# MakeNoise<VectorDomain<AtomDomain<T>, LpDistance<P, QI>, MO> for IntExpFamily<P>

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

Proves soundness of the implementation of MakeNoise over vectors for IntExpFamily in mod.rs at commit f5bb719 (outdated1).

This mechanism samples from the IntExpFamily distribution, where the tails are clipped to the smallest and largest representable integers in the native integer type. This is done by first sampling from the ZExpFamily, the equivalent distribution supported on all integers, and then clipping the tails by clamping. The clamping is done by saturating the cast of the sampled value to the native integer type.

# 1 Hoare Triple

#### Precondition

### Compiler-Verified

- Generic T implements trait Integer and SaturatingCast<IBig> The saturating cast is for infallible postprocessing of big into back to type T.
- Const-generic P is of type usize
- Generic QI implements trait Integer
- Generic MO implements trait Measure
- Type IBig implements trait From<T>. This infallible exact cast is for converting integers to big ints in the preprocessing transformation.
- Type RBig implements trait TryFrom<QI>. This is for fallible casting from input sensitivity of type QI to a rational in the privacy map.
- Type ZExpFamily<P> implements trait NoisePrivacyMap<LpDistance<P, RBig>, MO>. This bound requires that it must be possible to construct a privacy map for this combination of noise distribution, distance type and privacy measure.

#### **User-Verified**

None

 $<sup>^1\</sup>mathrm{See}$  new changes with git diff f5bb719..4500163 rust/src/measurements/noise/nature/integer/mod.rs

#### Pseudocode

```
class IntExpFamily:
    def make_noise(
        self, input_space: tuple[VectorDomain[AtomDomain[T]], LpDistance[P, QI]]

) -> Measurement[VectorDomain[AtomDomain[T]], T, LpDistance[P, QI], MO]:
    distribution = ZExpFamily(
        scale=integerize_scale(self.scale, 0)
    ) #

t_int = make_int_to_bigint(input_space)
    m_noise = distribution.make_noise(t_int.output_space())
    return t_int >> m_noise >> then_saturating_cast()
```

## Postcondition

Theorem 1.1. For every setting of the input parameters (self, input\_space, MO, T, P, QI) to make\_noise such that the given preconditions hold, make\_noise raises an error (at compile time or run time) or returns a valid measurement. A valid measurement 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. (Privacy 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\_measure, if x, x' are d\_in-close under input\_metric, privacy\_map(d\_in) does not raise an error, and privacy\_map(d\_in) = d\_out, then function(x), function(x') are d\_out-close under output\_measure.

*Proof.* Line 7 constructs a new random variable following a distribution equivalent to IntExpFamily, but without clipped tails.

Neither constructor make\_int\_to\_bigint nor MakeNoise.make\_noise have manual preconditions, and the postconditions guarantee a valid transformation and valid measurement, respectively. then\_saturating\_cast also does not have preconditions, and its postcondition guarantees that it returns a valid postprocessor.

The chain of a valid transformation, valid measurement and valid postprocessor is a valid measurement.