

# Combine by Example

OpenTelemetry Exponential Histograms

By Alex Van Boxel



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Austin, 2022



# Combine are just more Fn

# OpenTelemetry



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# OpenTelemetry

**An observability framework for cloud-native software.**

OpenTelemetry is a collection of tools, APIs, and SDKs. You use it to instrument, generate, collect, and export telemetry data (metrics, logs, and traces) for analysis in order to understand your software's performance and behavior.

# Exponential

Histogram, what is it?

The magic of scale



# Power of math

$$\text{base} = (2^{(2^{\text{-scale}})})$$
$$(\text{base}^{\text{index}}) \leq \text{value} < (\text{base}^{\text{(index+1)}})$$



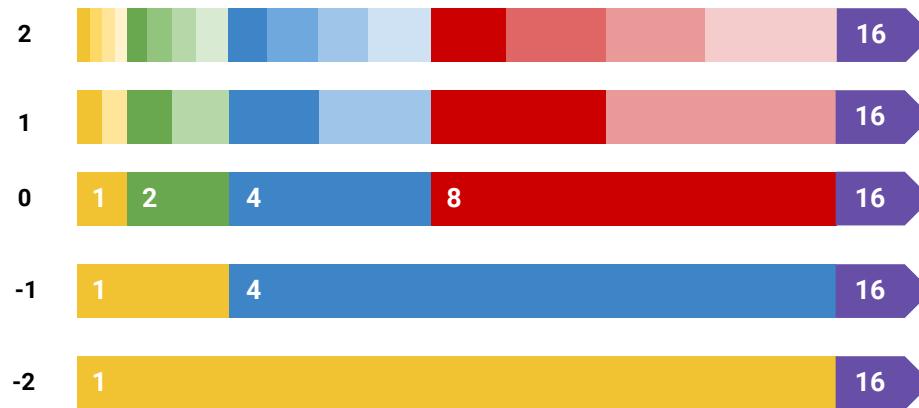
# Scale

scale	-2	-1	0	1	2	3	4	5
base	16	4	2	1.414213562	1.189207115	1.090507733	1.044273782	1.021897149
0	1	1	1	1	1	1	1	1
1	16	4	2	1.414213562	1.189207115	1.090507733	1.044273782	1.021897149
2	256	16	4	2	1.414213562	1.189207115	1.090507733	1.044273782
3	4096	64	8	2.828427125	1.681792831	1.296839555	1.138788635	1.067140401
4	65536	256	16	4	2	1.414213562	1.189207115	1.090507733
5	1048576	1024	32	5.656854249	2.37841423	1.542210825	1.241857812	1.114386743
6	16777216	4096	64	8	2.828427125	1.681792831	1.296839555	1.138788635
7	268435456	16384	128	11.3137085	3.363585661	1.834008086	1.354255547	1.163724859
8	4294967296	65536	256	16	4	2	1.414213562	1.189207115
9	68719476736	262144	512	22.627417	4.75682846	2.181015465	1.476826146	1.21524736
10	1099511627776	1048576	1024	32	5.656854249	2.37841423	1.542210825	1.241857812
11	1759218604441	4194304	2048	45.254834	6.727171322	2.593679109	1.610490332	1.269050957
12	2814749767106	16777216	4096	64	8	2.828427125	1.681792831	1.296839555
13	4.5036E+15	67108864	8192	90.50966799	9.51365692	3.084421651	1.75625216	1.325236643
14	7.20576E+16	268435456	16384	128	11.3137085	3.363585661	1.834008086	1.354255547
15	1.15292E+18	1073741824	32768	181.019336	13.45434264	3.668016173	1.915206561	1.383909882
16	1.84467E+19	4294967296	65536	256	16	4	2	1.414213562
17	2.95148E+20	17179869184	131072	362.038672	19.02731384	4.362030931	2.088547565	1.445180807



# Scale

scale



# Protobuf

The OpenTelemetry Exponential Histogram

A deep dive into the Exponential Histograms original Protobuf



# OpenTelemetry Metric Protobuf

```
// ExponentialHistogramDataPoint is a single data point in a timeseries that describes the  
// time-varying values of a ExponentialHistogram of double values. A ExponentialHistogram contains  
// summary statistics for a population of values, it may optionally contain the  
// distribution of those values across a set of buckets.
```

```
message ExponentialHistogramDataPoint {  
  
    // The set of key/value pairs that uniquely identify the timeseries from  
    // where this point belongs. The list may be empty (may contain 0 elements).  
    repeated opentelemetry.proto.common.v1.KeyValue attributes = 1;
```



# OpenTelemetry Metric Protobuf

```
fixed64 start_time_unix_nano = 2;    // StartTimeUnixNano is optional but strongly encouraged.  
  
fixed64 time_unix_nano = 3;    // TimeUnixNano is required.  
  
fixed64 count = 4;    // count is the number of values in the population.  
  
optional double sum = 5;    // sum of the values in the population. If count is zero then this field  
optional double min = 12;   // min is the minimum value over (start_time, end_time].  
optional double max = 13;   // max is the maximum value over (start_time, end_time].  
}
```



# OpenTelemetry Metric Protobuf

```
// scale describes the resolution of the histogram.  
sint32 scale = 6;  
  
// zero_count is the count of values that are either exactly zero or  
fixed64 zero_count = 7;  
  
// positive carries the positive range of exponential bucket counts.  
Buckets positive = 8;  
  
// negative carries the negative range of exponential bucket counts.  
Buckets negative = 9;
```



# OpenTelemetry Metric Protobuf

```
// Buckets are a set of bucket counts, encoded in a contiguous array
// of counts.

message Buckets {

    // Offset is the bucket index of the first entry in the bucket_counts array.
    sint32 offset = 1;    // Note: This uses a varint encoding as a simple form of compression.

    // Count is an array of counts, where count[i] carries the count of the bucket
    // at index (offset+i).  count[i] is the count of
    // values greater than or equal to base^(offset+i) and less than base^(offset+i+1).
    // This field is expected to have many buckets, especially zeros, so uint64 has been selected to ensure
    // varint encoding.

    repeated uint64 bucket_counts = 2;
}
```

# Design Decisions

# Implementing

CombineFn



# Implementing a CombineFn

```
class CombineFn<InputT, AccumT, OutputT> {  
  
    // Adds the given input value to the given accumulator, returning the new accumulator value.  
    public abstract AccumT addInput(AccumT mutableAccumulator, InputT input);  
  
    // Returns an accumulator representing the accumulation of all the input values accumulated in  
    // the merging accumulators. (only the first accumulator can be modified)  
    public abstract AccumT mergeAccumulators(Iterable<AccumT> accumulators);  
  
    // Returns the output value that is the result of combining all the input values represented by  
    // the given accumulator. (accumulator can be modified for efficiency)  
    public abstract OutputT extractOutput(AccumT accumulator);
```



# Implementing a CombineFn

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    // Returns the output value that is the result of combining all the input values represented by  
    // the given accumulator. (accumulator can be modified for efficiency)  
    public abstract OutputT extractOutput(AccumT accumulator);
```

# Design Choice

No Proto as the accumulator

Why not to use the pure protobuf as  
an accumulator?



# Mutable

```
/** Mutable Datapoint Accumulator for Exponential Histograms */
public final class ExponentialHistogramDatapointAccum implements Serializable {

    private final int scale;

    private final double base;

    ArrayList<Long> positiveBucket;

    long zeroBucket = 0;

    double sum = 0;
    long count = 0;
```

# Design Choice

**Output is an accumulator**

Why is the Protobuf not the output?



# Accum out

```
public abstract class ExponentialHistogramCombineFn<T>
    extends Combine.CombineFn<
        T, ExponentialHistogramDatapointAccum, ExponentialHistogramDatapointAccum> {

    @Override
    public ExponentialHistogramDatapointAccum extractOutput(
        ExponentialHistogramDatapointAccum accumulator) {
        return accumulator;
    }
}
```

# Design Choice

Separate Accumulator to Proto  
**DoFn**

What about time?



# Need for time

**@ProcessElement**

```
public void exponentialHistogramDataPointAccum2MetricFn(  
    ProcessContext context, BoundedWindow window) {  
  
    ExponentialHistogramDataPoint.Builder pbDataPointBuilder =  
        MetricUtil.add1MinuteBounds(  
            ExponentialHistogramDataPoint.newBuilder(),  
            window.maxTimestamp().getMillis());
```

# Design Choice

Put most of the logic in the  
accumulator

Accumulators are mutable anyway



# Most of the logic in the Accum

```
public void addValue(double value) {  
    if (value < 1.0) {  
        increaseZeroCount(1);  
    } else {  
        increaseIndexCount(indexOf(value), 1);  
    }  
    count++;  
    sum += value;  
}  
  
private int indexOf(double value) {  
    Preconditions.checkArgument(value >= 1.0, "indexOf must not be called for values < 1.0");  
    return (int) (Math.log(value) / Math.log(base));  
}
```



# Keep Combine simple

```
@Override
public ExponentialHistogramDatapointAccum mergeAccumulators(
    Iterable<ExponentialHistogramDatapointAccum> accumulators) {
    // Search for the lowest scale, so no unnecessary compression needs to happen mid-calculation
    int scale = preferredScale;
    for (ExponentialHistogramDatapointAccum accumulator : accumulators) {
        scale = Math.min(scale, accumulator.getScale());
    }
    ExponentialHistogramDatapointAccum mutable = new ExponentialHistogramDatapointAccum(scale);
    // Add all histograms using the add function in the accumulators
    for (ExponentialHistogramDatapointAccum accumulator : accumulators) {
        mutable.add(accumulator);
    }
    return mutable;
}
```



# Keep Combine simple

```
@Override
public ExponentialHistogramDatapointAccum mergeAccumulators(
    Iterable<ExponentialHistogramDatapointAccum> accumulators) {
    // Search for the lowest scale, so no unnecessary compression needs to happen mid-calculation
    int scale = preferredScale;
    for (ExponentialHistogramDatapointAccum accumulator : accumulators) {
        scale = Math.min(scale, accumulator.
    }
    ExponentialHistogramDatapointAccum mutable = new ExponentialHistogramDatapointAccum(scale);
    // Add all histograms using the add function in the accumulators
    for (ExponentialHistogramDatapointAccum accumulator : accumulators) {
        mutable.add(accumulator);
    }
    return mutable;
}
```

Opportunity for optimization

# Design Choice

Have an **abstract** Combine

Exponential Histogram takes a  
double



# Abstract Combine

```
@Override  
public abstract ExponentialHistogramDatapointAccum addInput(  
    ExponentialHistogramDatapointAccum mutableAccumulator, T value);  
  
public class DoubleToExponentialHistogramDataPointCombineFn  
    extends ExponentialHistogramCombineFn<Double> {  
  
    @Override  
    public ExponentialHistogramDatapointAccum addInput(  
        ExponentialHistogramDatapointAccum mutableAccumulator, Double value) {  
        mutableAccumulator.addValue(value);  
        return mutableAccumulator;  
    }  
}
```



# Abstract Combine

```
@Override  
public abstract ExponentialHistogramDatapointAccum addInput(  
    ExponentialHistogramDatapointAccum mutableAccumulator, T value);  
  
public class ExponentialHistogramDataPointAccumCombineFn  
    extends ExponentialHistogramCombineFn<ExponentialHistogramDatapointAccum> {  
  
    @Override  
    public ExponentialHistogramDatapointAccum addInput(  
        ExponentialHistogramDatapointAccum mutableAccumulator,  
        ExponentialHistogramDatapointAccum value) {  
        mutableAccumulator.add(value);  
        return mutableAccumulator;  
    }  
}
```

# Combine in action



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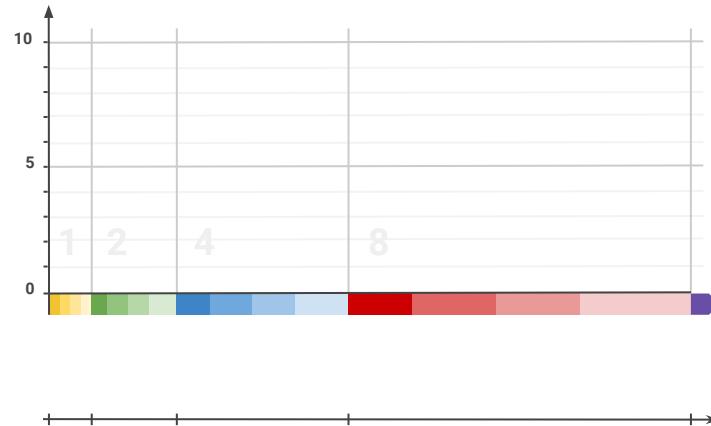
# Combine

Doubles into Exponential  
Histograms

Visual guide to Exponential  
Histograms



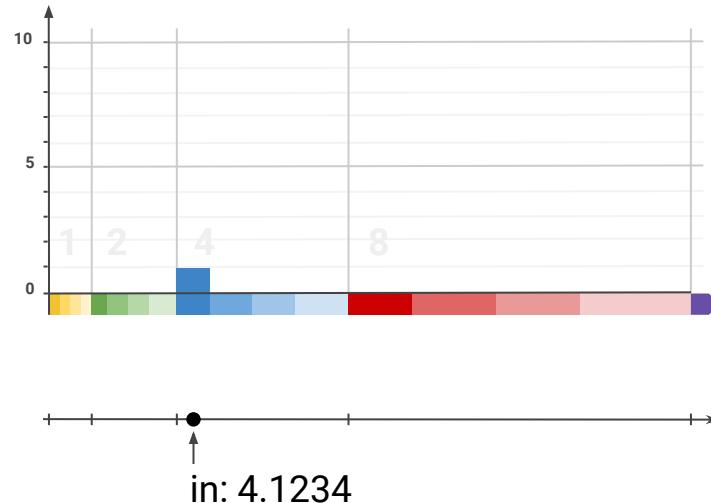
# From Double to Exp. Histogram



<b>SUM</b>	0
<b>COUNT</b>	0
<b>AVG</b>	0
<b>SIZE (by)</b>	32
<b>SIZE in (by)</b>	0



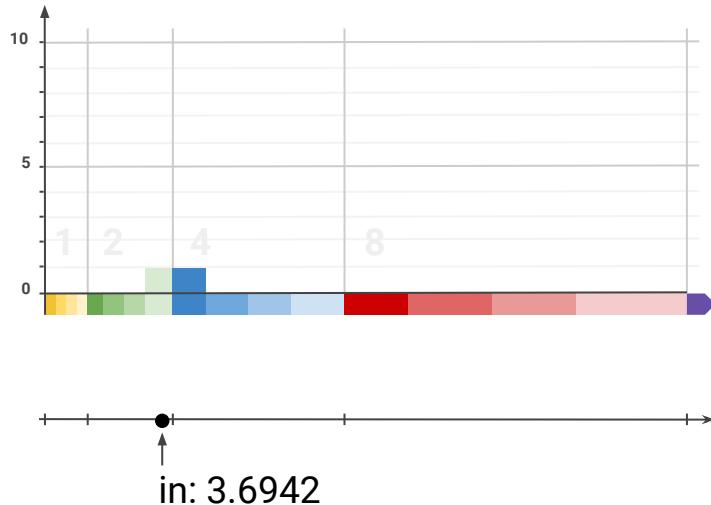
# From Double to Exp. Histogram



<b>SUM</b>	4.1234
<b>COUNT</b>	1
<b>AVG</b>	4.1234
<b>SIZE (by)</b>	68 (32+36)
<b>SIZE in (by)</b>	4



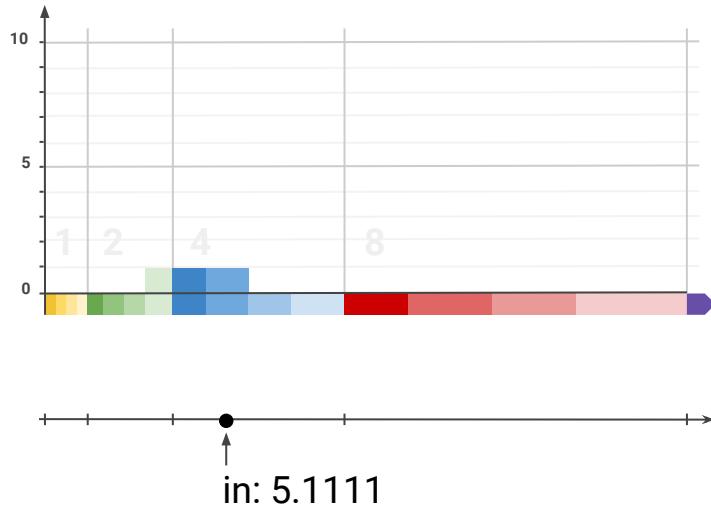
# From Double to Exp. Histogram



<b>SUM</b>	7.8176
<b>COUNT</b>	2
<b>AVG</b>	3.9088
<b>SIZE (by)</b>	68 (32+36)
<b>SIZE in (by)</b>	8



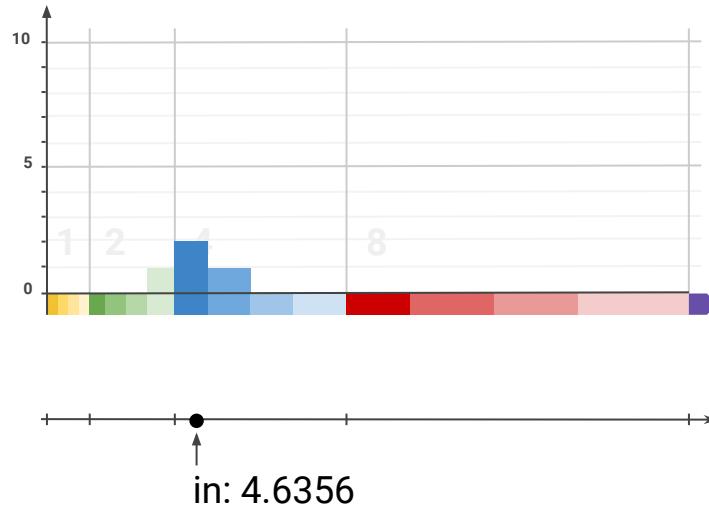
# From Double to Exp. Histogram



<b>SUM</b>	12.9287
<b>COUNT</b>	3
<b>AVG</b>	4.3096
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	12



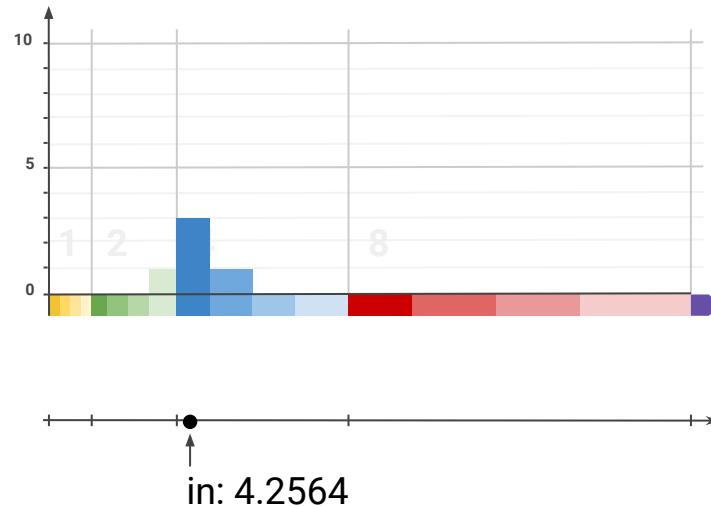
# From Double to Exp. Histogram



<b>SUM</b>	17.5643
<b>COUNT</b>	4
<b>AVG</b>	4.3911
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	16



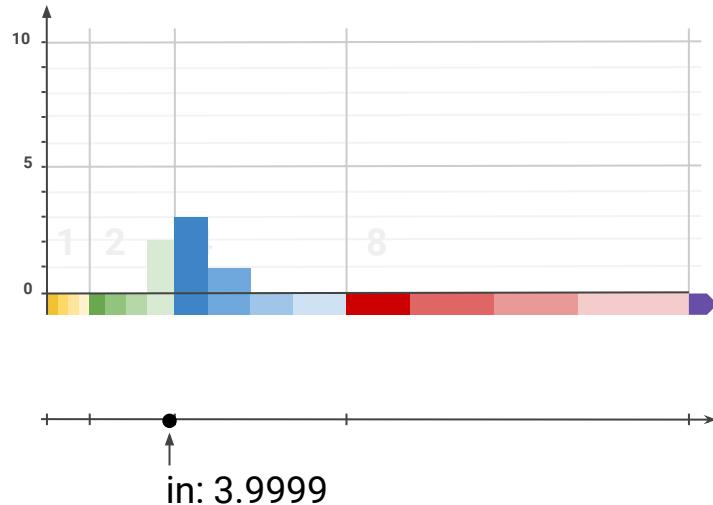
# From Double to Exp. Histogram



<b>SUM</b>	21.8207
<b>COUNT</b>	5
<b>AVG</b>	4.3641
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	20



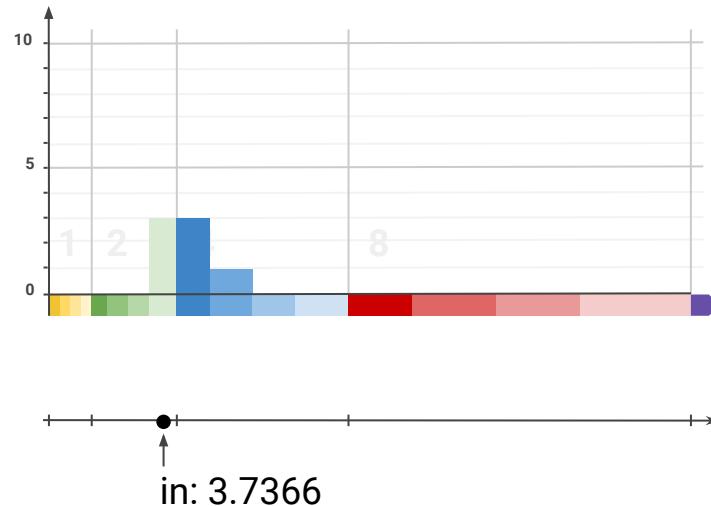
# From Double to Exp. Histogram



<b>SUM</b>	25.8206
<b>COUNT</b>	6
<b>AVG</b>	4.3034
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	24



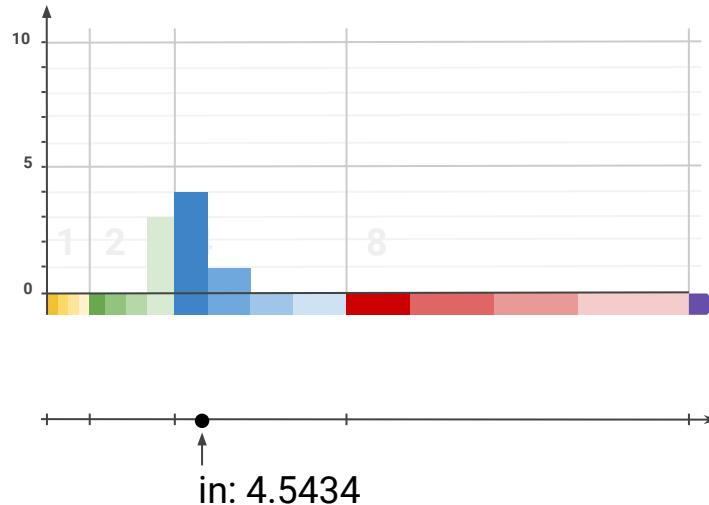
# From Double to Exp. Histogram



<b>SUM</b>	29.5572
<b>COUNT</b>	7
<b>AVG</b>	4.2225
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	28



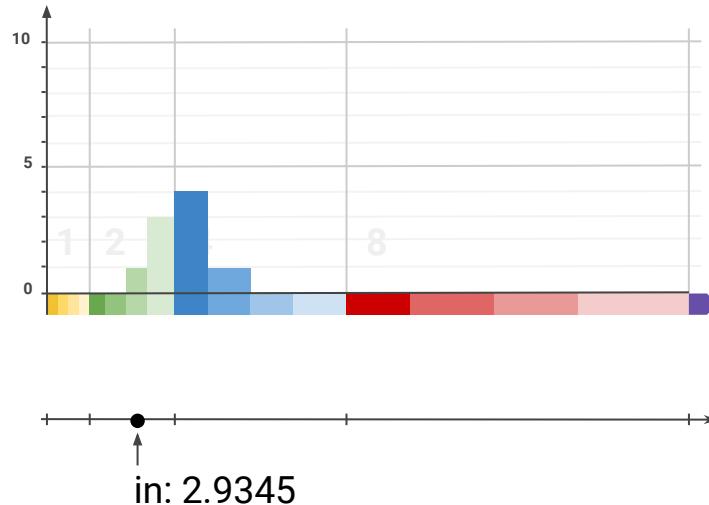
# From Double to Exp. Histogram



<b>SUM</b>	34.1006
<b>COUNT</b>	8
<b>AVG</b>	4.2626
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	32



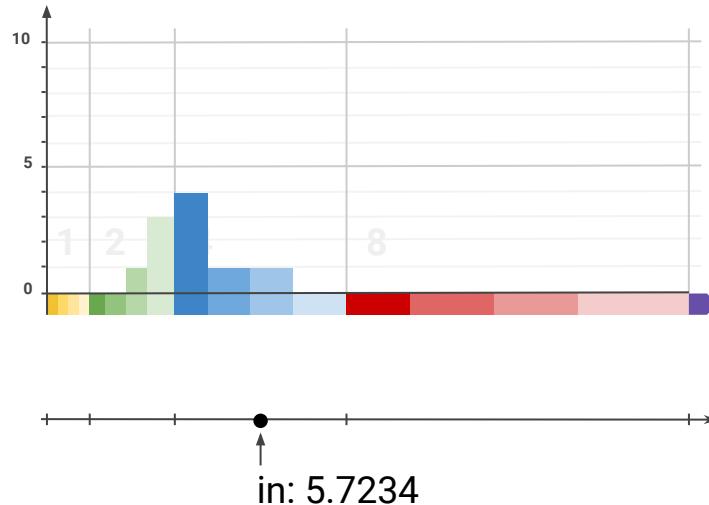
# From Double to Exp. Histogram



<b>SUM</b>	37.0351
<b>COUNT</b>	9
<b>AVG</b>	4.1150
<b>SIZE (by)</b>	72 (32+40)
<b>SIZE in (by)</b>	36



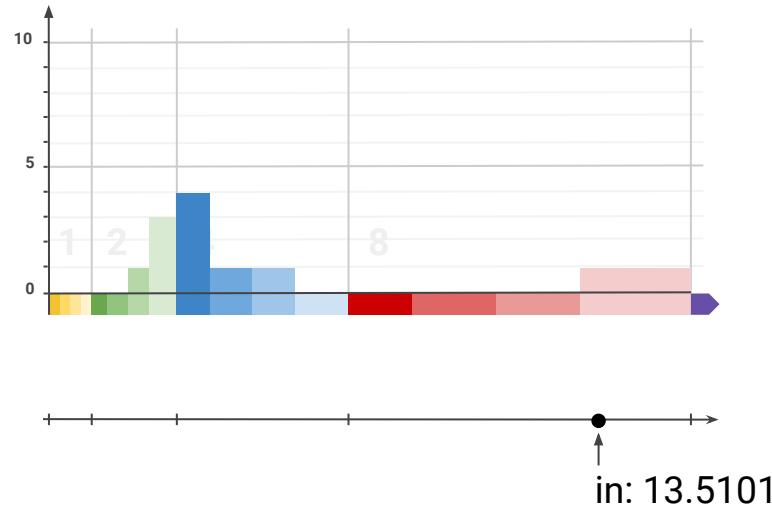
# From Double to Exp. Histogram



<b>SUM</b>	42.7585
<b>COUNT</b>	10
<b>AVG</b>	4.2759
<b>SIZE (by)</b>	76 (32+44)
<b>SIZE in (by)</b>	40



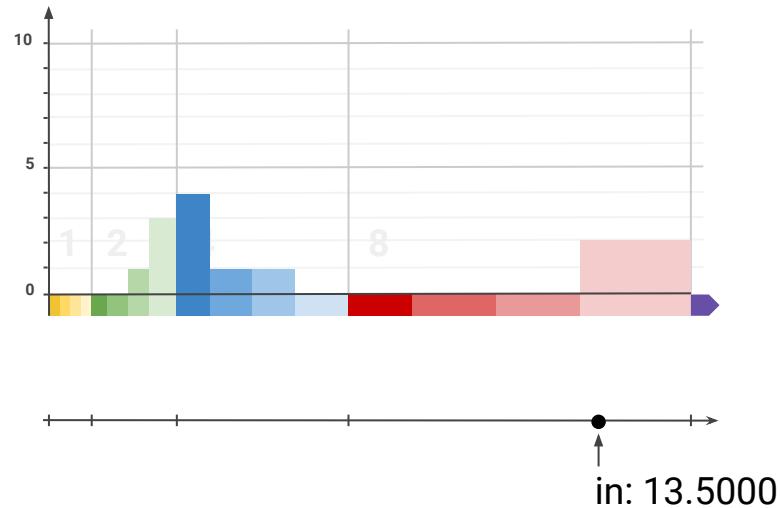
# From Double to Exp. Histogram



<b>SUM</b>	42.7585
<b>COUNT</b>	11
<b>AVG</b>	5.1153
<b>SIZE (by)</b>	96 (32+64)
<b>SIZE in (by)</b>	44



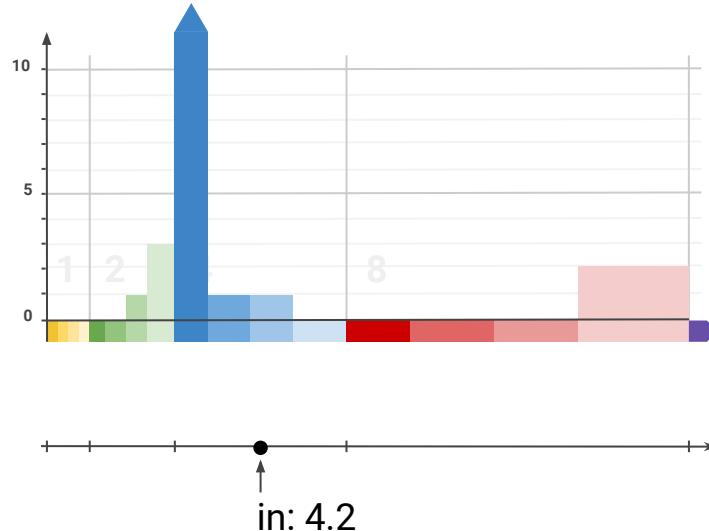
# From Double to Exp. Histogram



<b>SUM</b>	42.7585
<b>COUNT</b>	12
<b>AVG</b>	5.8141
<b>SIZE (by)</b>	96 (32+64)
<b>SIZE in (by)</b>	48



# From Double to Exp. Histogram



<b>SUM</b>	153.7686
<b>COUNT</b>	32
<b>AVG</b>	4.8053
<b>SIZE (by)</b>	96 (32+64)
<b>SIZE in (by)</b>	128

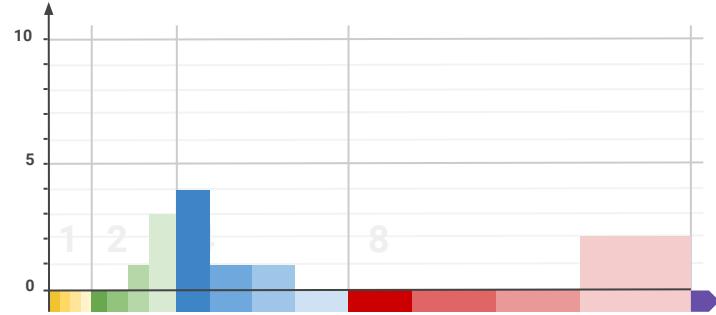
# Combine

Exponential Histograms with  
each other

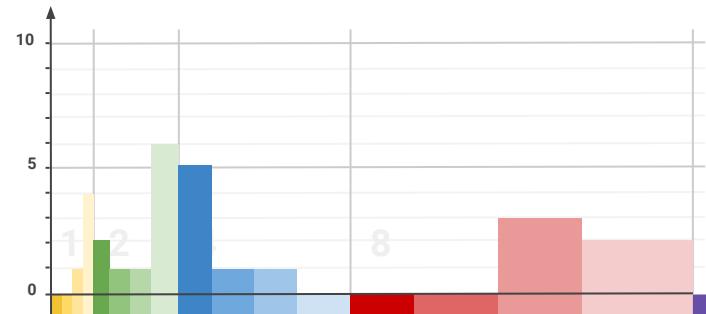
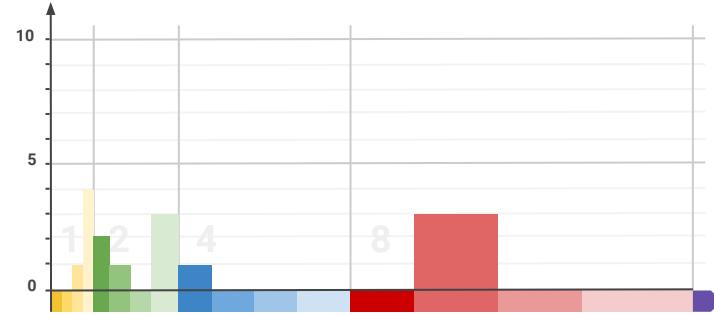
Visual guide to Exponential  
Histograms



# Combine with same scale

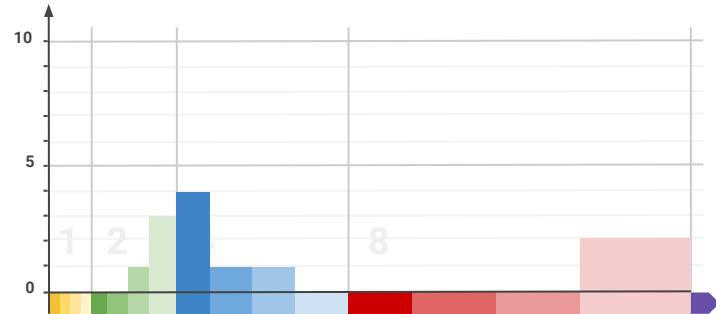


combine  
↓

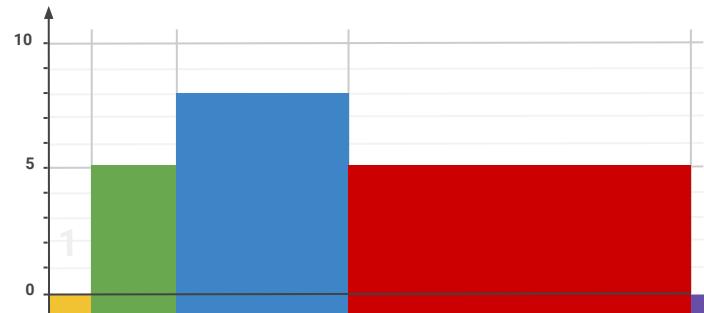
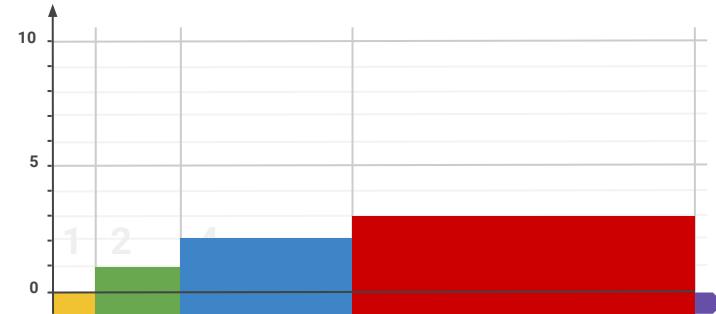




# Combine with different scale



combine  
↓

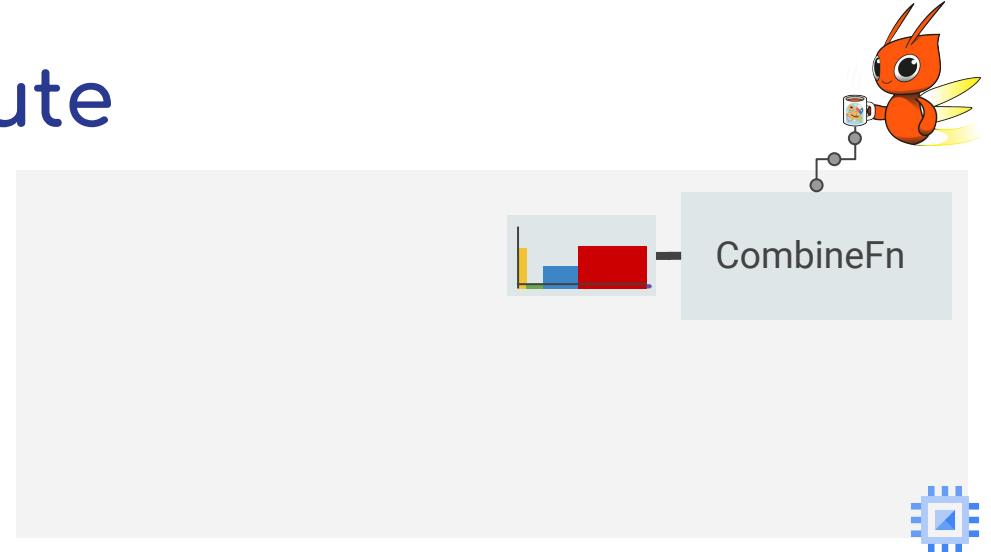


# Combine

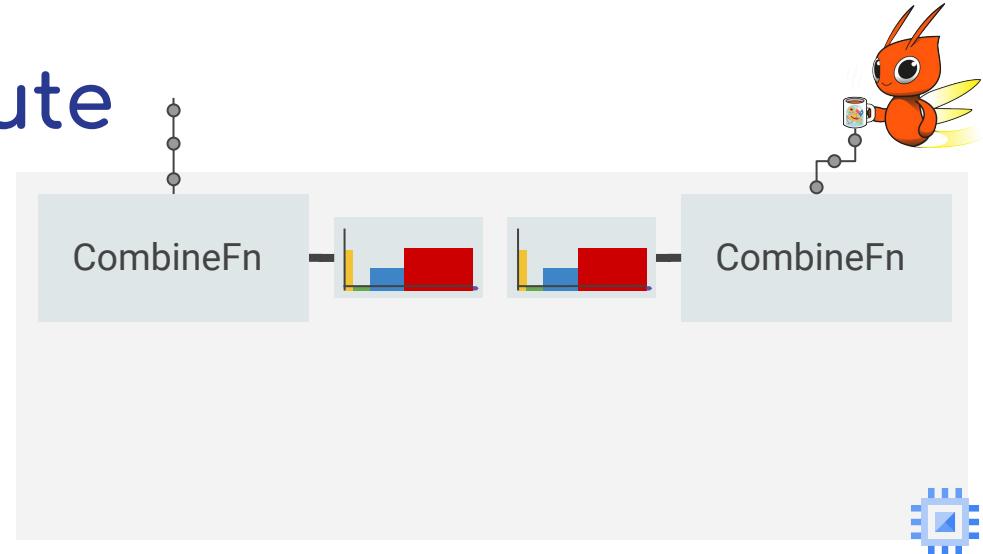
on the compute

Visual guide to Exponential  
Histograms

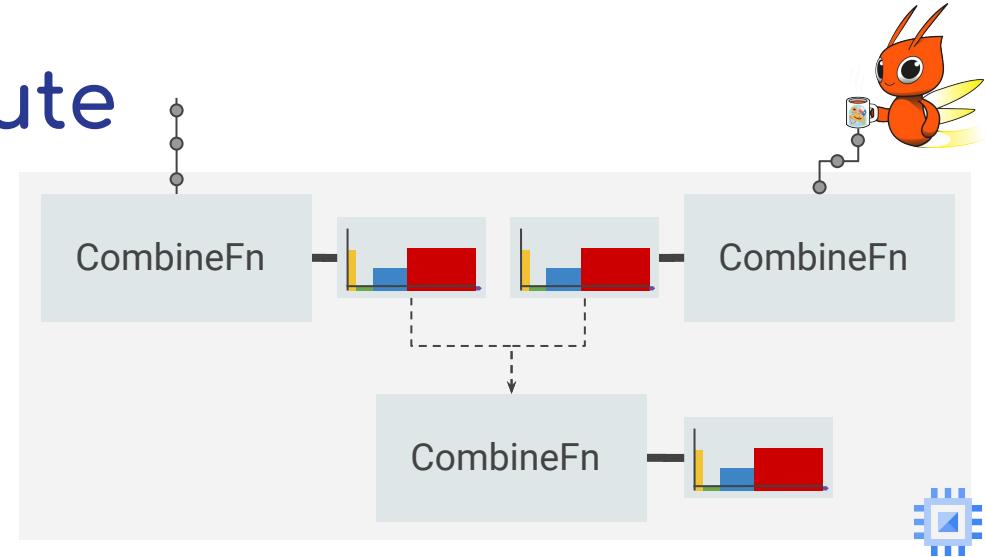
# Combine on compute



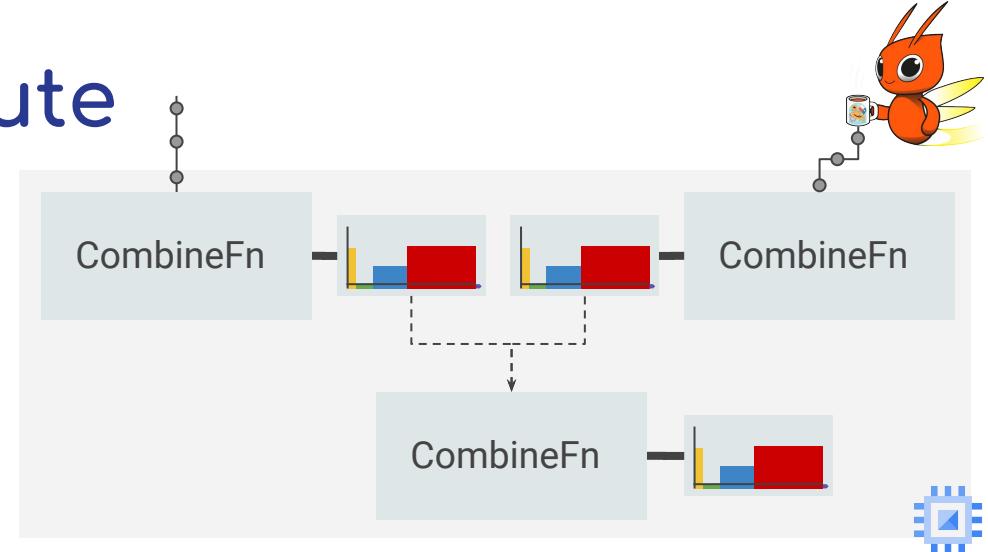
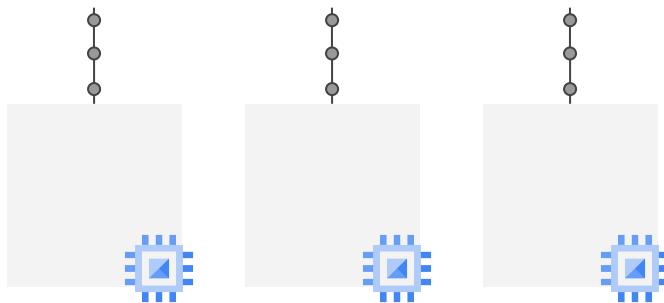
# Combine on compute



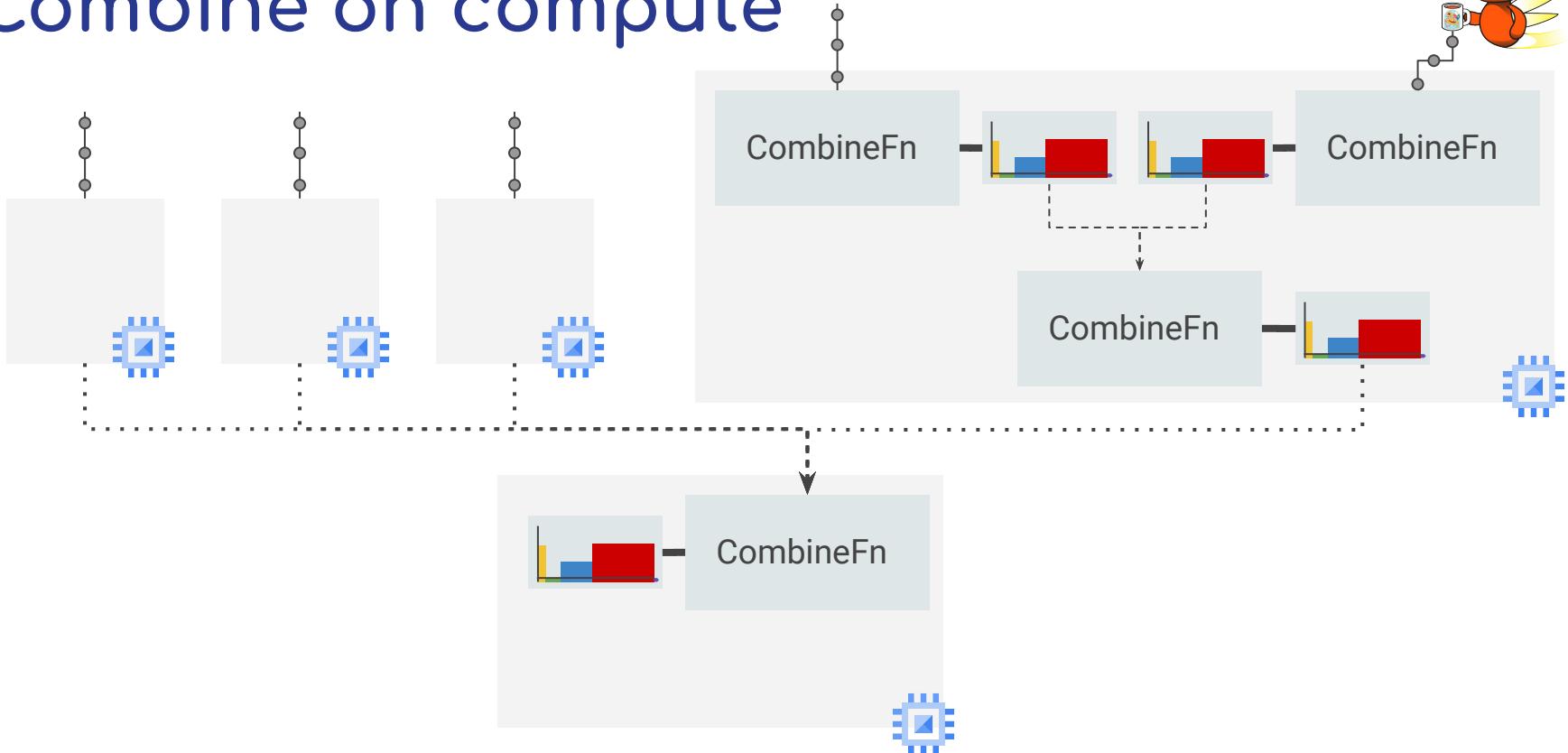
# Combine on compute



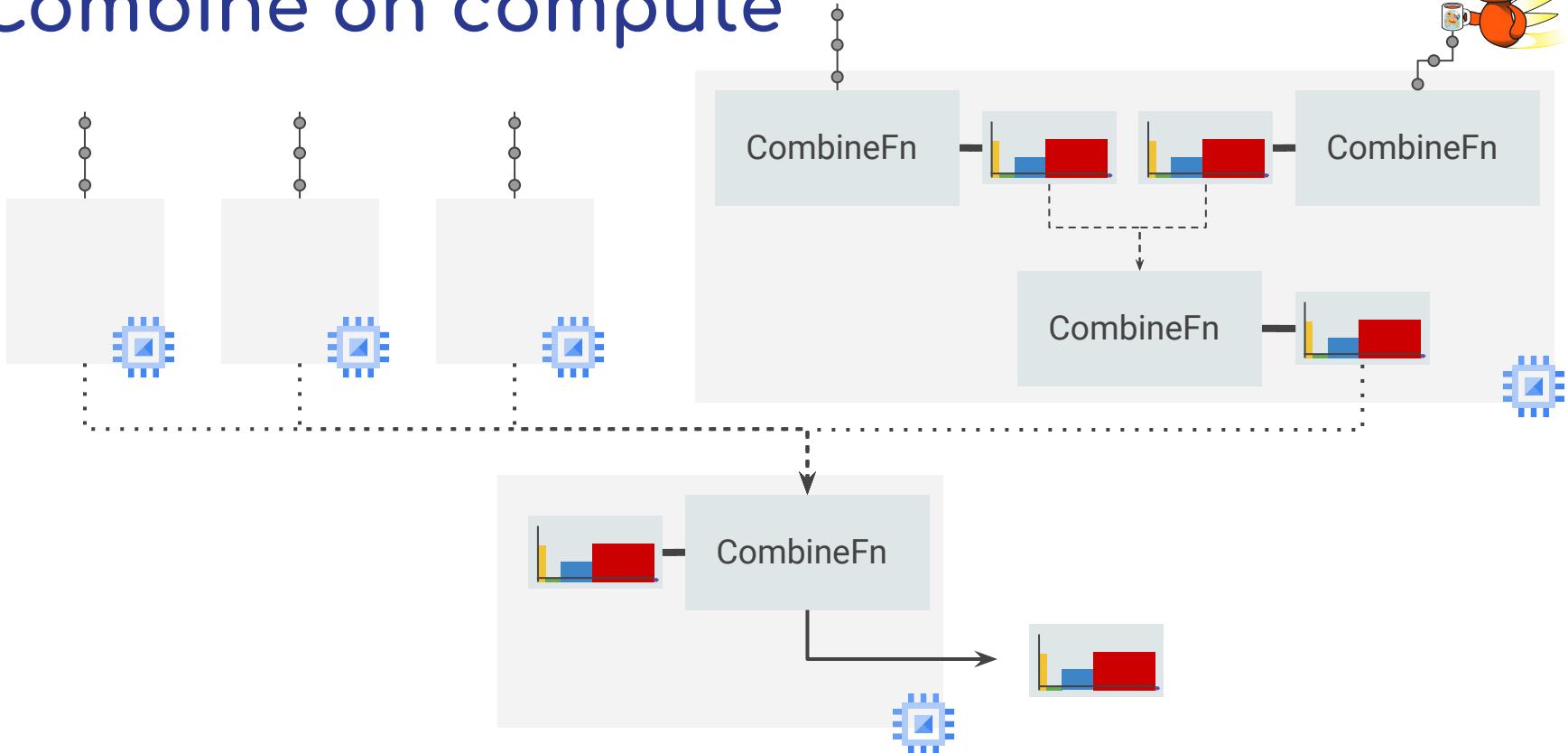
# Combine on compute



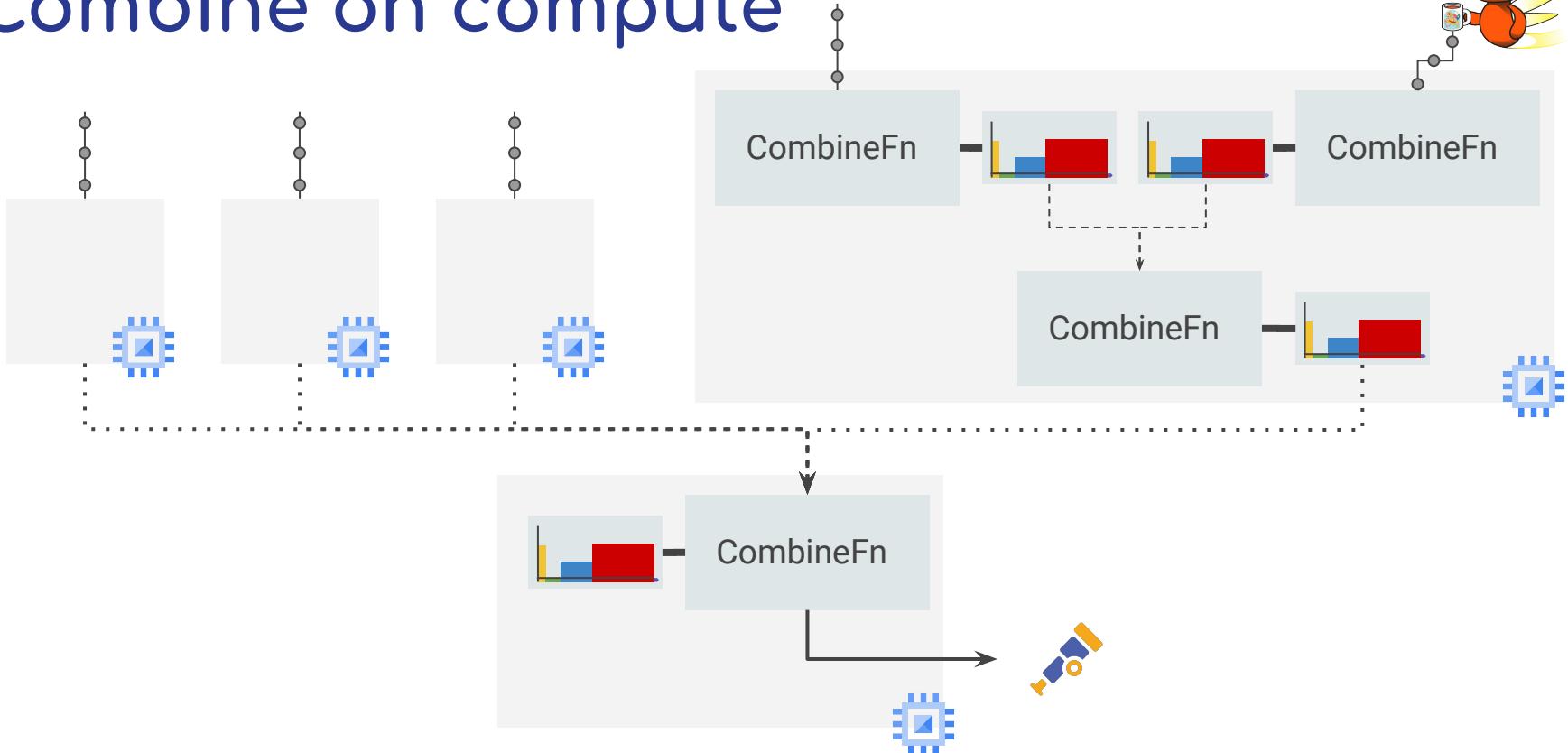
# Combine on compute



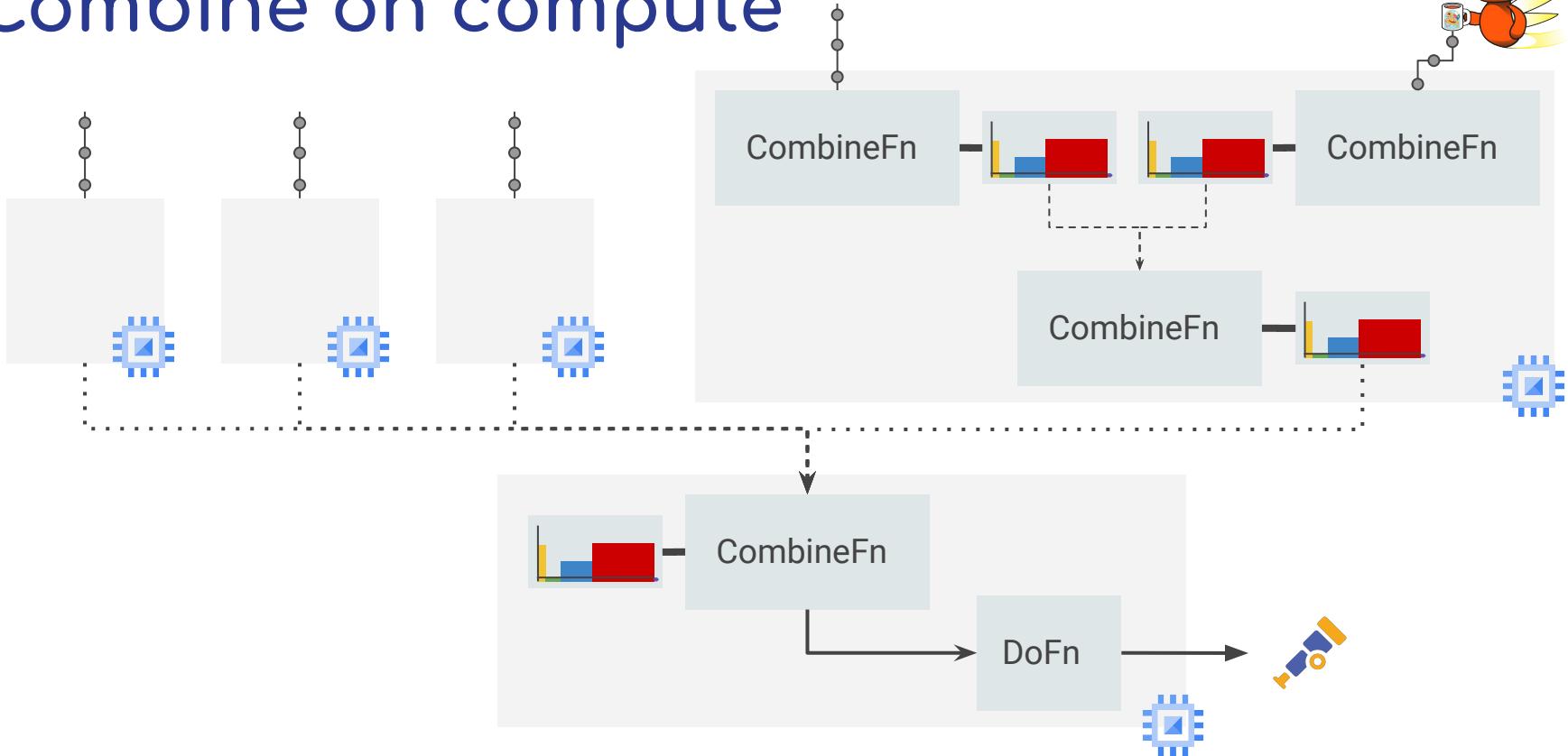
# Combine on compute



# Combine on compute



# Combine on compute



# Conclusion



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# Next...

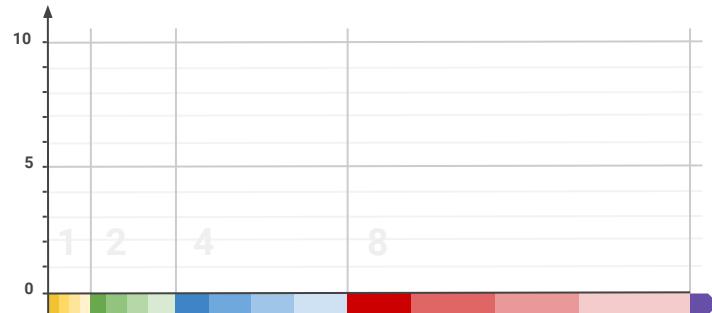
- Redo it in Go (as it's now in GA)
- Full implementation
- Examplar support
- PromQL Beam (it's just an idea)

# Questions?

Maybe some contact info  
here? Twitter @alexvb  
GitHub alexvanboxel  
Or whatever you want



# Scale





# Scale

scale

