# fn make\_row\_by\_row\_fallible

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This proof resides in "contrib" because it has not completed the vetting process.

Proves soundness of make\_row\_by\_row\_fallible in mod.rs at commit f5bb719 (outdated<sup>1</sup>).

make\_row\_by\_row\_fallible returns a Transformation that applies a user-specified function to each record in the input dataset. The function is permitted to return a data-independent error.

#### **Vetting History**

• Pull Request #562

## 1 Hoare Triple

#### Precondition

- DI (input domain) is a type with trait RowByRowDomain<DO>. This trait provides a way to apply a map function to each record in the input dataset to retrieve a dataset that is a member of the output domain, of type DO. The trait further implies that DatasetDomain is also implemented for DI.
- DO (output domain) is a type with trait DatasetDomain. DatasetDomain is used to define the type of the row domain.
- M (metric) is a type with trait DatasetMetric. DatasetMetric is used to restrict the set of valid metrics to those which measure distances between datasets.
- MetricSpace is implemented for (DI, M). Therefore M is a valid metric on DI.
- MetricSpace is implemented for (DO, M).
- row\_function has no side-effects.
- If the input to row\_function is a member of input\_domain's row domain, then the output is a member of output\_row\_domain, or a data-independent error.

#### Pseudocode

<sup>&</sup>lt;sup>1</sup>See new changes with git diff f5bb719..d3f2c25 rust/src/transformations/manipulation/mod.rs

```
# where .translate is defined by the RowByRowDomain trait
9
      output_domain = input_domain.translate(output_row_domain)
11
12
       def function(data: DI_Carrier) -> DO_Carrier:
           \mbox{\tt\#} where .apply_rows is defined by the RowByRowDomain trait
13
           return DI.apply_rows(data, row_function)
14
15
       stability_map = new_stability_map_from_constant(1) #
17
18
      return Transformation (
           \verb"input_domain", output_domain", function",
19
           input_metric, input_metric, stability_map)
```

## Postcondition

Theorem 1.1. For every setting of the input parameters (input\_domain, input\_metric, output\_domain, row\_function, DI, DO, M) to make\_row\_by\_row such that the given preconditions hold, make\_row\_by\_row raises an exception (at compile time or run time) or returns a valid transformation. A valid transformation has the following properties:

- 1. (Appropriate output domain). For every element x in input\_domain, function(x) is in output\_domain or raises a data-independent runtime exception.
- 2. (Stability guarantee). For every pair of elements x, x' in input\_domain and for every pair (d\_in, d\_out), where d\_in has the associated type for input\_metric and d\_out has the associated type for output\_metric, if x, x' are d\_in-close under input\_metric, stability\_map(d\_in) does not raise an exception, and stability\_map(d\_in)  $\leq$  d\_out, then function(x), function(x') are d\_out-close under output\_metric.

## 2 Proofs

Proof. (Part 1 - appropriate output domain). By the definition of RowByRowDomain, DI.apply\_rows(data, row\_function) returns a dataset in input\_domain.translate(output\_row\_domain), if row\_function is a mapping between input\_domain's row domain to output\_row\_domain. This is satisfied by the precondition on row\_function. Thus, for all settings of input arguments, the function returns a dataset in the output domain.

Before proceeding with proving the validity of the stability map, we first provide a lemma.

**Lemma 2.1.** Let f denote the row\_function. For any choice u, v of input arguments in the input domain, and any choice M for which DatasetMetric is implemented for,  $d_M([f(u_1), f(u_2), ...], [f(v_1), f(v_2), ...]) \le d_M([u_1, u_2, ...], [v_1, v_2, ...]).$ 

Proof. Assume WLOG that any source of randomness is fixed when f is computed on u vs v. Given this assumption, and the precondition that f has no side-effects, if  $u_i = v_i$ , then  $f(u_i) = f(v_i)$ . That is, the row function cannot increase the distance between corresponding rows in any adjacent dataset. On the other hand, it is possible for  $f(u_i) = f(v_i)$ , even if  $u_i \neq v_i$ . For example, if f is a constant function, then  $f(u_i) = f(v_i)$  for all i. Therefore, by any of the metrics that DatasetMetric is implemented for, f can only make datasets more similar.

*Proof.* (Part 2 – stability map). Take any two elements u, v in the input\_domain and any pair (d\_in, d\_out), where d\_in has the associated type for input\_metric and d\_out has the associated type for output\_metric. Assume u, v are d\_in-close under input\_metric and that stability\_map(d\_in)  $\leq$  d\_out.

```
d_M(\mathtt{function}(u),\mathtt{function}(v)) = d_M([f(u_1),f(u_2),\ldots],[f(v_1),f(v_2),\ldots])
                                                                                    since {\tt DO} is a {\tt DatasetDomain}
                                  \leq d_M([u_1, u_2, ...], [v_1, v_2, ...])
                                                                                    by 2.1
                                  =d_M(u,v)
                                                                                    since DI is a {\tt DatasetDomain}
                                  = d_in
                                                                                    by the first assumption
                                  by InfCast

    T0.one().inf_mul(T0.inf_cast(d_in))

                                                                                    by InfMul
                                  = stability_map(d_in)
                                                                                    by pseudocode line 16
                                  \leq \texttt{d\_out}
                                                                                    by the second assumption
```

It is shown that function(u), function(v) are  $d_out$ -close under output\_metric.