

# fn make\_count

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This proof resides in “**contrib**” because it has not completed the vetting process.

Proves soundness of **make\_count** in **mod.rs** at commit **f5bb719** (outdated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

**make\_count** returns a Transformation that computes a count of the number of records in a vector. The length of the vector, of type **usize**, is exactly casted to a user specified output type **T0**. If the length is too large to be represented exactly by **T0**, the cast saturates at the maximum value of type **T0**.

## Vetting History

- [Pull Request #513](#)

## 1 Hoare Triple

### Precondition

- TIA (atomic input type) is a type with trait **Primitive**. **Primitive** implies TIA has the trait bound:
  - **CheckNull** so that TIA is a valid atomic type for **AtomDomain**
- T0 (output type) is a type with trait **Number**. **Number** further implies T0 has the trait bounds:
  - **InfSub** so that the output domain is compatible with the output metric
  - **CheckNull** so that T0 is a valid atomic type for **AtomDomain**
  - **ExactIntCast** for casting a vector length index of type **usize** to T0. **ExactIntCast** further implies T0 has the trait bound:
    - \* **ExactIntBounds**, which gives the **MAX\_CONSECUTIVE** value of type T0
  - **One** provides a way to retrieve T0’s representation of 1
  - **DistanceConstant** to satisfy the preconditions of **new\_stability\_map\_from\_constant**

### Pseudocode

```
1 def make_count(  
2     input_domain: VectorDomain[AtomDomain[TIA]],  
3     input_metric: SymmetricDistance  
4 ):  
5     output_domain = AtomDomain(T0) #  
6  
7     def function(data: Vec[TIA]) -> T0: #  
8         size = input_domain.size(data) #  
9         try: #  
10             return T0.exact_int_cast(size) #
```

<sup>1</sup>See new changes with `git diff f5bb719..0abb9d rust/src/transformations/count/mod.rs`

```

11         except FailedCast:
12             return TO.MAX_CONSECUTIVE #
13
14         output_metric = AbsoluteDistance(TO)
15
16         stability_map = new_stability_map_from_constant(TO.one()) #
17
18         return Transformation(
19             input_domain, output_domain, function,
20             input_metric, output_metric, stability_map)

```

## Postcondition

**Theorem 1.1.** For every setting of the input parameters (TIA, TO) to `make_count` such that the given preconditions hold, `make_count` raises an error (at compile time or run time) or returns a valid transformation. A valid transformation has the following properties:

1. (Data-independent runtime errors). For every pair of members  $x$  and  $x'$  in `input_domain`, `invoke(x)` and `invoke(x')` either both return the same error or neither return an error.
2. (Appropriate output domain). For every member  $x$  in `input_domain`, `function(x)` is in `output_domain` or raises a data-independent runtime error.
3. (Stability guarantee). For every pair of members  $x$  and  $x'$  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 error, and `stability_map(d_in) = d_out`, then `function(x), function(x')` are  $d\_out$ -close under `output_metric`.

## 2 Proofs

*Proof. (Part 1 – appropriate output domain).* The `output_domain` is `AtomDomain(TO)`, so it is sufficient to show that `function` always returns non-null values of type `TO`. By the definition of the `ExactIntCast` trait, `TO.exact_int_cast` always returns a non-null value of type `TO` or raises an exception. If an exception is raised, the function returns `TO.MAXIMUM_CONSECUTIVE`, which is also a non-null value of type `TO`. Thus, in all cases, the function (from line 9) returns a non-null value of type `TO`.  $\square$

Before proceeding with proving the validity of the stability map, we provide a couple lemmas.

**Lemma 2.1.**  $|\text{function}(u) - \text{function}(v)| \leq |\text{len}(u) - \text{len}(v)|$ , where `len` is an alias for `input_domain.size`.

*Proof.* By `CollectionDomain`, we know `size` on line 8 is of type `usize`, so it is non-negative and integral. Therefore, by the definition of `ExactIntCast`, the invocation of `TO.exact_int_cast` on line 10 can only fail if the argument is greater than `TO.MAX_CONSECUTIVE`. In this case, the value is replaced with `TO.MAX_CONSECUTIVE`. Therefore,  $\text{function}(u) = \min(\text{len}(u), c)$ , where  $c = \text{TO.MAX_CONSECUTIVE}$ . We use this equality to prove the lemma:

$$\begin{aligned}
 |\text{function}(u) - \text{function}(v)| &= |\min(\text{len}(u), c) - \min(\text{len}(v), c)| \\
 &\leq |\text{len}(u) - \text{len}(v)| \quad \text{since clamping is stable}
 \end{aligned}$$

$\square$

**Lemma 2.2.** For vector  $v$  with each element  $\ell \in v$  drawn from domain  $\mathcal{X}$ ,  $\text{len}(v) = \sum_{z \in \mathcal{X}} h_v(z)$ .

*Proof.* Every element  $\ell \in v$  is drawn from domain  $\mathcal{X}$ , so summing over all  $z \in \mathcal{X}$  will sum over every element  $\ell \in v$ . Recall that the definition of `SymmetricDistance` states that  $h_v(z)$  will return the number of occurrences of value  $z$  in vector  $v$ . Therefore,  $\sum_{z \in \mathcal{X}} h_v(z)$  is the sum of the number of occurrences of each unique value; this is equivalent to the total number of items in the vector.

Since `CollectionDomain` is implemented for `VectorDomain<AtomDomain<TIA>`, we depend on the correctness of the implementation Conditioned on the correctness of the implementation of `CollectionDomain` for `VectorDomain<AtomDomain<TIA>`, the variable `size` is of type `usize` containing the number of elements in `arg`. Therefore,  $\sum_{z \in \mathcal{X}} h_v(z)$  is equivalent to `size`.  $\square$

*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) ≤ d_out`. These assumptions are used to establish the following inequality:

$$\begin{aligned}
|\text{function}(u) - \text{function}(v)| &\leq |\text{len}(u) - \text{len}(v)| && \text{by 2.1} \\
&= \left| \sum_{z \in \mathcal{X}} h_u(z) - \sum_{z \in \mathcal{X}} h_v(z) \right| && \text{by 2.2} \\
&= \left| \sum_{z \in \mathcal{X}} (h_u(z) - h_v(z)) \right| && \text{by algebra} \\
&\leq \sum_{z \in \mathcal{X}} |h_u(z) - h_v(z)| && \text{by triangle inequality} \\
&= d_{Sym}(u, v) && \text{by SymmetricDistance} \\
&\leq d\_in && \text{by the first assumption} \\
&\leq T0.\text{inf\_cast}(d\_in) && \text{by InfCast} \\
&\leq T0.\text{one}().\text{inf\_mul}(T0.\text{inf\_cast}(d\_in)) && \text{by InfMul} \\
&= \text{stability\_map}(d\_in) && \text{by pseudocode line 16} \\
&\leq d\_out && \text{by the second assumption}
\end{aligned}$$

It is shown that `function(u), function(v)` are `d_out`-close under `output_metric`.  $\square$