TensorFlow

# tf.contrib.distributions.BetaWithSoftplusConcentration

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# Class BetaWithSoftplusConcentration

Inherits From: Beta

Defined in tensorflow/python/ops/distributions/beta.py.

Beta with softplus transform of concentration1 and concentration0.

# **Properties**

### allow\_nan\_stats

Python bool describing behavior when a stat is undefined.

Stats return +/- infinity when it makes sense. E.g., the variance of a Cauchy distribution is infinity. However, sometimes the statistic is undefined, e.g., if a distribution's pdf does not achieve a maximum within the support of the distribution, the mode is undefined. If the mean is undefined, then by definition the variance is undefined. E.g. the mean for Student's T for df = 1 is undefined (no clear way to say it is either + or - infinity), so the variance = E[(X - mean)\*\*2] is also undefined.

#### Returns:

• allow\_nan\_stats: Python bool.

### batch\_shape

Shape of a single sample from a single event index as a TensorShape.

May be partially defined or unknown.

The batch dimensions are indexes into independent, non-identical parameterizations of this distribution.

#### Returns:

batch\_shape: TensorShape, possibly unknown.

### concentration0

Concentration parameter associated with a 0 outcome.

#### concentration1

Concentration parameter associated with a 1 outcome.

### dtype

The DType of Tensor's handled by this Distribution.

### event\_shape

Shape of a single sample from a single batch as a TensorShape.

May be partially defined or unknown.

#### Returns:

• event\_shape: TensorShape, possibly unknown.

#### name

Name prepended to all ops created by this **Distribution**.

#### parameters

Dictionary of parameters used to instantiate this **Distribution**.

#### reparameterization\_type

Describes how samples from the distribution are reparameterized.

Currently this is one of the static instances  ${f distributions.FULLY\_REPARAMETERIZED}$  or  ${f distributions.NOT\_REPARAMETERIZED}$  .

#### Returns:

An instance of ReparameterizationType.

### total\_concentration

Sum of concentration parameters.

### validate\_args

Python **bool** indicating possibly expensive checks are enabled.

### Methods

```
__init__(
   concentration1,
   concentration0,
   validate_args=False,
   allow_nan_stats=True,
   name='BetaWithSoftplusConcentration'
```

### batch\_shape\_tensor

```
batch_shape_tensor(name='batch_shape_tensor')
```

Shape of a single sample from a single event index as a 1-D Tensor.

The batch dimensions are indexes into independent, non-identical parameterizations of this distribution.

#### Args:

• name: name to give to the op

#### Returns:

• batch\_shape: Tensor.

### cdf

```
cdf(
    value,
    name='cdf'
```

Cumulative distribution function.

Given random variable X, the cumulative distribution function cdf is:

```
cdf(x) := P[X \le x]
```

Additional documentation from Beta:



Note: x must have dtype self.dtype and be in [0, 1]. It must have a shape compatible with self.batch\_shape().

## Args:

- value: float or double Tensor.
- name: The name to give this op.

#### Returns:

cdf: a Tensor of shape sample\_shape(x) + self.batch\_shape with values of type self.dtype.

#### copy

```
copy(**override_parameters_kwargs)
```

Creates a deep copy of the distribution.



Note: the copy distribution may continue to depend on the original initialization arguments.

#### Args:

\*\*override\_parameters\_kwargs: String/value dictionary of initialization arguments to override with new values.

#### Returns:

distribution: A new instance of type(self) initialized from the union of self.parameters and override\_parameters\_kwargs, i.e., dict(self.parameters, \*\*override\_parameters\_kwargs).

#### covariance

```
covariance(name='covariance')
```

#### Covariance.

Covariance is (possibly) defined only for non-scalar-event distributions.

For example, for a length-k, vector-valued distribution, it is calculated as,

```
Cov[i, j] = Covariance(X_i, X_j) = E[(X_i - E[X_i]) (X_j - E[X_j])]
```

where Cov is a (batch of)  $k \times k$  matrix,  $0 \leftarrow (i, j) \leftarrow k$ , and E denotes expectation.

Alternatively, for non-vector, multivariate distributions (e.g., matrix-valued, Wishart), **Covariance** shall return a (batch of) matrices under some vectorization of the events, i.e.,

```
Cov[i, j] = Covariance(Vec(X)_i, Vec(X)_j) = [as above]
```

where Cov is a (batch of)  $k' \times k'$  matrices,  $0 \le (i, j) \le k' = reduce\_prod(event\_shape)$ , and Vec is some function mapping indices of this distribution's event dimensions to indices of a length-k' vector.

#### Args:

name: The name to give this op.

#### Returns:

• covariance: Floating-point **Tensor** with shape **[B1, ..., Bn, k', k']** where the first **n** dimensions are batch coordinates and **k' = reduce\_prod(self.event\_shape)**.

### entropy

```
entropy(name='entropy')
```

Shannon entropy in nats.

### event\_shape\_tensor

```
event_shape_tensor(name='event_shape_tensor')
```

Shape of a single sample from a single batch as a 1-D int32 Tensor.

### Args:

• name: name to give to the op

#### Returns:

• event\_shape: Tensor.

### is\_scalar\_batch

```
is_scalar_batch(name='is_scalar_batch')
```

Indicates that batch\_shape == [].

### Args:

• name: The name to give this op.

#### Returns:

• is\_scalar\_batch: bool scalar Tensor.

### is\_scalar\_event

```
is_scalar_event(name='is_scalar_event')
```

Indicates that **event\_shape == []**.

### Args:

• name: The name to give this op.

#### Returns:

• is\_scalar\_event: bool scalar Tensor.

### log\_cdf

```
log_cdf(
    value,
    name='log_cdf'
)
```

Log cumulative distribution function.

Given random variable X, the cumulative distribution function cdf is:

```
log\_cdf(x) := Log[P[X \le x]]
```

Often, a numerical approximation can be used for  $log_cdf(x)$  that yields a more accurate answer than simply taking the logarithm of the cdf when x << -1.

Additional documentation from Beta:



Note: x must have dtype self.dtype and be in [0, 1]. It must have a shape compatible with self.batch\_shape().

#### Args:

- value: float or double Tensor.
- name: The name to give this op.

### Returns:

• logcdf: a Tensor of shape sample\_shape(x) + self.batch\_shape with values of type self.dtype.

### log\_prob

```
log_prob(
    value,
    name='log_prob'
)
```

Log probability density/mass function.

Additional documentation from Beta:



Note: x must have dtype self.dtype and be in [0, 1]. It must have a shape compatible with self.batch\_shape().

### Args:

- value: float or double Tensor.
- name: The name to give this op.

#### Returns:

log\_prob: a Tensor of shape sample\_shape(x) + self.batch\_shape with values of type self.dtype.

#### log\_survival\_function

```
log_survival_function(
    value,
    name='log_survival_function'
)
```

Log survival function.

Given random variable X, the survival function is defined:

```
log\_survival\_function(x) = Log[ P[X > x] ]
= Log[ 1 - P[X <= x] ]
= Log[ 1 - cdf(x) ]
```

Typically, different numerical approximations can be used for the log survival function, which are more accurate than 1 - cdf(x) when x >> 1.

### Args:

- value: float or double Tensor.
- name: The name to give this op.

#### Returns:

Tensor of shape sample\_shape(x) + self.batch\_shape with values of type self.dtype.

#### mean

```
mean(name='mean')
```

Mean.

#### mode

```
mode(name='mode')
```

Mode.

Additional documentation from Beta:



Note: The mode is undefined when concentration1 <= 1 or concentration0 <= 1. If self.allow\_nan\_stats is True,
NaN is used for undefined modes. If self.allow\_nan\_stats is False an exception is raised when one or more modes are
undefined.

### param\_shapes

```
param_shapes(
    cls,
    sample_shape,
    name='DistributionParamShapes'
)
```

Shapes of parameters given the desired shape of a call to sample().

This is a class method that describes what key/value arguments are required to instantiate the given **Distribution** so that a particular shape is returned for that instance's call to **sample()**.

Subclasses should override class method \_param\_shapes .

### Args:

• sample\_shape: Tensor or python list/tuple. Desired shape of a call to sample().

• name: name to prepend ops with.

#### Returns:

dict of parameter name to Tensor shapes.

#### param\_static\_shapes

```
param_static_shapes(
    cls,
    sample_shape
)
```

param\_shapes with static (i.e. TensorShape ) shapes.

This is a class method that describes what key/value arguments are required to instantiate the given **Distribution** so that a particular shape is returned for that instance's call to **sample()**. Assumes that the sample's shape is known statically.

Subclasses should override class method \_param\_shapes to return constant-valued tensors when constant values are fed.

### Args:

sample\_shape: TensorShape or python list/tuple. Desired shape of a call to sample().

#### Returns:

dict of parameter name to TensorShape.

### Raises:

• ValueError: if sample\_shape is a TensorShape and is not fully defined.

### prob

```
prob(
   value,
   name='prob'
)
```

Probability density/mass function.

Additional documentation from Beta:



Note: x must have dtype self.dtype and be in [0, 1]. It must have a shape compatible with self.batch\_shape().

#### Args:

- value: float or double Tensor.
- name: The name to give this op.

#### Returns:

• prob: a Tensor of shape sample\_shape(x) + self.batch\_shape with values of type self.dtype.

### quantile

```
quantile(
   value,
   name='quantile'
)
```

Quantile function. Aka "inverse cdf" or "percent point function".

Given random variable X and p in [0, 1], the quantile is:

```
quantile(p) := x such that P[X \le x] == p
```

### Args:

- value: float or double Tensor.
- name: The name to give this op.

#### Returns:

• quantile: a Tensor of shape sample\_shape(x) + self.batch\_shape with values of type self.dtype.

### sample

```
sample(
    sample_shape=(),
    seed=None,
    name='sample'
)
```

Generate samples of the specified shape.

Note that a call to sample() without arguments will generate a single sample.

#### Args:

- sample\_shape: 0D or 1D int32 Tensor. Shape of the generated samples.
- · seed: Python integer seed for RNG
- name: name to give to the op.

### Returns:

• samples: a Tensor with prepended dimensions sample\_shape.

### stddev

```
stddev(name='stddev')
```

Standard deviation.

Standard deviation is defined as,

```
stddev = E[(X - E[X])**2]**0.5
```

where X is the random variable associated with this distribution, E denotes expectation, and stddev.shape = batch\_shape + event\_shape .

### Args:

• name: The name to give this op.

#### Returns:

stddev: Floating-point Tensor with shape identical to batch\_shape + event\_shape , i.e., the same shape as self.mean().

### survival\_function

```
survival_function(
   value,
   name='survival_function'
)
```

Survival function.

Given random variable X, the survival function is defined:

```
survival\_function(x) = P[X > x]
= 1 - P[X \le x]
= 1 - cdf(x).
```

### Args:

- value: float or double Tensor.
- name: The name to give this op.

#### Returns:

Tensor of shape  $sample_shape(x) + self.batch_shape$  with values of type self.dtype.

### variance

```
variance(name='variance')
```

Variance.

Variance is defined as,

```
Var = E[(X - E[X])**2]
```

where X is the random variable associated with this distribution, E denotes expectation, and Var.shape = batch\_shape + event\_shape .

### Args:

• name: The name to give this op.

#### Returns:

• variance: Floating-point **Tensor** with shape identical to **batch\_shape + event\_shape**, i.e., the same shape as **self.mean()**.

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