
A	0	7
В	4	0
В	4	8
В	1	5
$\mathbf{C}$	4	5
$\mathbf{C}$	1	0
<u>C</u>	1	9

# Reshape data

```
Melt data (from wide to long)
```

\_unnamed [18 3]:

:V4	:variable	:value
A	:V1	0
A	:V1	0
A	:V1	0
В	:V1	4
В	:V1	4
В	:V1	1
$\mathbf{C}$	:V1	4
$\mathbf{C}$	:V1	1
$\mathbf{C}$	:V1	1
A	:V2	3
A	:V2	4
A	:V2	7
В	:V2	0
В	:V2	8
В	:V2	5
$\mathbf{C}$	:V2	5
$\mathbf{C}$	:V2	0
$\mathbf{C}$	:V2	9

Cast data (from long to wide)

```
(-> mDS
   (api/pivot->wider :variable :value {:fold-fn vec})
   (api/update-columns ["V1" "V2"] (partial map count)))
```

\_unnamed [3 3]:

:V4	V1	V2
В	3	3
A	3	3
$\mathbf{C}$	3	3

```
(-> mDS
    (api/pivot->wider :variable :value {:fold-fn vec})
    (api/update-columns ["V1" "V2"] (partial map dfn/sum)))
```

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

:V4	V1	V2
В	9.000	13.00
A	0.000	14.00
$\mathbf{C}$	6.000	14.00

```
(-> mDS
    (api/map-columns :value #(> % 5))
    (api/pivot->wider :value :variable {:fold-fn vec})
    (api/update-columns ["true" "false"] (partial map #(if (sequential? %) (count %) 1))))
```

\_unnamed [3 3]:

:V4	true	false
$\overline{\mathrm{C}}$	1	5
A	1	5
В	1	5

Split

```
(api/group-by DS :V4 {:result-type :as-map})
```

{"A" Group: A [3 3]:

:V4	:V1	:V2
A	0	3
A	0	4
A	0	7

, "B" Group: B [3 3]:

:V1	:V2
4	0
4	8
1	5
	4 4

, "C" Group: C [3 3]:

:V4	:V1	:V2
$\overline{\mathbf{C}}$	4	5
$\mathbf{C}$	1	0
$\mathbf{C}$	1	9

}

Split and transpose a vector/column

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

:V1	:V2
A	a
В	b
$\mathbf{C}$	$\mathbf{c}$

# ${\bf Other}$

 ${\bf Skipped}$ 

## Join/Bind data sets

х у

\_unnamed [4 3]:

Id	X1	XY
A	1	x2
В	3	x4
$\mathbf{C}$	5	x6
С	7	x8

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

$\operatorname{Id}$	Y1	XY
A	1	y1
В	3	y3
В	5	y5
D	7	у7

#### Join

Join matching rows from y to x

(api/left-join x y "Id")

left-outer-join [5 6]:

$\overline{\operatorname{Id}}$	X1	XY	right.Id	Y1	right.XY
A	1	x2	A	1	y1
В	3	x4	В	3	y3
В	3	x4	В	5	y5
$\mathbf{C}$	5	x6			
С	7	x8			

Join matching rows from  $\mathbf{x}$  to  $\mathbf{y}$ 

(api/right-join x y "Id")

right-outer-join [4 6]:

$\overline{\operatorname{Id}}$	X1	XY	right.Id	Y1	right.XY
A	1	x2	A	1	y1
В	3	x4	В	3	y3
В	3	x4	В	5	y5
			D	7	y7

Join matching rows from both  $\mathbf{x}$  and  $\mathbf{y}$ 

(api/inner-join x y "Id")

inner-join [3 5]:

Id	X1	XY	Y1	right.XY
A	1	x2	1	y1
В	3	x4	3	y3
В	3	x4	5	y5

Join keeping all the rows

# (api/full-join x y "Id")

full-join [6 6]:

$\overline{\operatorname{Id}}$	X1	XY	right.Id	Y1	right.XY
A	1	x2	A	1	y1
В	3	x4	В	3	y3
В	3	x4	В	5	y5
$\mathbf{C}$	5	x6			
$\mathbf{C}$	7	x8			
			D	7	y7

Return rows from x matching y

```
(api/semi-join x y "Id")
```

semi-join  $[2\ 3]$ :

Id	X1	XY
A	1	x2
В	3	x4

Return rows from  ${\bf x}$  not matching  ${\bf y}$ 

anti-join  $[2\ 3]$ :

$\overline{\mathrm{Id}}$	X1	XY
$\overline{\mathrm{C}}$	5	x6
С	7	x8

# More joins

Select columns while joining

right-outer-join [4 4]:

Id	X1	${\rm right. Id}$	XY
A	1	A	y1
В	3	В	y3
В	3	В	y5
		D	y7

right-outer-join [4 4]:

$\overline{\operatorname{Id}}$	XY	right.Id	right.XY
A	x2	A	y1
В	x4	В	y3
В	x4	В	y5
		D	y7

Aggregate columns while joining

right-outer-join [4 2]:

right.Id	X1Y1
A	1.0
В	24.0
С	NaN
$\mathbf{C}$	NaN

Update columns while joining

right-outer-join [4 4]:

SqX1	right.Id	Y1	XY
1	A	1	y1
9	В	3	y3
9	В	5	y5
	D	7	y7

Adds a list column with rows from y matching x (nest-join)

```
(-> (api/left-join x y "Id")
     (api/drop-columns ["right.Id"])
```

## (api/fold-by (api/column-names x)))

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

XY	X1	Id	Y1	right.XY
x4	3	В	[3 5]	["y3" "y5"]
x6	5	$\mathbf{C}$		
x8	7	$\mathbf{C}$	[]	[]
x2	1	A	[1]	["y1"]

Some joins are skipped

Cross join

```
(def cjds (api/dataset {:V1 [[2 1 1]] :V2 [[3 2]]}))
```

cjds

 $\underline{\phantom{a}}$ unnamed [1 2]:

(reduce #(api/unroll %1 %2) cjds (api/column-names cjds))

 $\underline{\phantom{a}}$ unnamed [6 2]:

$$\begin{array}{c|cc} \hline \\ \hline \\ \hline \\ 2 & 3 \\ 2 & 2 \\ 1 & 3 \\ 1 & 2 \\ 1 & 3 \\ 1 & 2 \\ \end{array}$$

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

:V1	:V2
2	3
2	2
1	3
1	2

# Bind

```
(\texttt{def x (api/dataset } \{: \texttt{V1} \ [1\ 2\ 3]\}))
(def y (api/dataset {:V1 [4 5 6]}))
(def z (api/dataset {:V1 [7 8 9]
                             :V2 [0 0 0]}))
x y z
_unnamed [3 1]:

\frac{:V1}{1} \\
2 \\
3

\underline{\phantom{a}}unnamed [3 1]:
                                                              :V1
                                                              4
5
6
\underline{\phantom{a}}unnamed [3 2]:
                                                          :V1
                                                                 :V2
                                                                  0
                                                          8
                                                                  0
                                                                  0
Bind rows
(api/bind x y)
null [6 1]:
                                                              :V1
                                                              1
                                                              2
3
                                                              4
                                                              5
                                                              6
(api/bind x z)
```

null  $[6\ 2]$ :

:V2
0
0
0

Bind rows using a list

```
(->> [x y]
      (map-indexed #(api/add-column %2 :id (repeat %1)))
      (apply api/bind))
```

null [6 2]:

:V1	:id
1	0
2	0
3	0
4	1
5	1
6	1

Bind columns

```
(api/append x y)
```

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

## Set operations

```
(def x (api/dataset {:V1 [1 2 2 3 3]}))
(def y (api/dataset {:V1 [2 2 3 4 4]}))
x y
```

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

```
\frac{\overline{\text{V1}}}{1}

2
2
3
```

	$\frac{\overline{.V1}}{3}$	
_unnamed [5 1]:		
	:V1 2 2 3 4 4	
Intersection		
<pre>(api/intersect x y)</pre>		
intersection [2 1]:	: <u>V1</u> 2 3	
Difference		
(oni/difference r r)		
<pre>(api/difference x y)</pre>		
difference [1 1]:	:V1 1	
	<del></del>	
difference [1 1]:	<del></del>	
difference [1 1]:  Union	<del></del>	

null [10 1]:

:V1
1
2
2
3
3
2
2
3
4
4

Equality not implemented