

tf.contrib.distributions.bijectors.SinhArcsinh

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Inherits From: [Bijector](#)Defined in [tensorflow/contrib/distributions/python/ops/bijectors/sinh_arcsinh_impl.py](#).Compute $Y = g(X) = \text{Sinh}(\text{Arcsinh}(X) + \text{skewness} * \text{tailweight})$.

For skewness in $(-\infty, \infty)$ and tailweight in $(0, \infty)$, this transformation is a diffeomorphism of the real line $(-\infty, \infty)$. The inverse transform is $X = g^{-1}(Y) = \text{Sinh}(\text{ArcSinh}(Y) / \text{tailweight} - \text{skewness})$.

The **SinhArcsinh** transformation of the Normal is described in [Sinh-arcsinh distributions](#). This Bijector allows a similar transformation of any distribution supported on $(-\infty, \infty)$.

Meaning of the parameters

- If $\text{skewness} = 0$ and $\text{tailweight} = 1$, this transform is the identity.
- Positive (negative) skewness leads to positive (negative) skew.
- positive skew means, for unimodal X centered at zero, the mode of Y is "tilted" to the right.
- positive skew means positive values of Y become more likely, and negative values become less likely.
- Larger (smaller) tailweight leads to fatter (thinner) tails.
- Fatter tails mean larger values of $|Y|$ become more likely.
- If X is a unit Normal, $\text{tailweight} < 1$ leads to a distribution that is "flat" around $Y = 0$, and a very steep drop-off in the tails.
- If X is a unit Normal, $\text{tailweight} > 1$ leads to a distribution more peaked at the mode with heavier tails.

To see the argument about the tails, note that for $|X| \gg 1$ and $|X| \gg (|\text{skewness}| * \text{tailweight})^{1/\text{tailweight}}$, we have $Y \approx 0.5 X^{2\text{tailweight}} e^{(\text{sign}(X) \text{skewness} * \text{tailweight})}$.

Properties

dtype

dtype of **Tensor**s transformable by this distribution.

event_ndims

Returns then number of event dimensions this bijector operates on.

graph_parents

Returns this `Bijector` 's graph_parents as a Python list.

is_constant_jacobian

Returns true iff the Jacobian is not a function of x.

★ **Note:** Jacobian is either constant for both forward and inverse or neither.

Returns:

- `is_constant_jacobian`: Python `bool`.

name

Returns the string name of this `Bijector`.

skewness

The `skewness` in: $Y = \text{Sinh}((\text{Arcsinh}(X) + \text{skewness}) * \text{tailweight})$.

tailweight

The `tailweight` in: $Y = \text{Sinh}((\text{Arcsinh}(X) + \text{skewness}) * \text{tailweight})$.

validate_args

Returns True if Tensor arguments will be validated.

Methods

`__init__`

```
__init__(
    skewness=0.0,
    tailweight=1.0,
    event_ndims=0,
    validate_args=False,
    name='sinh_arcsinh'
)
```

Instantiates the `SinhArcsinh` bijector.

Args:

- `skewness`: Skewness parameter. Float-type `Tensor`.

- `tailweight` : Tailweight parameter. Positive `Tensor` of same `dtype` as `skewness` and broadcastable `shape` .
- `event_ndims` : Python scalar indicating the number of dimensions associated with a particular draw from the distribution.
- `validate_args` : Python `bool` indicating whether arguments should be checked for correctness.
- `name` : Python `str` name given to ops managed by this object.

forward

```
forward(
    x,
    name='forward'
)
```

Returns the forward `Bijector` evaluation, i.e., $X = g(Y)$.

Args:

- `x` : `Tensor` . The input to the "forward" evaluation.
- `name` : The name to give this op.

Returns:

`Tensor` .

Raises:

- `TypeError` : if `self.dtype` is specified and `x.dtype` is not `self.dtype` .
- `NotImplementedError` : if `_forward` is not implemented.

forward_event_shape

```
forward_event_shape(input_shape)
```

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

Same meaning as `forward_event_shape_tensor` . May be only partially defined.

Args:

- `input_shape` : `TensorShape` indicating event-portion shape passed into `forward` function.

Returns:

- `forward_event_shape_tensor` : `TensorShape` indicating event-portion shape after applying `forward` . Possibly unknown.

forward_event_shape_tensor

```
forward_event_shape_tensor(
    input_shape,
    name='forward_event_shape_tensor'
)
```

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

Args:

- `input_shape`: `Tensor`, `int32` vector indicating event-portion shape passed into `forward` function.
- `name`: name to give to the op

Returns:

- `forward_event_shape_tensor`: `Tensor`, `int32` vector indicating event-portion shape after applying `forward`.

forward_log_det_jacobian

```
forward_log_det_jacobian(
    x,
    name='forward_log_det_jacobian'
)
```

Returns both the `forward_log_det_jacobian`.

Args:

- `x`: `Tensor`. The input to the "forward" Jacobian evaluation.
- `name`: The name to give this op.

Returns:

`Tensor`, if this bijector is injective. If not injective this is not implemented.

Raises:

- `TypeError`: if `self.dtype` is specified and `y.dtype` is not `self.dtype`.
- `NotImplementedError`: if neither `_forward_log_det_jacobian` nor `{_inverse, _inverse_log_det_jacobian}` are implemented, or this is a non-injective bijector.

inverse

```
inverse(
    y,
    name='inverse'
)
```

Returns the inverse `Bijector` evaluation, i.e., $X = g^{-1}(Y)$.

Args:

- `y` : **Tensor** . The input to the "inverse" evaluation.
- `name` : The name to give this op.

Returns:

Tensor , if this bijector is injective. If not injective, returns the k-tuple containing the unique `k` points `(x1, ..., xk)` such that $g(x_i) = y$.

Raises:

- **TypeError** : if `self.dtype` is specified and `y.dtype` is not `self.dtype` .
- **NotImplementedError** : if `_inverse` is not implemented.

inverse_event_shape

```
inverse_event_shape(output_shape)
```

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

Same meaning as `inverse_event_shape_tensor` . May be only partially defined.

Args:

- `output_shape` : **TensorShape** indicating event-portion shape passed into `inverse` function.

Returns:

- `inverse_event_shape_tensor` : **TensorShape** indicating event-portion shape after applying `inverse` . Possibly unknown.

inverse_event_shape_tensor

```
inverse_event_shape_tensor(
    output_shape,
    name='inverse_event_shape_tensor'
)
```

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

Args:

- `output_shape` : **Tensor** , **int32** vector indicating event-portion shape passed into `inverse` function.
- `name` : name to give to the op

Returns:

- `inverse_event_shape_tensor` : **Tensor** , **int32** vector indicating event-portion shape after applying `inverse` .

inverse_log_det_jacobian

```
inverse_log_det_jacobian(  
    y,  
    name='inverse_log_det_jacobian'  
)
```

Returns the $(\log \circ \det \circ \text{Jacobian} \circ \text{inverse})(y)$.

Mathematically, returns: $\log(\det(dX/dY))(Y)$. (Recall that: $X=g^{-1}(Y)$.)

Note that `forward_log_det_jacobian` is the negative of this function, evaluated at $g^{-1}(y)$.

Args:

- `y`: `Tensor`. The input to the "inverse" Jacobian evaluation.
- `name`: The name to give this op.

Returns:

`Tensor`, if this bijector is injective. If not injective, returns the tuple of local log det Jacobians, $\log(\det(Dg_i^{-1}(y)))$, where g_i is the restriction of g to the i th partition D_i .

Raises:

- `TypeError`: if `self.dtype` is specified and `y.dtype` is not `self.dtype`.
- `NotImplementedError`: if `_inverse_log_det_jacobian` is not implemented.

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