# impl SelectionMeasure for RangeDivergence

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April 29, 2025

# 1 Hoare Triple

### Precondition

#### Compiler-verified

- Associated Type RV
  - RV must implement trait InverseCDF.
- Method random\_variable Types consistent with pseudocode.
- $\bullet \ \ {\rm Method} \ {\tt privacy\_map} \ \ {\it Types} \ consistent \ with \ pseudocode.$

#### Caller-verified

- Associated Type RV (no caller verified preconditions)
- Method random\_variable
  - scale is positive (cannot be null due to FBig dtype).
- Method privacy\_map
  - d\_in is non-null and non-negative.
  - scale is non-null and non-negative.

#### Pseudocode

```
class RangeDivergence(SelectionMeasure):
      ONE_SHOT = True
      RV = GumbelRV
      @staticmethod
      def random_variable(shift: FBig, scale: FBig) -> GumbelRV:
          return GumbelRV(shift=shift, scale=scale)
      @staticmethod
9
      def privacy_map(d_in: f64, scale: f64, k: usize) -> f64:
10
          if d_in < 0:</pre>
11
               raise ValueError("input distance must be non-negative")
12
13
          if scale.is_zero():
14
15
               return f64.INFINITY
16
          return d_in.inf_div(scale).inf_mul(f64.inf_cast(k))
```

#### Postcondition

Theorem 1.1. The implementation is consistent with all associated items in the SelectionMeasure trait.

- 1. Associated Type RV
- 2. Method random\_variable
- 3. Method privacy\_map

**Definition 1.2.** A random variable follows the Gumbel distribution if it has density

$$f(x) = \frac{1}{\beta} e^{-e^{-e^{-z}} - z} \tag{1}$$

where  $z = \frac{x-\mu}{\beta}$ ,  $\mu$  is the location parameter and  $\beta$  is the scale parameter.

Proof of valid associated type: RV. The associated type RV is defined as GumbelRV, which represents a random variable following the Gumbel distribution 1.2. The compiler verifies that GumbelRV implements the InverseCDF trait.

*Proof of valid method:* random\_variable. By the precondition on scale being positive, random\_variable returns a valid instance of GumbelRV.

Proof of valid method:  $privacy_map$ . By Lemma 4.2 of [2],  $\mathcal{M}_{Gumbel}^k(x)$  is equal in distribution to the peeling exponential mechanism, which is the k-fold composition of the exponential mechanism. Proposition 2 of [1] shows that the exponential mechanism satisfies the RangeDivergence privacy guarantee.

## References

- [1] Jinshuo Dong, David Durfee, and Ryan Rogers. Optimal differential privacy composition for exponential mechanisms and the cost of adaptivity. *CoRR*, abs/1909.13830, 2019.
- [2] David Durfee and Ryan Rogers. Practical differentially private top-k selection with pay-what-you-get composition. *CoRR*, abs/1905.04273, 2019.