# fn make\_gaussian\_threshold

## Michael Shoemate

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

Proves soundness of the implementation of make\_gaussian\_threshold in mod.rs at commit f5bb719 (outdated1).

Thresholded noise mechanisms may be parameterized along many different axes:

- key dtype: i8, i16, i32, i64, u8, u16, u32, u64, f32, f64, UBig, IBig, RBig
- metric dtype: i8, i16, i32, i64, u8, u16, u32, u64, f32, f64, UBig, IBig, RBig
- measure: max divergence, zero concentrated divergence
- distribution: laplace, gaussian

All parameterizations reduce to a single core mechanism that perturbs a signed big integers with noise sampled from the appropriate discrete distribution, and then thresholds and shuffles the result.

The implementation of this function constructs a random variable denoting the noise distribution to add, and then dispatches to the MakeNoiseThreshold<DI, MI, MO> trait which constructs the core mechanism and wraps it in pre-processing transformations and post-processors to match the desired parameterization.

# 1 Hoare Triple

### Precondition

### Compiler-Verified

- generic DI implements trait NoiseDomain
- generic MI implements trait Metric
- generic MO implements trait Measure
- generic DiscreteGaussian implements trait MakeNoiseThreshold
- type (DI, MI) implements trait MetricSpace

#### **User-Verified**

#### None

 $<sup>^{1}\</sup>mathrm{See}\ \mathrm{new}\ \mathrm{changes}\ \mathrm{with}\ \mathrm{git}\ \mathrm{diff}\ \mathrm{f5bb719..8cbc2c7f}\ \mathrm{rust/src/measurements/noise\_threshold/distribution/gaussian/mod.rs}$ 

## Pseudocode

## Postcondition

#### Theorem 1.1.

Theorem 1.2. For every setting of the input parameters (input\_domain, input\_metric, scale, threshold, k, DI, MI, MO) to make\_gaussian\_threshold such that the given preconditions hold, make\_gaussian\_threshold raises an exception (at compile time or run time) or returns a valid measurement. A valid measurement has the following property:

1. (Privacy 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\_measure, if x, x' are d\_in-close under input\_metric, privacy\_map(d\_in) does not raise an exception, and privacy\_map(d\_in)  $\leq$  d\_out, then function(x), function(x') are d\_out-close under output\_measure.

*Proof.* We first construct a random variable DiscreteGaussian representing the desired noise distribution. Since MakeNoiseThreshold.make\_noise\_threshold has no preconditions, the postcondition follows, which matches the postcondition for this function.