

## GaussianNoise

[\[source\]](#)

```
keras.layers.noise.GaussianNoise(sigma)
```

Apply to the input an additive zero-centered Gaussian noise with standard deviation `sigma`. This is useful to mitigate overfitting (you could see it as a kind of random data augmentation). Gaussian Noise (GS) is a natural choice as corruption process for real valued inputs.

As it is a regularization layer, it is only active at training time.

### Arguments

- **sigma**: float, standard deviation of the noise distribution.

### Input shape

Arbitrary. Use the keyword argument `input_shape` (tuple of integers, does not include the samples axis) when using this layer as the first layer in a model.

### Output shape

Same shape as input.

---

## GaussianDropout

[\[source\]](#)

```
keras.layers.noise.GaussianDropout(p)
```

Apply to the input an multiplicative one-centered Gaussian noise with standard deviation  `$\sqrt{p/(1-p)}$` .

As it is a regularization layer, it is only active at training time.

### Arguments

- **p**: float, drop probability (as with `Dropout`).

### Input shape

Arbitrary. Use the keyword argument `input_shape` (tuple of integers, does not include the samples axis) when using this layer as the first layer in a model.

## Output shape

Same shape as input.

## References

- [\\_\\_Dropout\\_\\_: A Simple Way to Prevent Neural Networks from Overfitting](#) Srivastava, Hinton, et al. 2014