

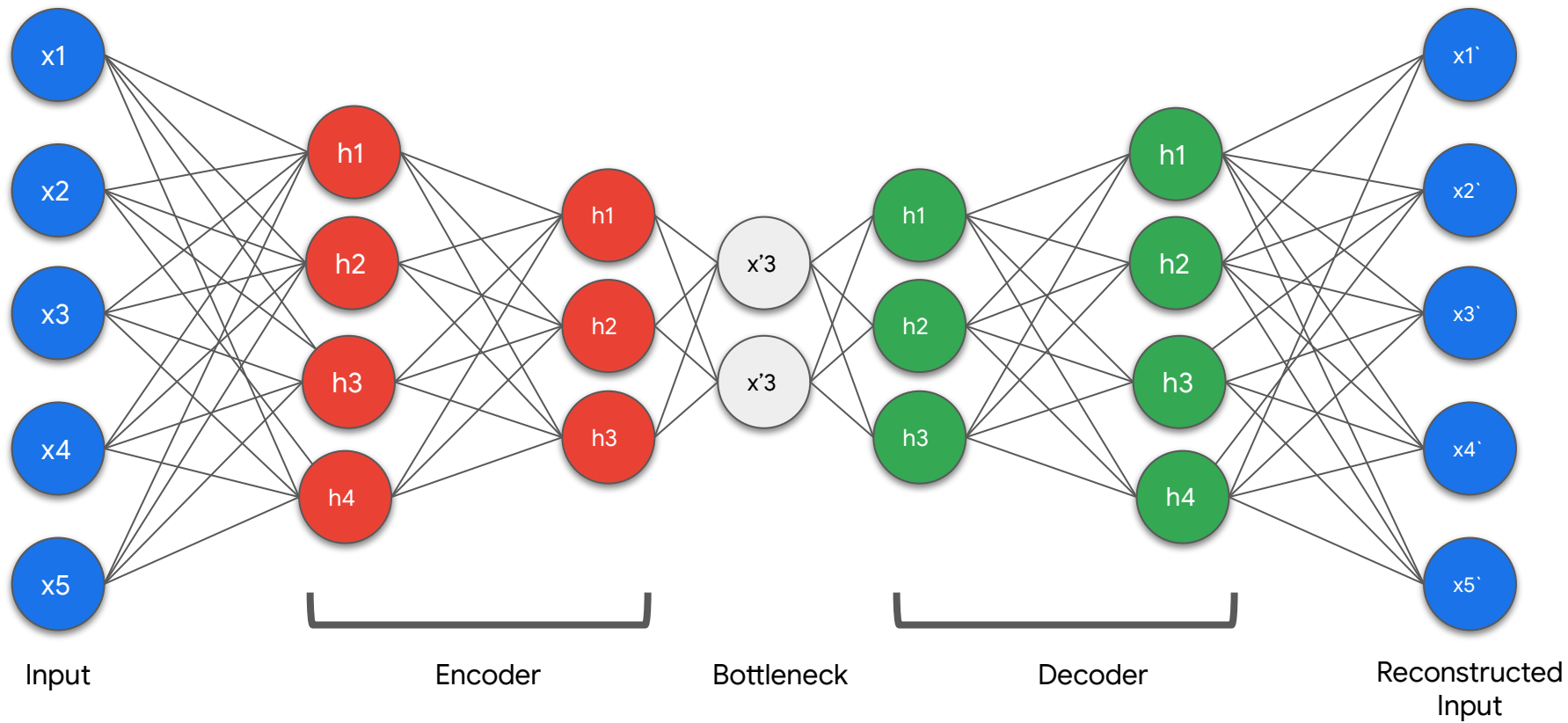
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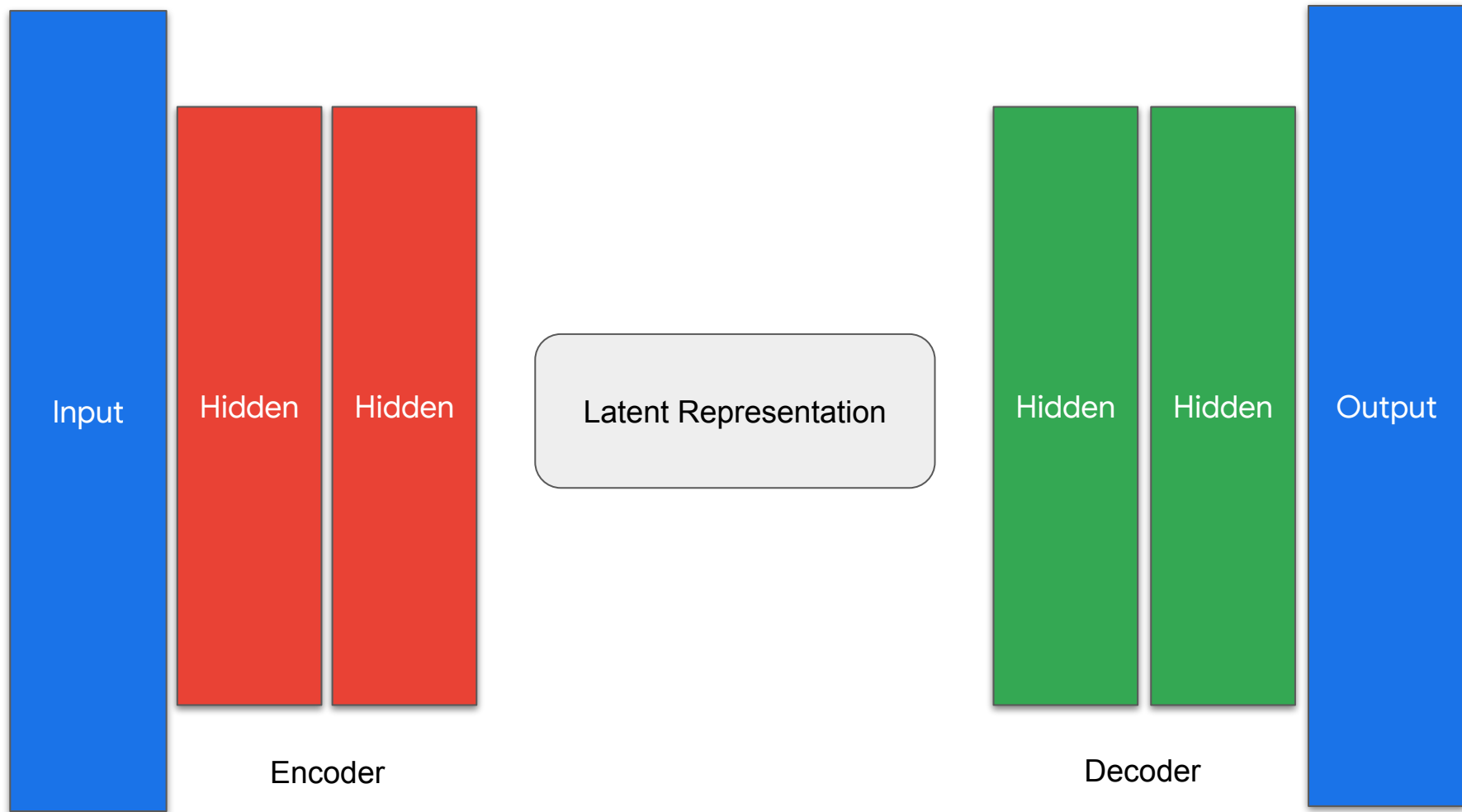
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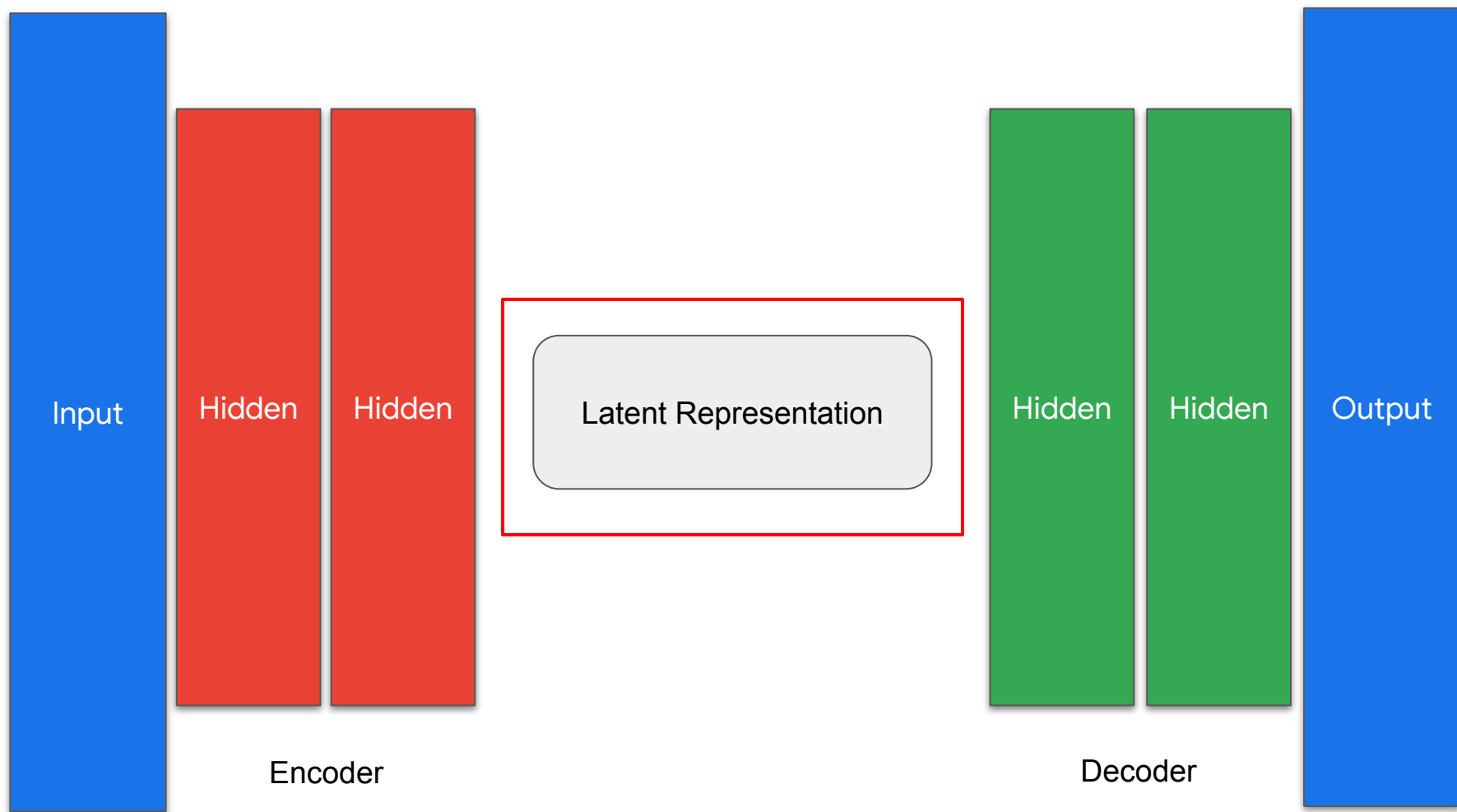
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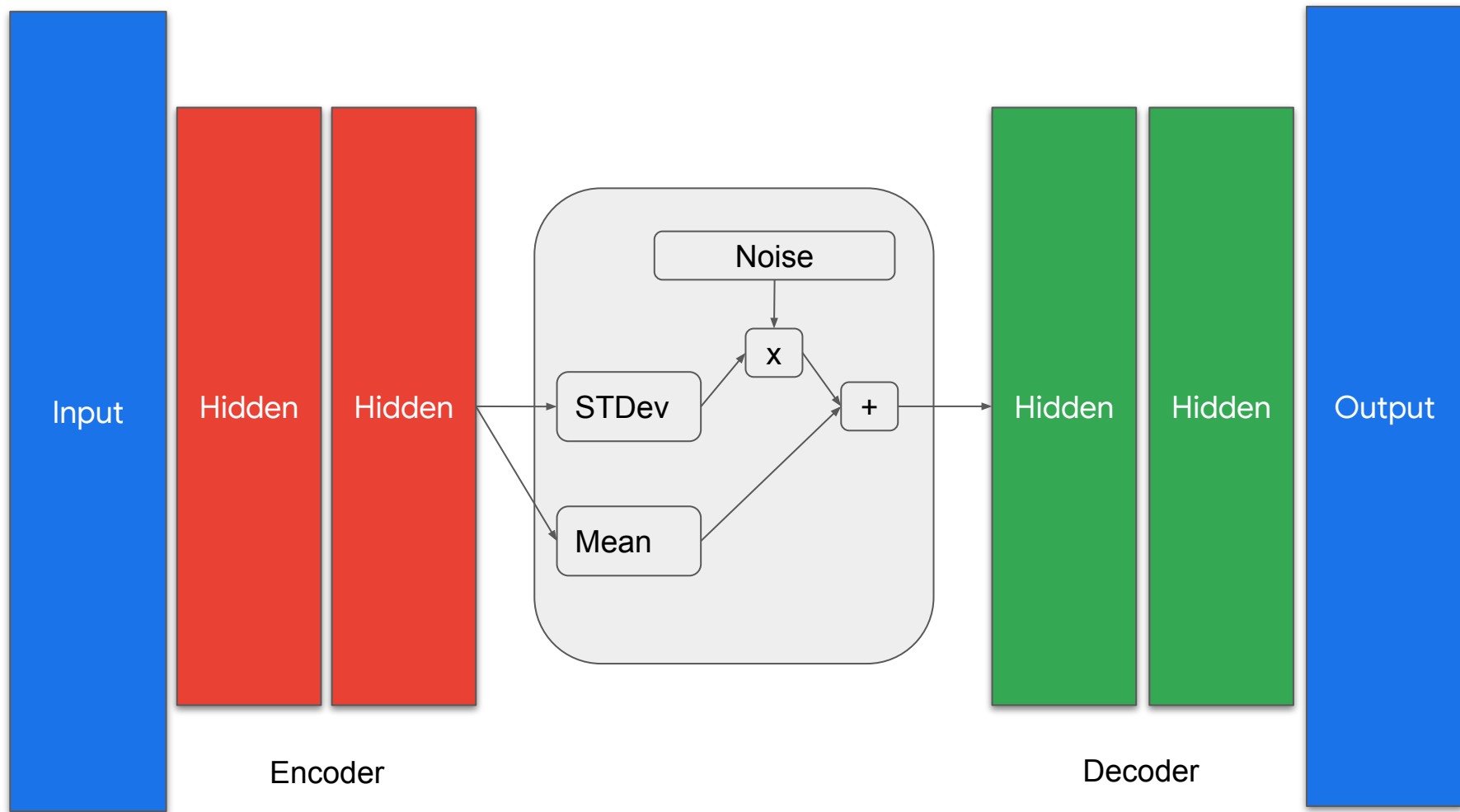
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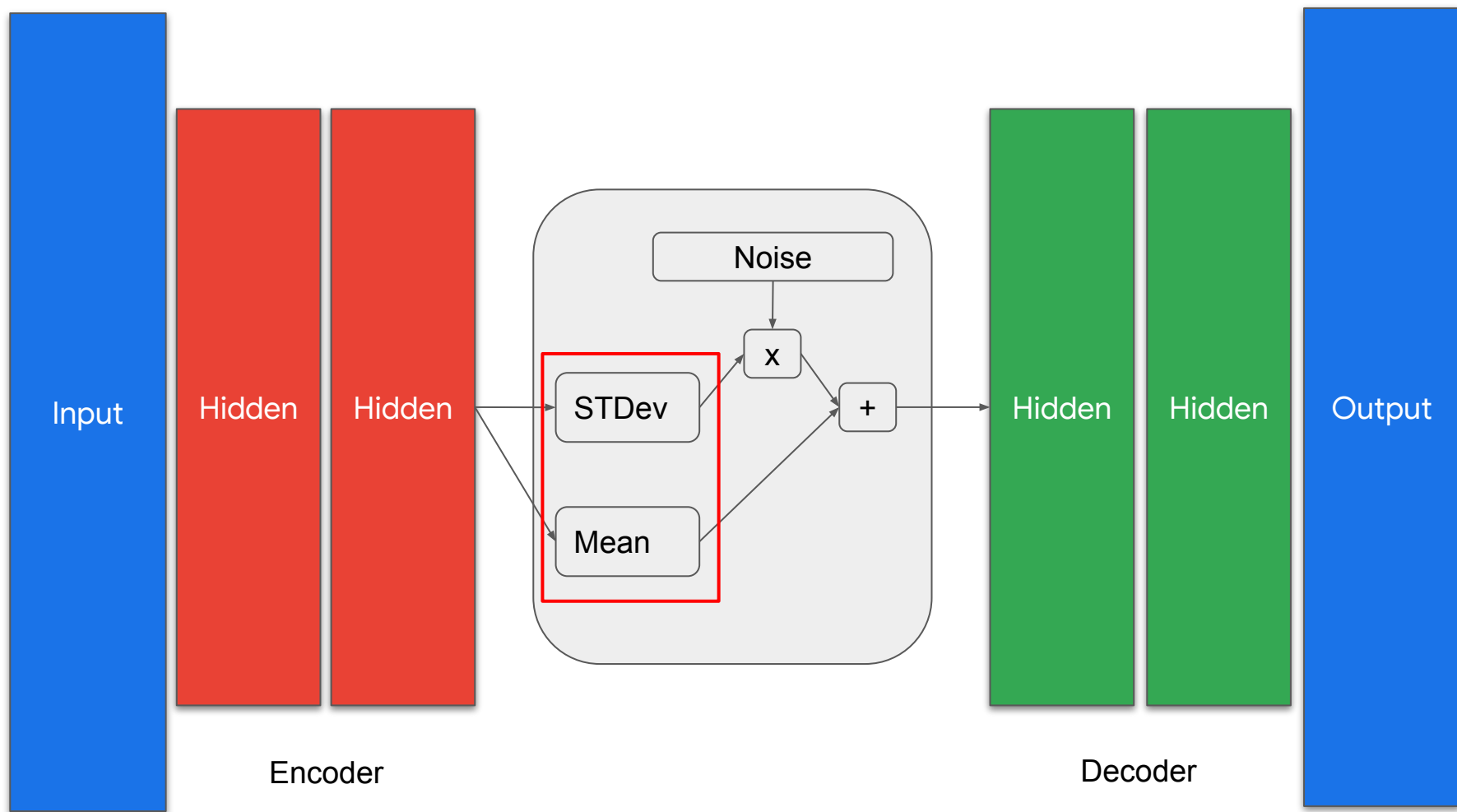
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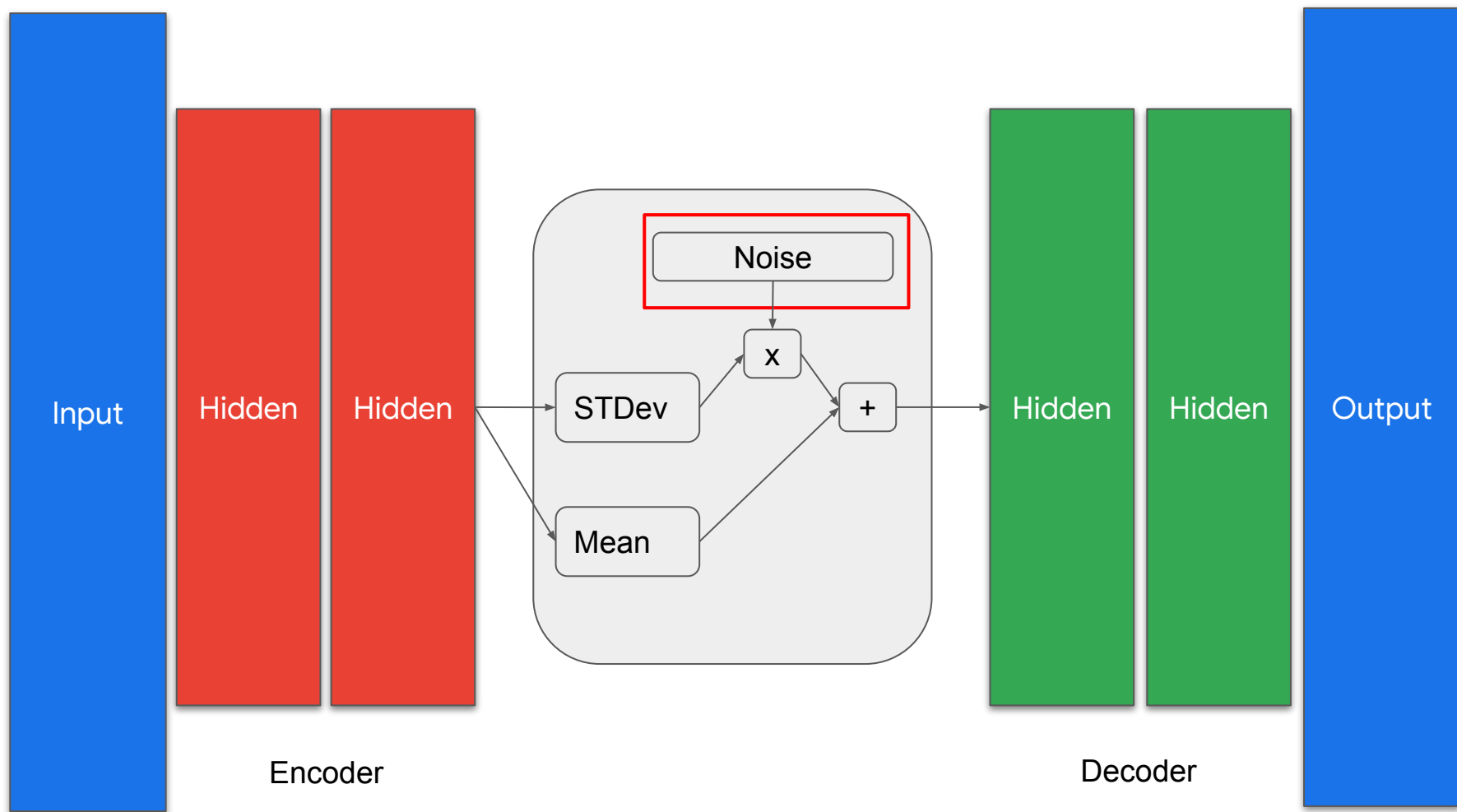


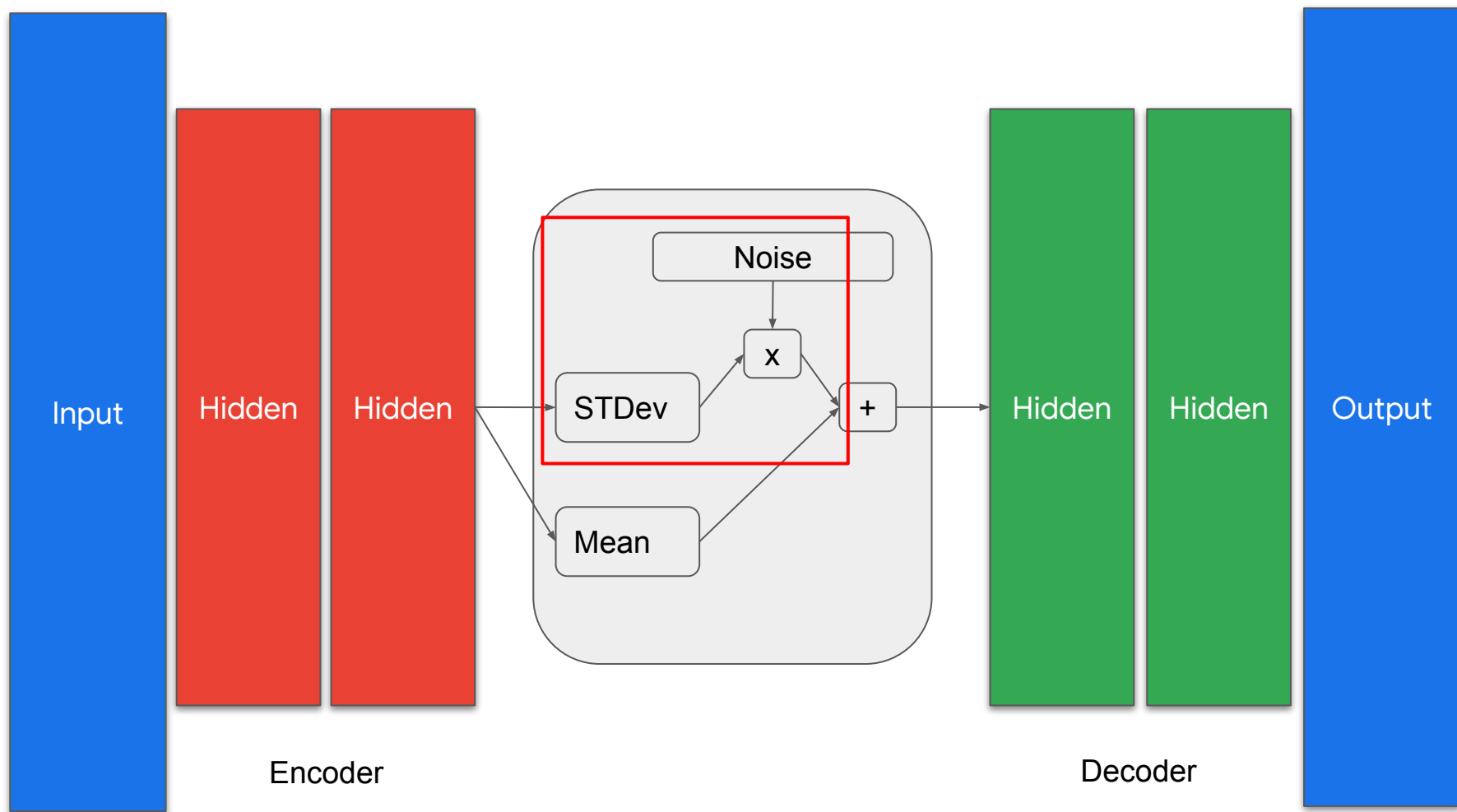


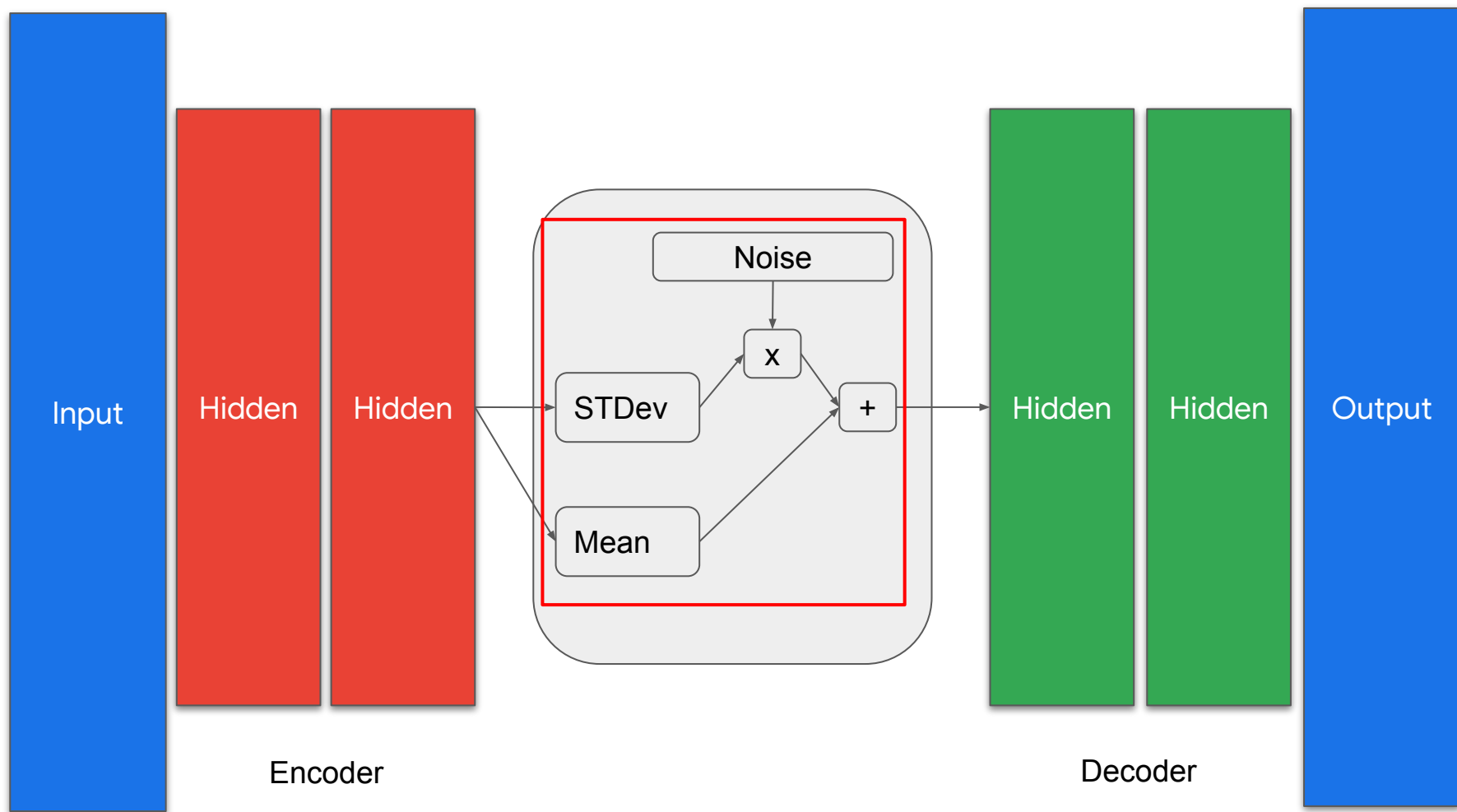


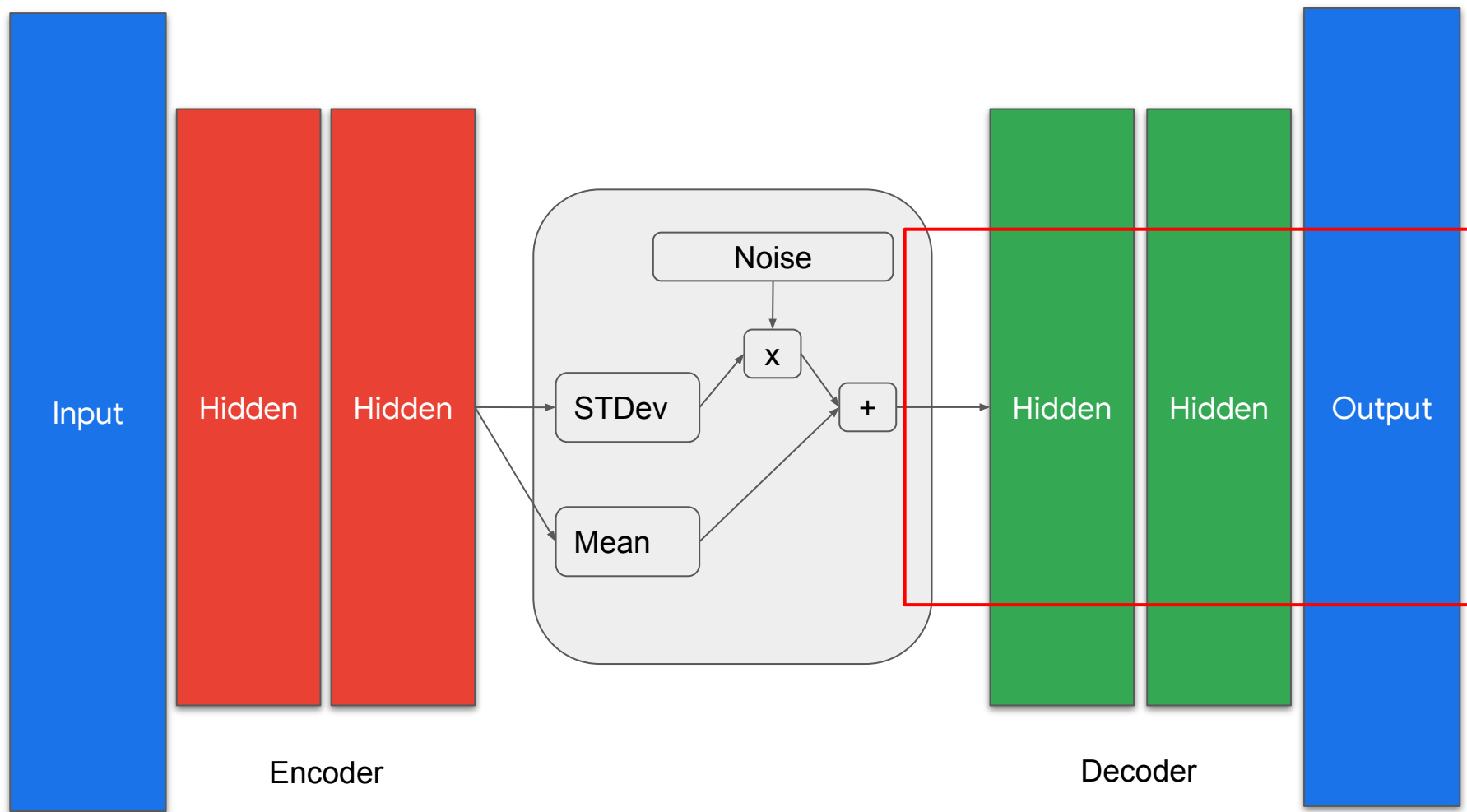


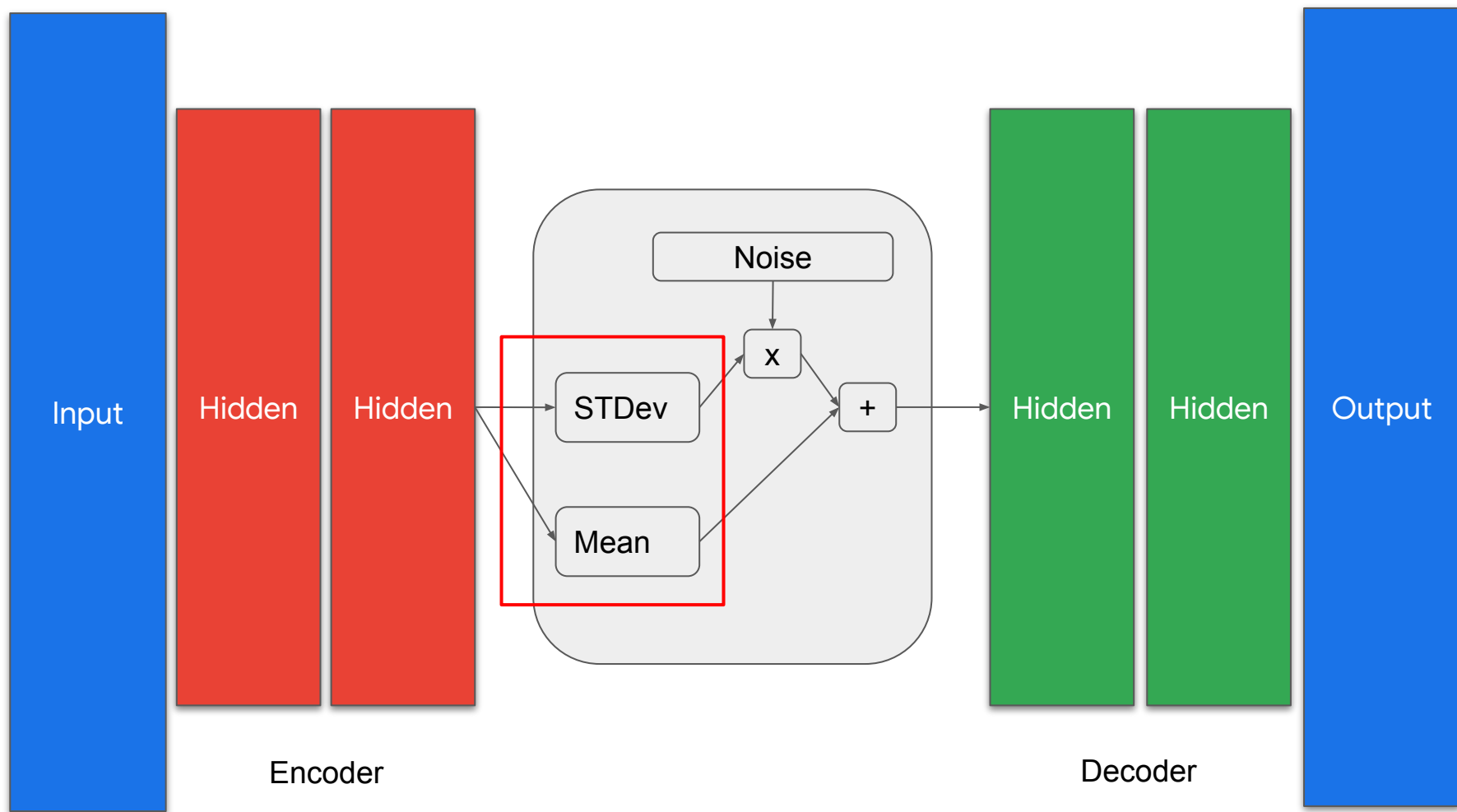


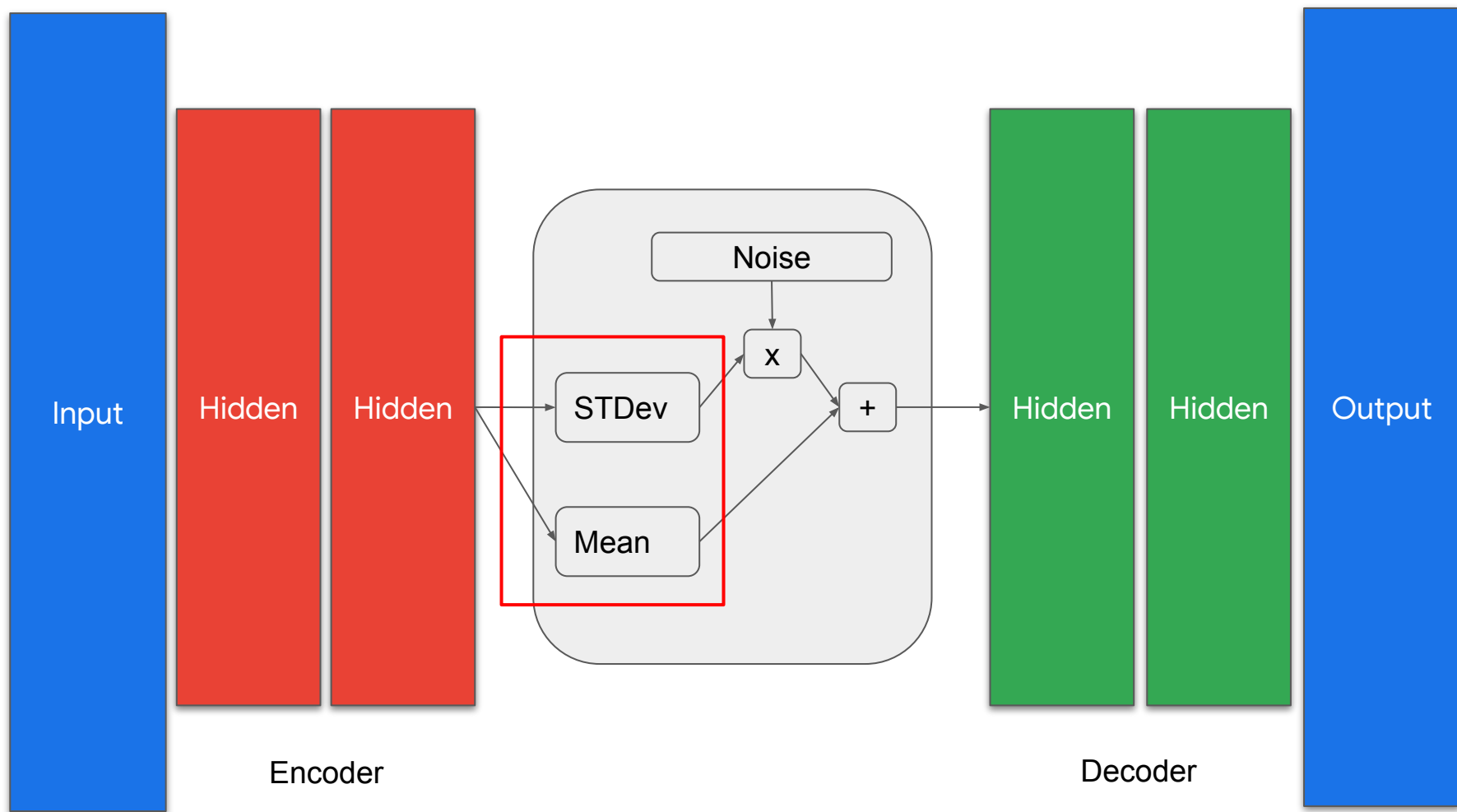


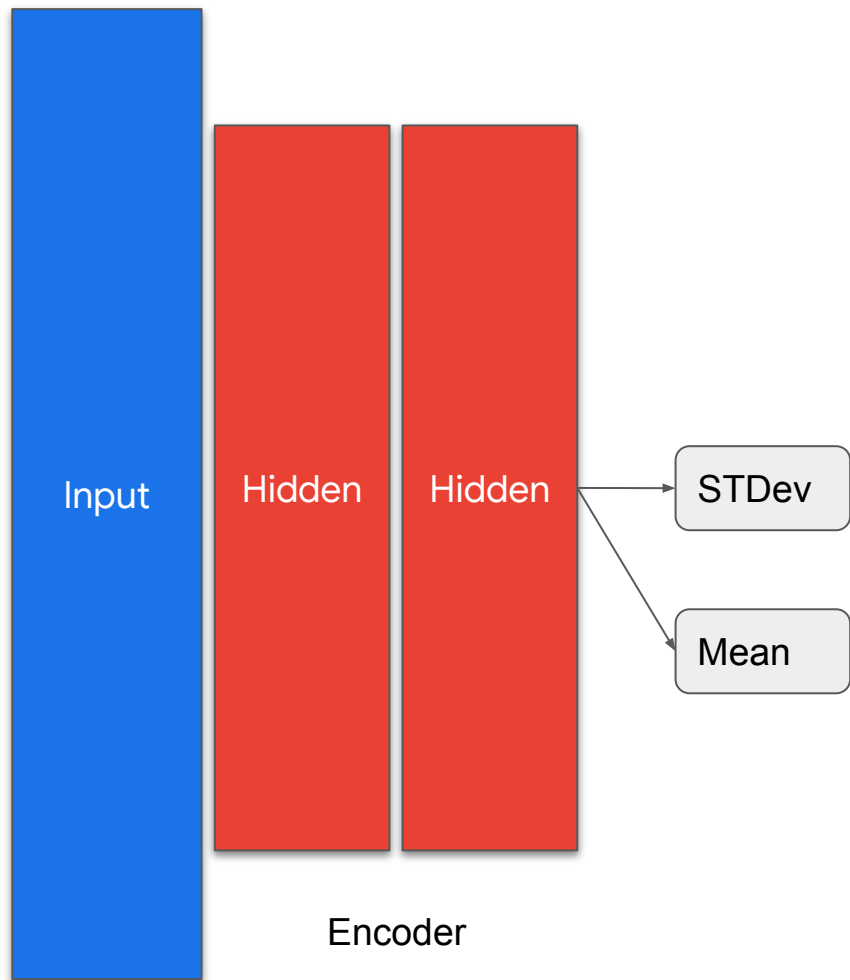


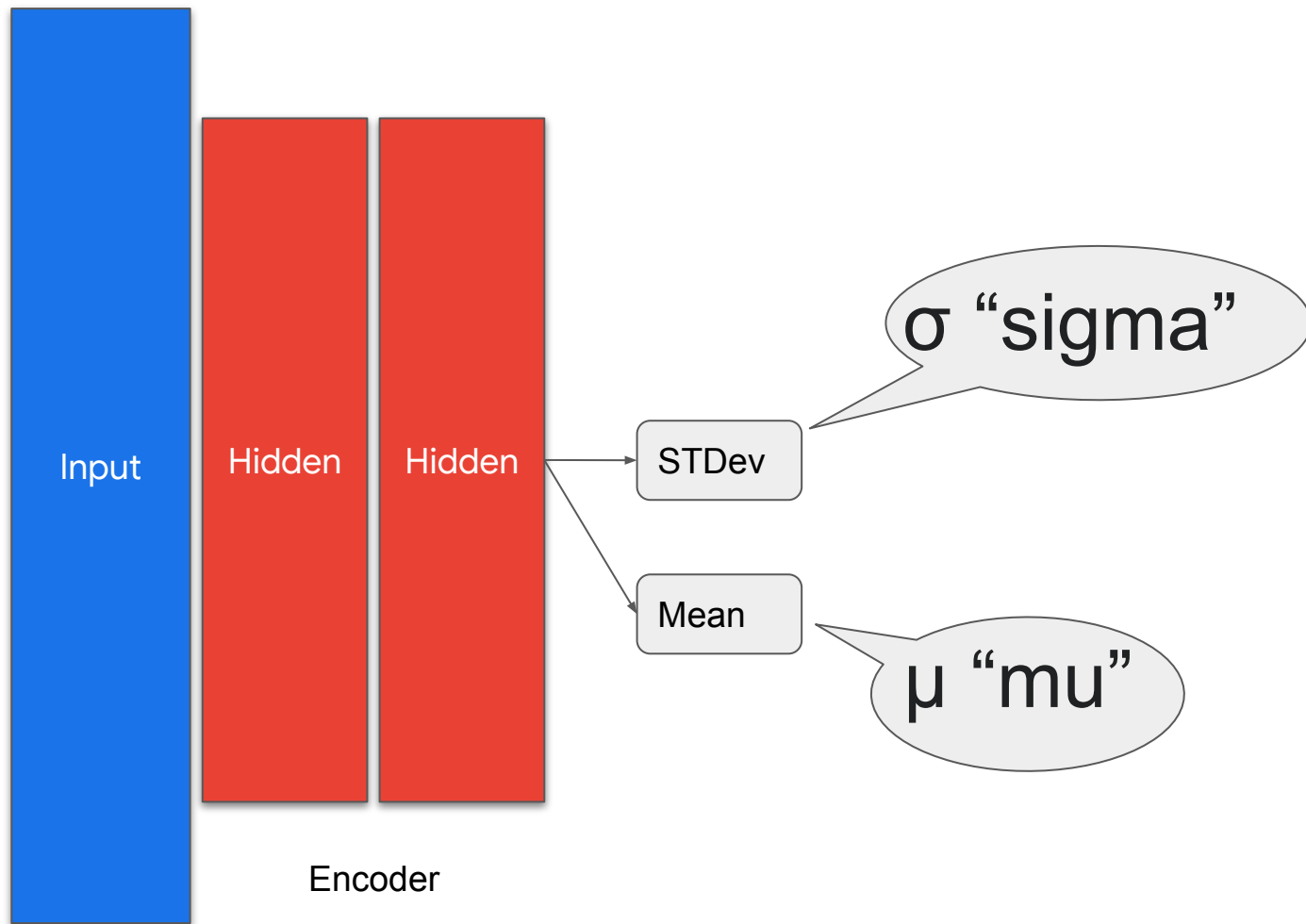












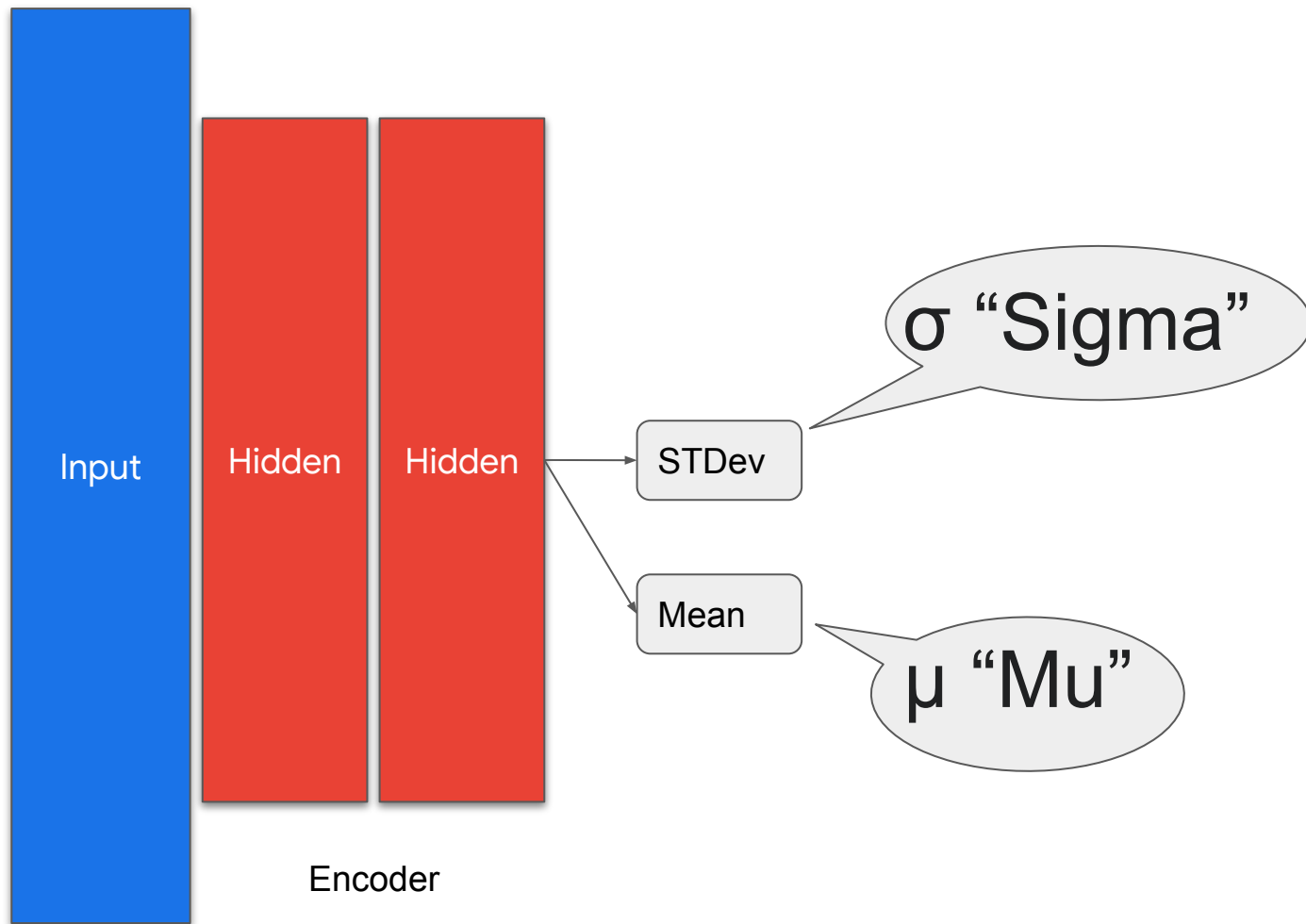
Probability Distribution

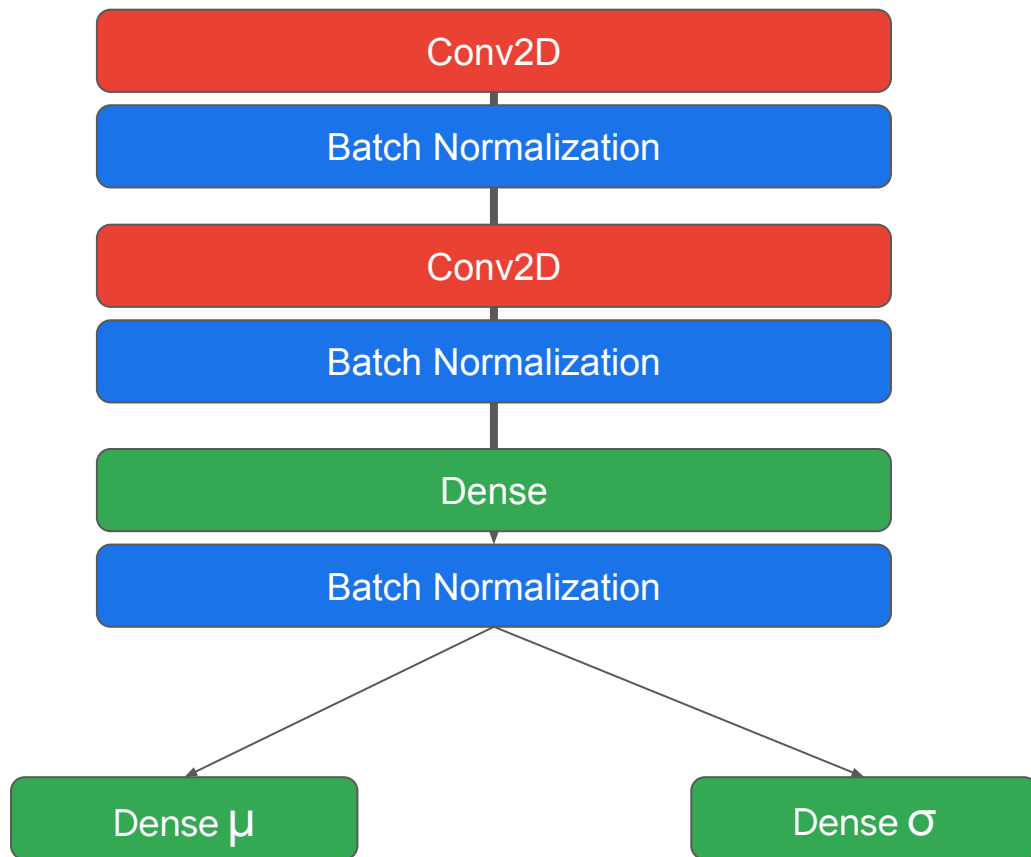
Gaussian probability density function or Normal Distribution.

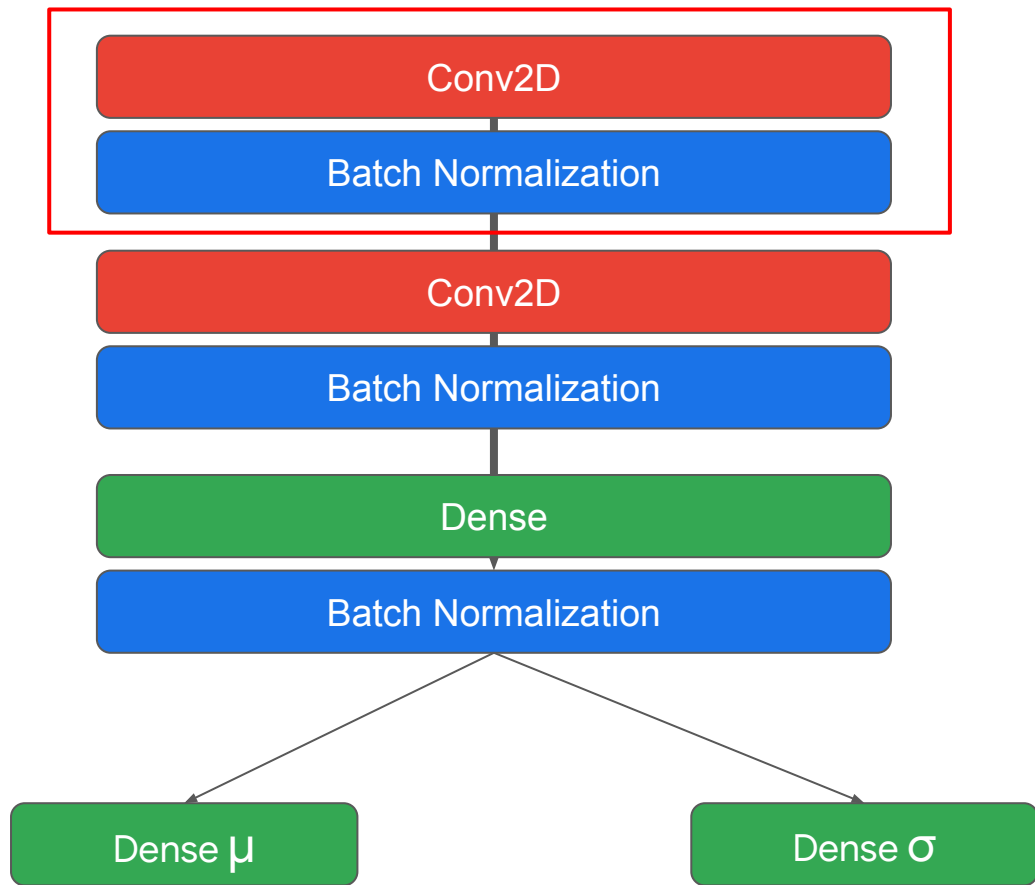
Normal Distribution is controlled by:

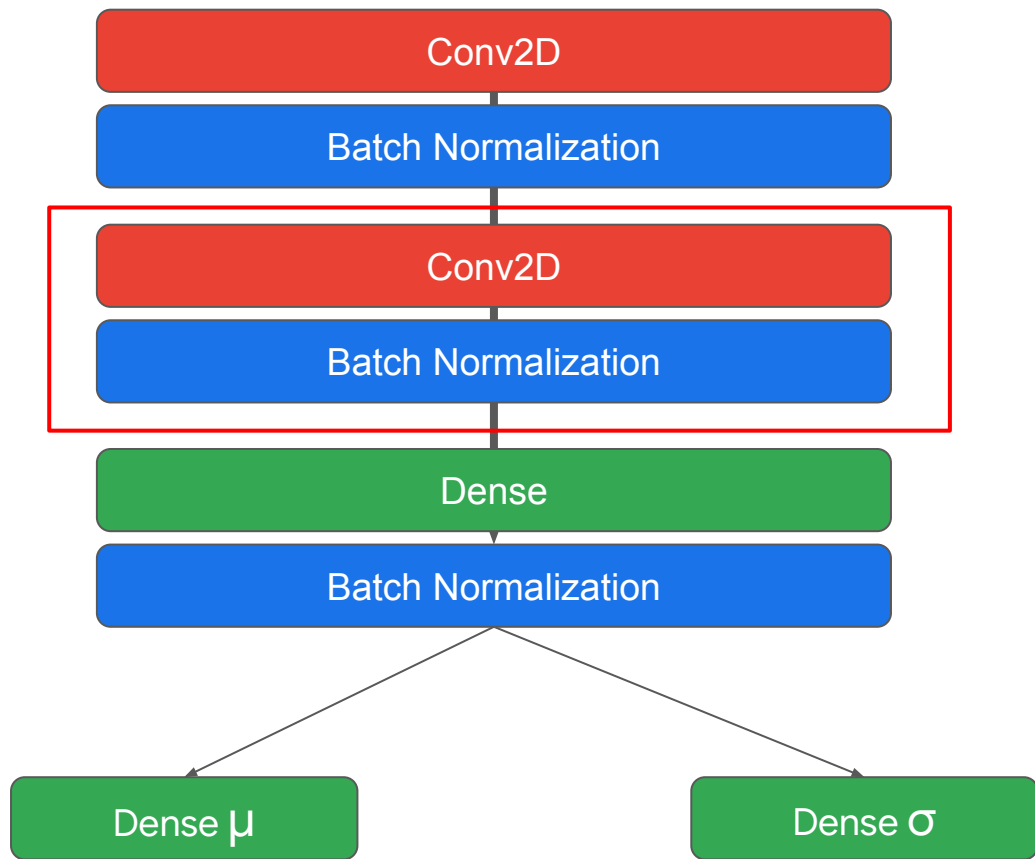
- μ “mean”
- σ “standard deviation”

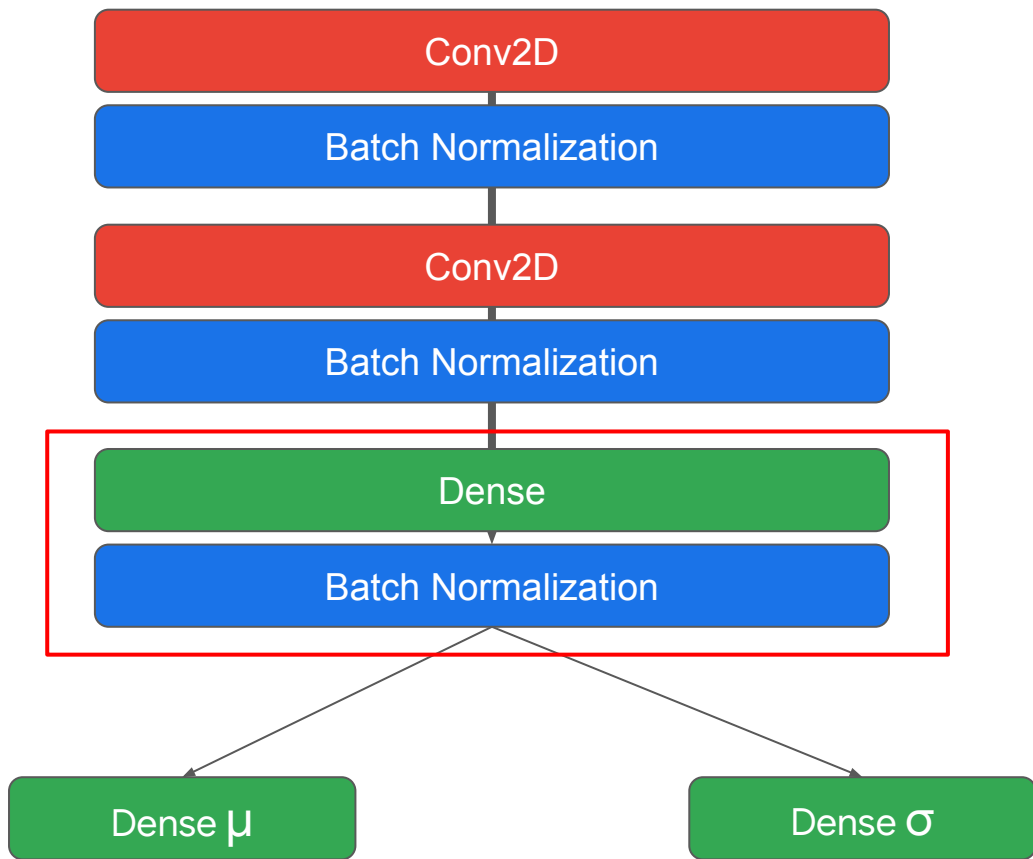
$$N(\mu, \sigma)$$

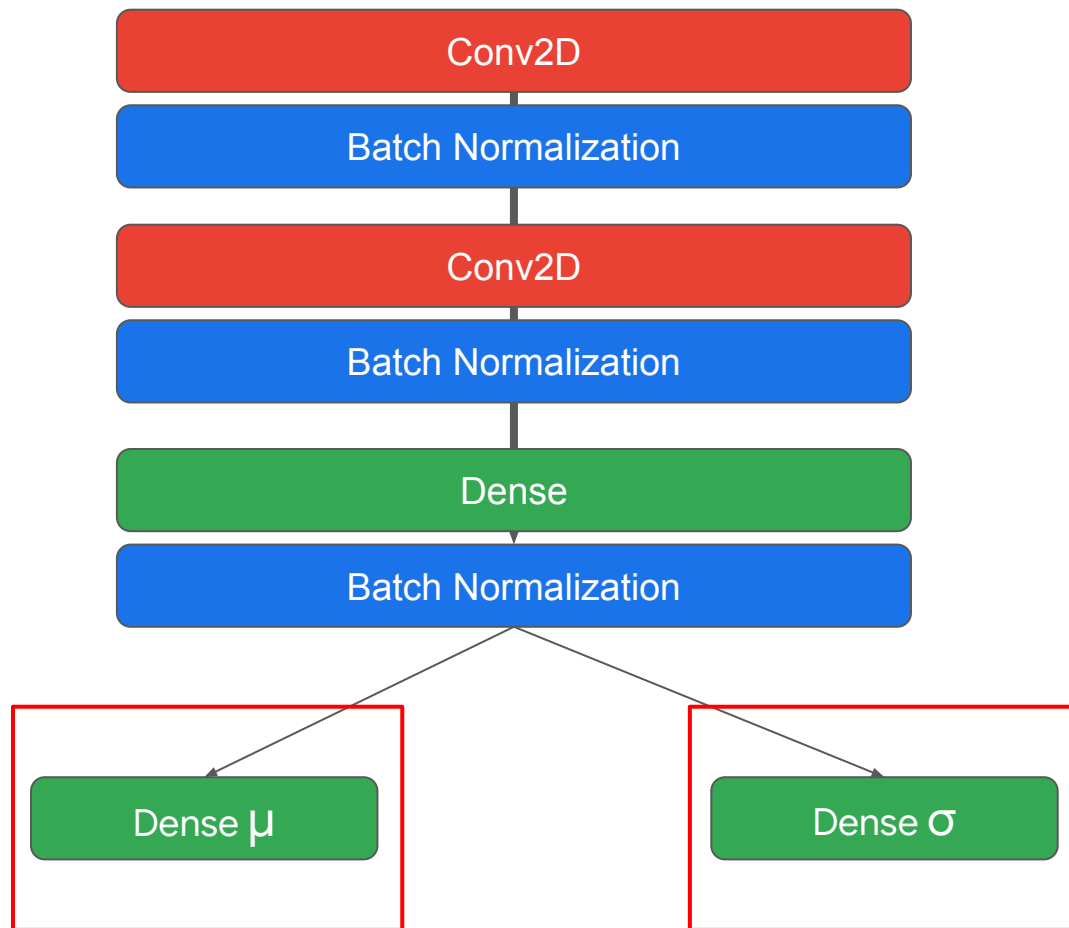


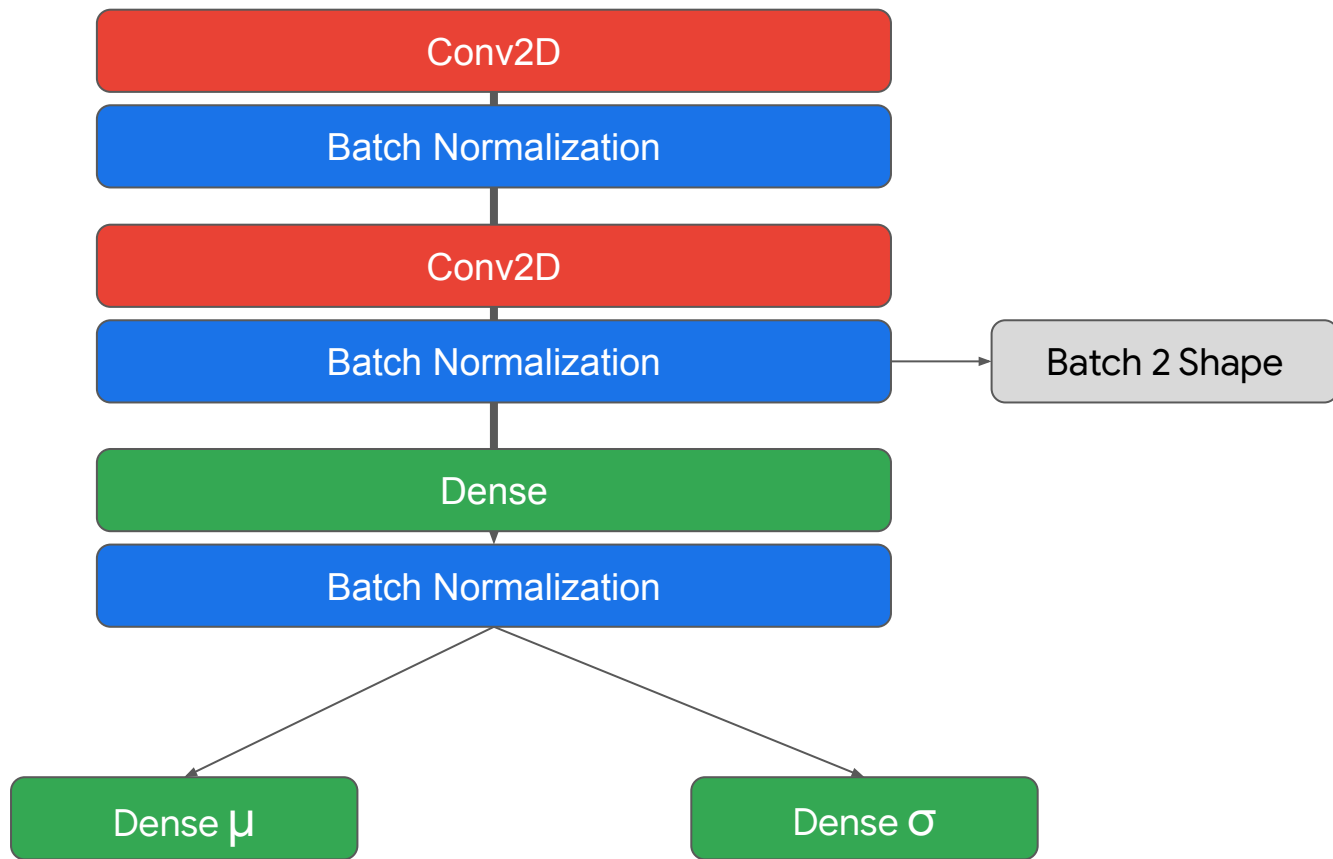












```
# This function defines the encoder's layers
def encoder_layers(inputs, latent_dim):
    x = tf.keras.layers.Conv2D(filters=32, kernel_size=3, strides=2,
                                padding="same", activation='relu',
                                name="encode_conv1")(inputs)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2D(filters=64, kernel_size=3, strides=2,
                                padding='same', activation='relu',
                                name="encode_conv2")(x)
    batch_2 = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Flatten(name="encode_flatten")(batch_2)
    x = tf.keras.layers.Dense(20, activation='relu', name="encode_dense")(x)
    x = tf.keras.layers.BatchNormalization()(x)

    mu = tf.keras.layers.Dense(latent_dim, name='latent_mu')(x)
    sigma = tf.keras.layers.Dense(latent_dim, name='latent_sigma')(x)

    return mu, sigma, batch_2.shape
```

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# This function defines the encoder's layers
```

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def encoder_layers(inputs, latent_dim):
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```
    x = tf.keras.layers.Conv2D(filters=32, kernel_size=3, strides=2,  
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    x = tf.keras.layers.Conv2D(filters=64, kernel_size=3, strides=2,  
                                padding='same', activation='relu',  
                                name="encode_conv2")(x)
```

```
    batch_2 = tf.keras.layers.BatchNormalization()(x)
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    x = tf.keras.layers.Flatten(name="encode_flatten")(batch_2)
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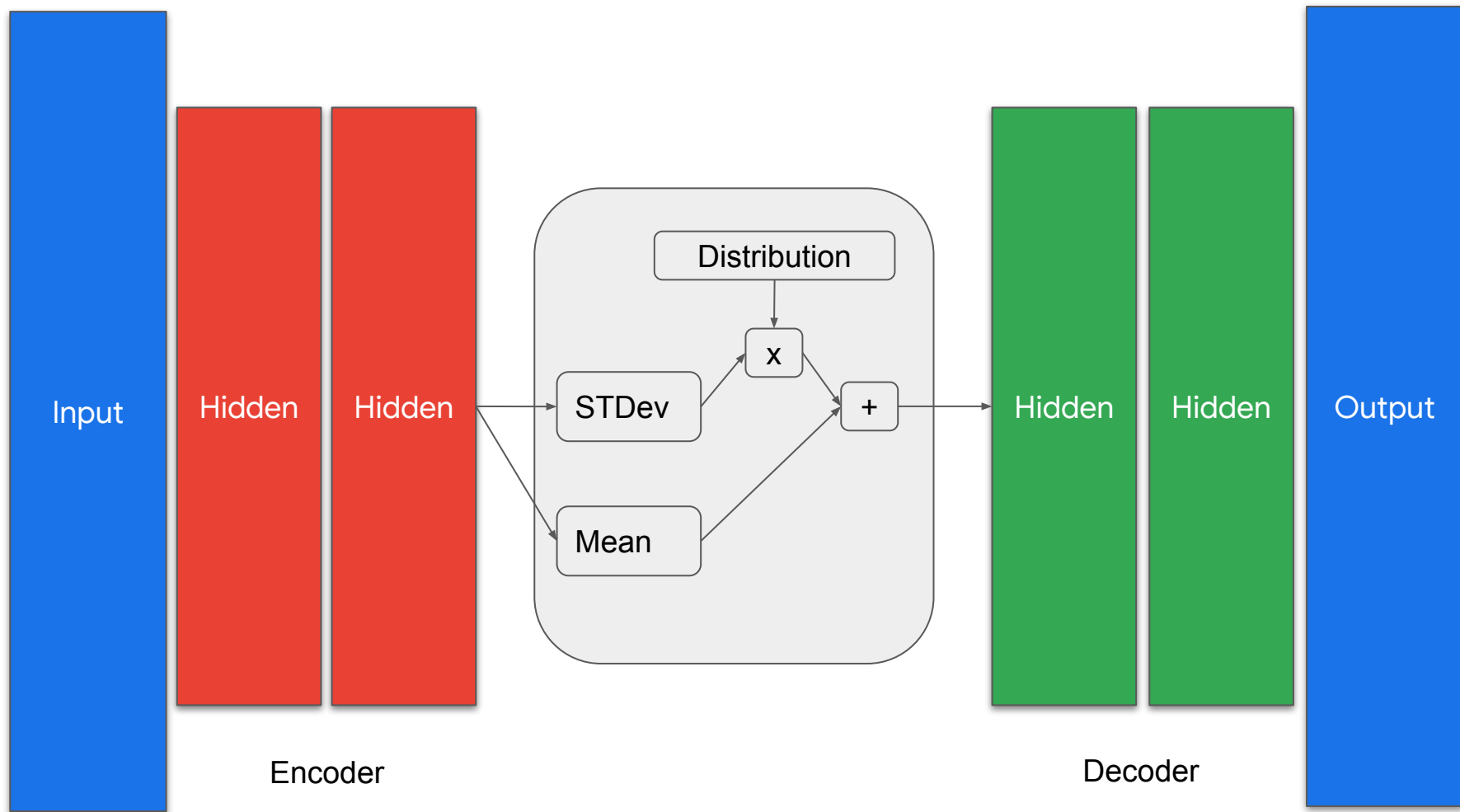
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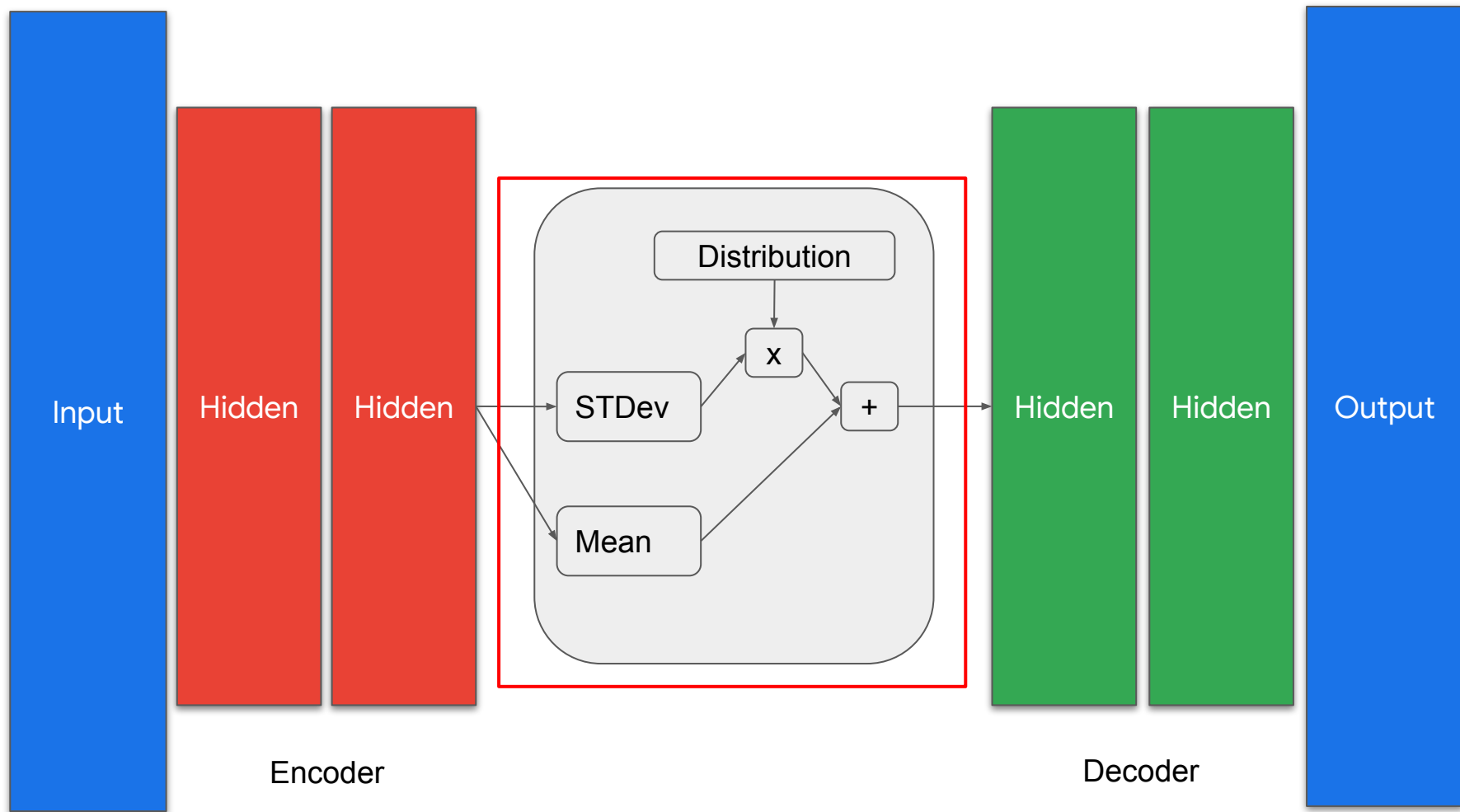
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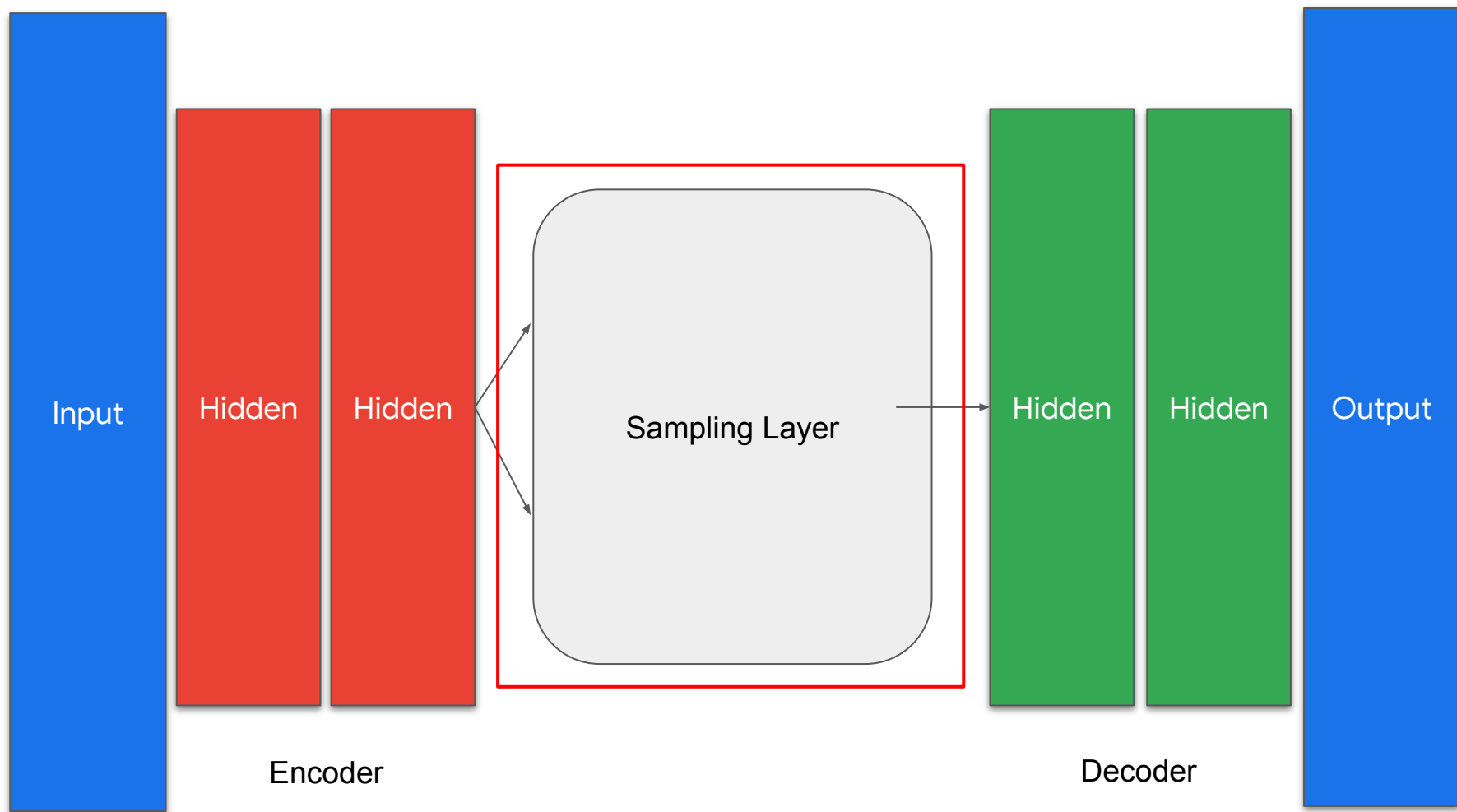
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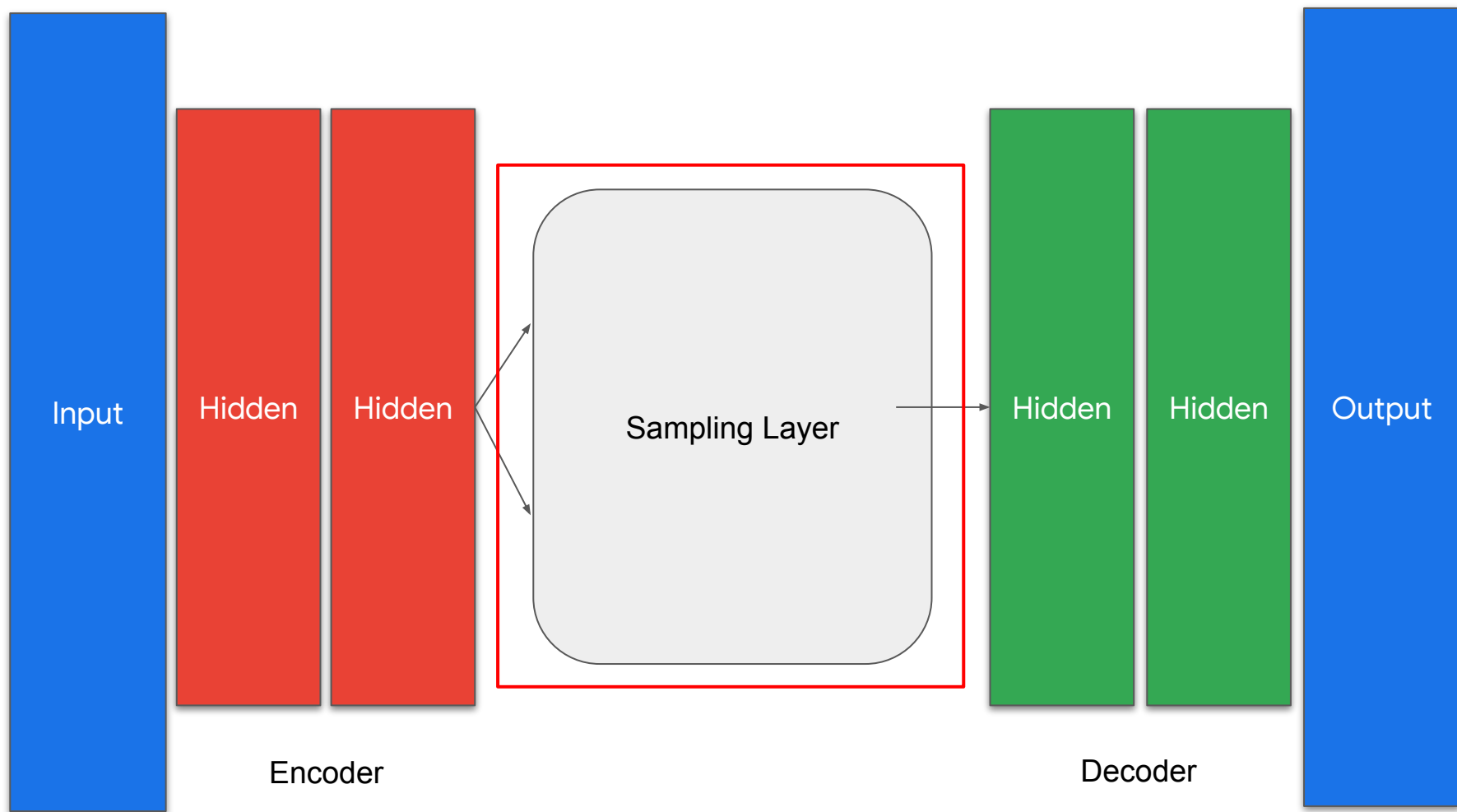

```
class Sampling(tf.keras.layers.Layer):  
    def call(self, inputs):  
        mu, sigma = inputs  
        batch = tf.shape(mu)[0]  
        dim = tf.shape(mu)[1]  
        epsilon = tf.keras.backend.random_normal(shape=(batch, dim))  
        return mu + tf.exp(0.5 * sigma) * epsilon
```

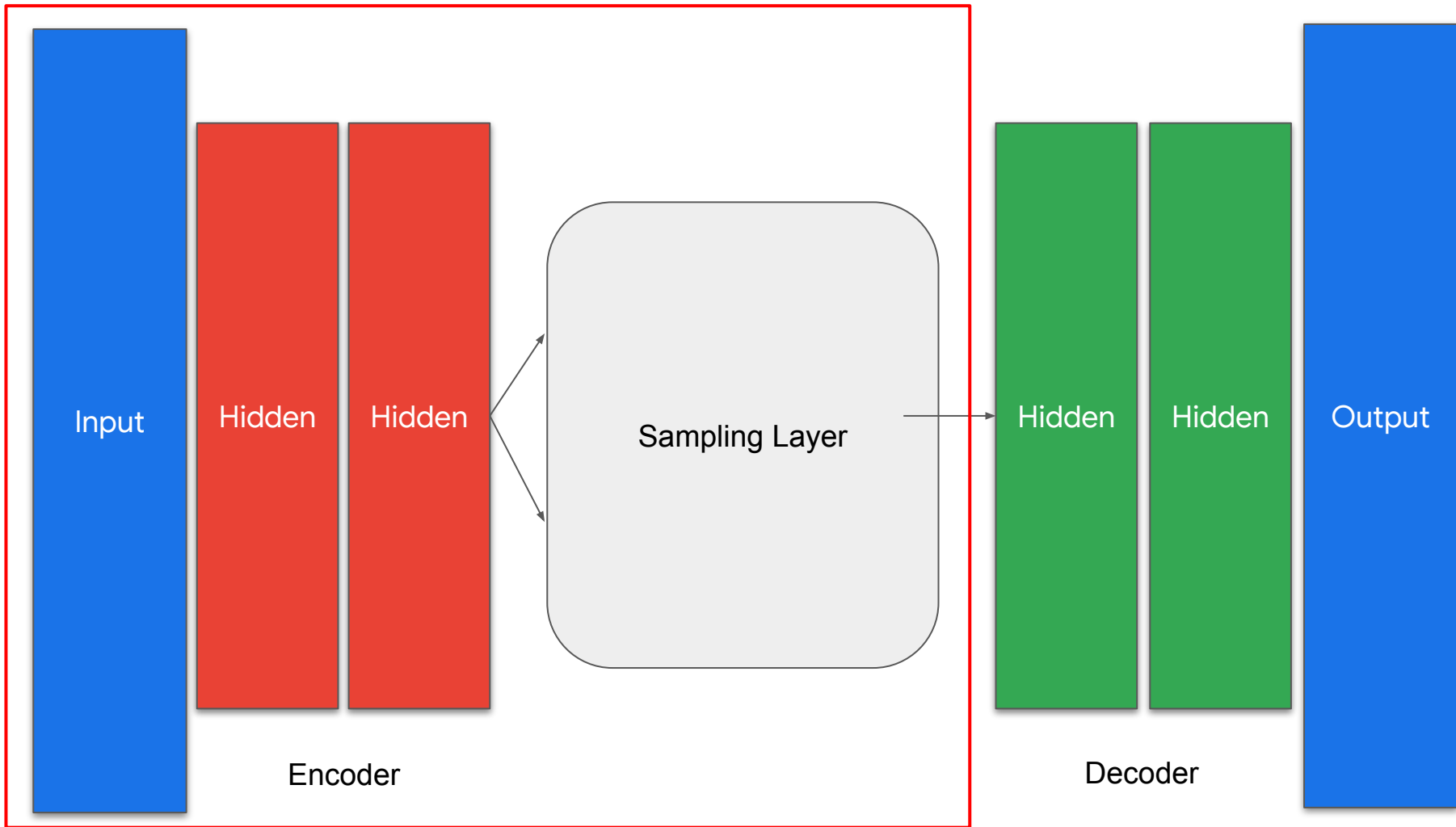
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```
def encoder_model(LATENT_DIM, input_shape):  
    inputs = tf.keras.layers.Input(shape=input_shape)  
    mu, sigma, conv_shape = encoder_layers(inputs, latent_dim=LATENT_DIM)  
    z = Sampling()(mu, sigma)  
    model = tf.keras.Model(inputs, outputs=[mu, sigma, z])  
    return model, conv_shape
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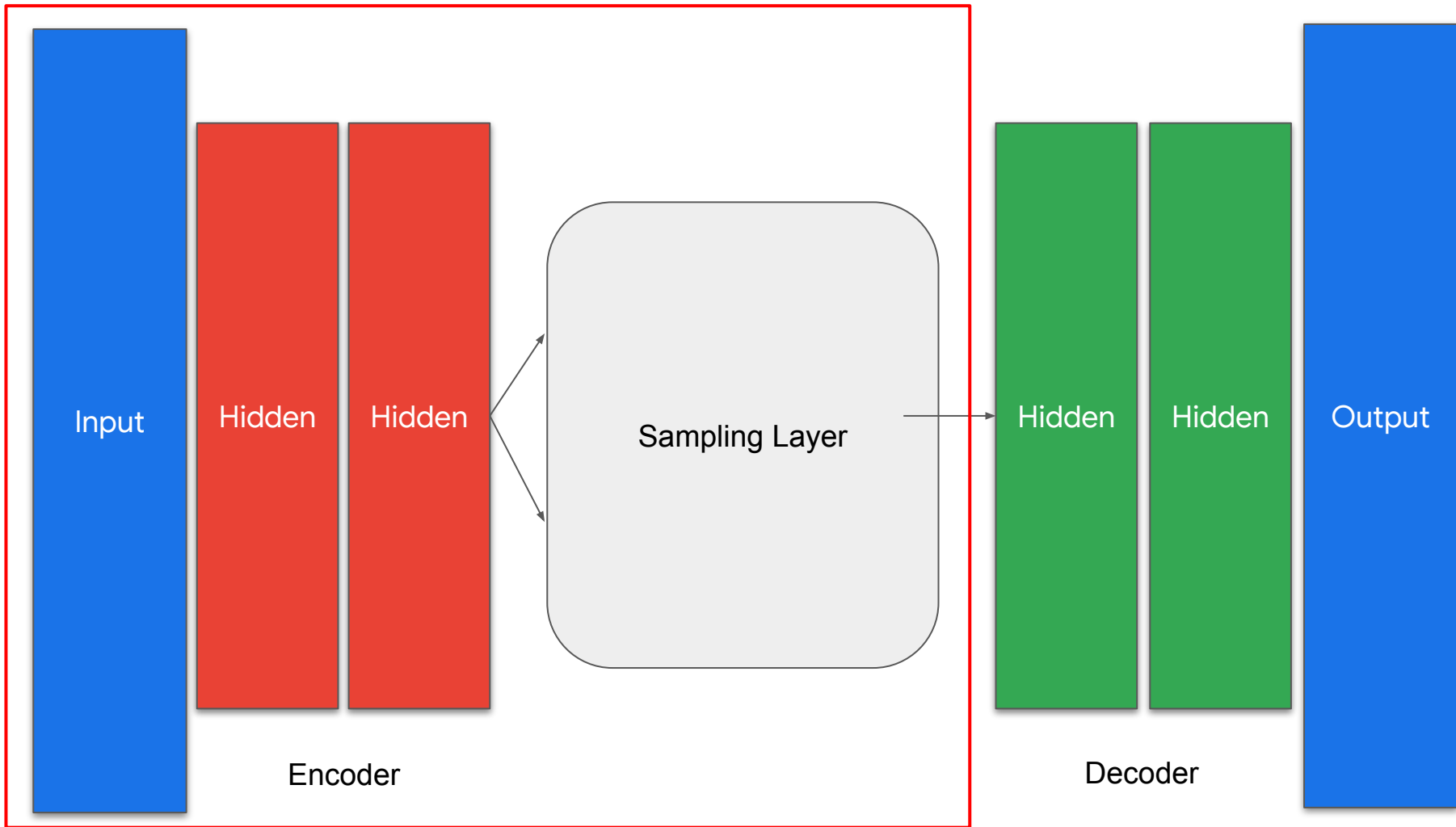
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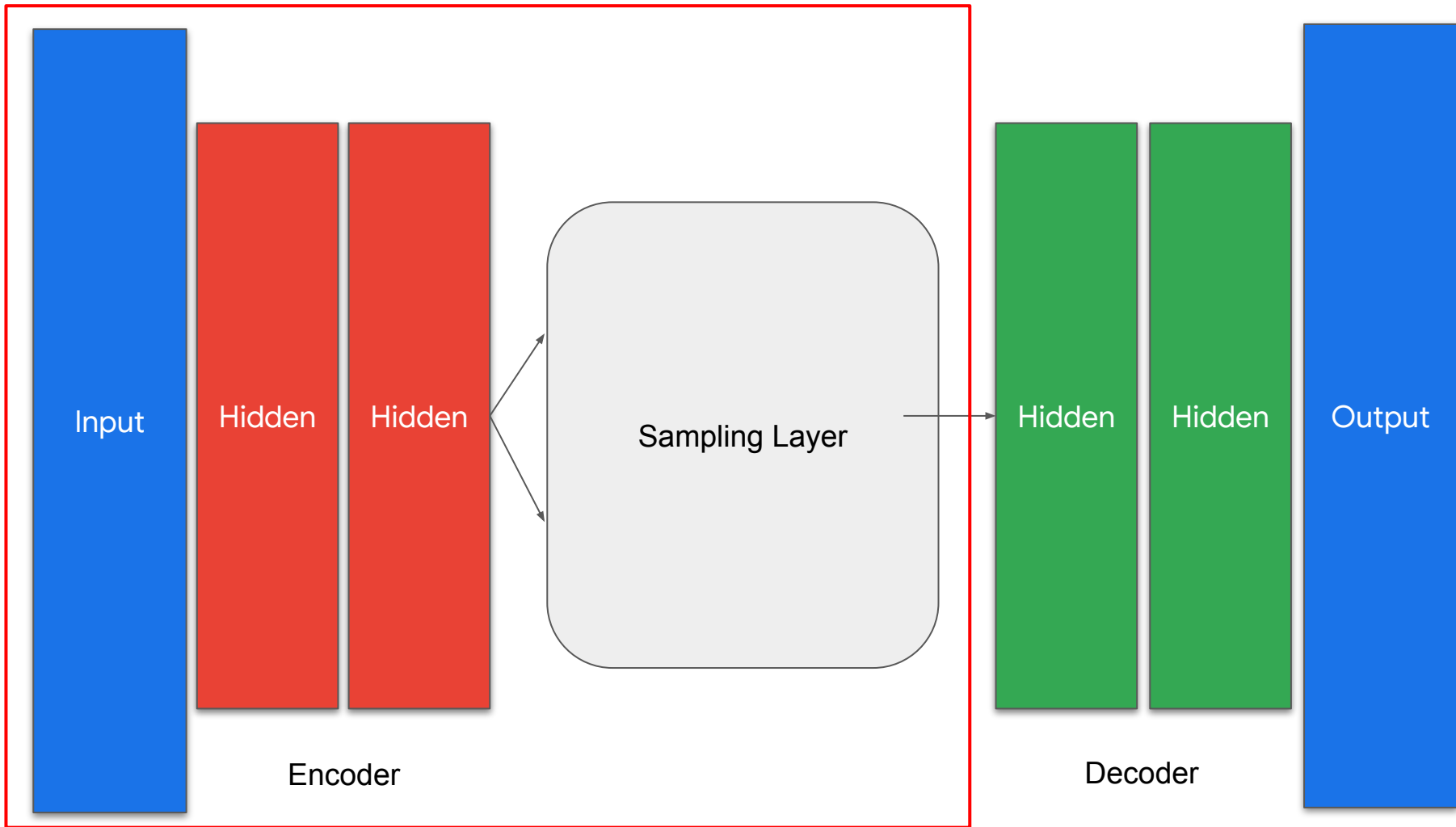
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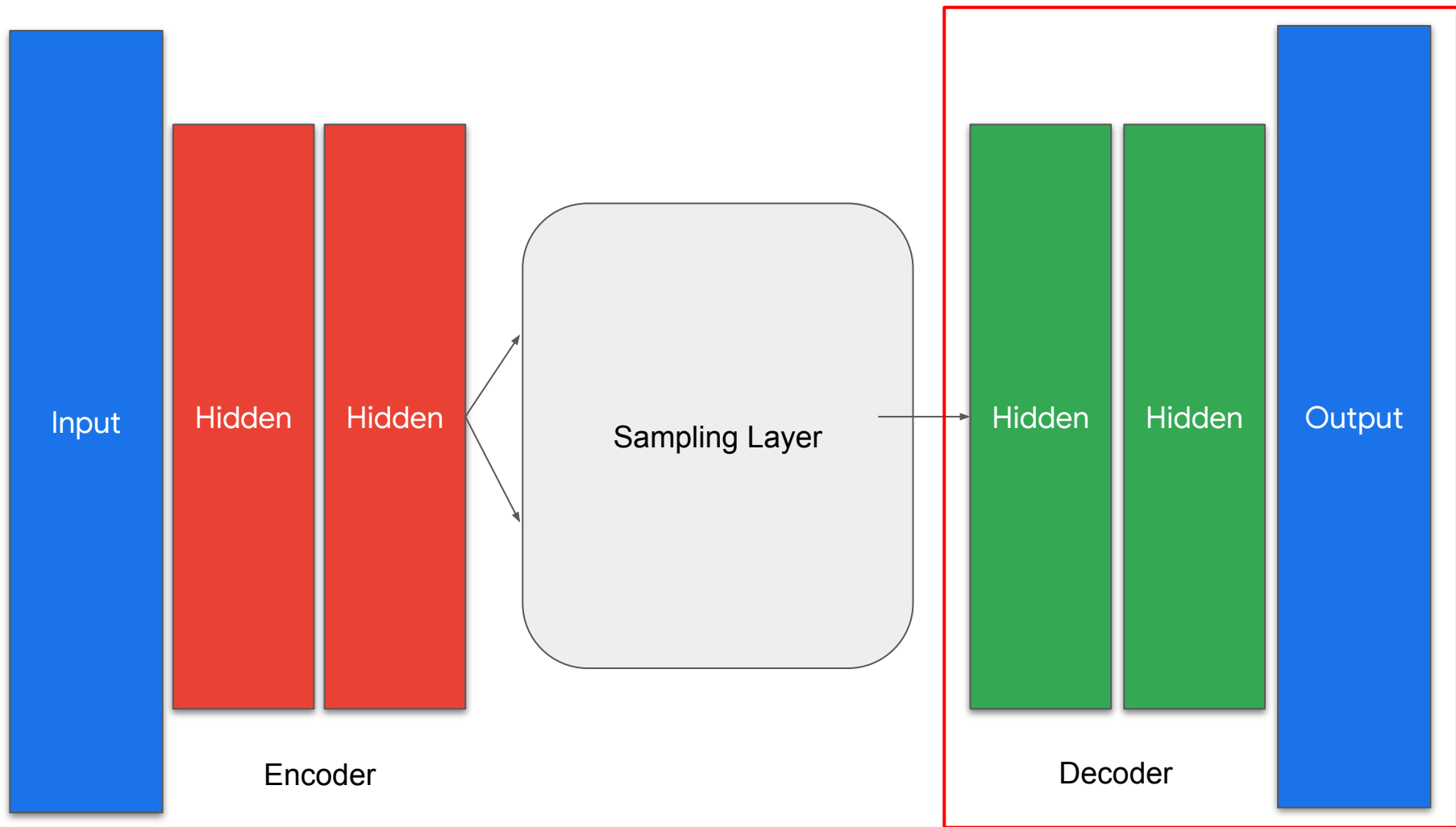
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```







```
def decoder_layers(inputs, conv_shape):
    units = conv_shape[1] * conv_shape[2] * conv_shape[3]
    x = tf.keras.layers.Dense(units, activation = 'relu',
                               name="decode_dense1")(inputs)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Reshape((conv_shape[1], conv_shape[2], conv_shape[3]),
                                  name="decode_reshape")(x)
    x = tf.keras.layers.Conv2DTranspose(filters=64, kernel_size=3, strides=2,
                                          padding='same', activation='relu',
                                          name="decode_conv2d_2")(x)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2DTranspose(filters=32, kernel_size=3, strides=2,
                                          padding='same', activation='relu',
                                          name="decode_conv2d3")(x)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2DTranspose(filters=1, kernel_size=3, strides=1, padding='same',
                                          activation='sigmoid', name="decode_final")(x)

    return x
```

```
def decoder_layers(inputs, conv_shape):
    units = conv_shape[1] * conv_shape[2] * conv_shape[3]
    x = tf.keras.layers.Dense(units, activation = 'relu',
                               name="decode_dense1")(inputs)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Reshape((conv_shape[1], conv_shape[2], conv_shape[3]),
                                name="decode_reshape")(x)
    x = tf.keras.layers.Conv2DTranspose(filters=64, kernel_size=3, strides=2,
                                         padding='same', activation='relu',
                                         name="decode_conv2d_2")(x)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2DTranspose(filters=32, kernel_size=3, strides=2,
                                         padding='same', activation='relu',
                                         name="decode_conv2d3")(x)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2DTranspose(filters=1, kernel_size=3, strides=1, padding='same',
                                         activation='sigmoid', name="decode_final")(x)

    return x
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                                         name="decode_conv2d3")(x)  
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                                         name="decode_conv2d_2")(x)
    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2DTranspose(filters=32, kernel_size=3, strides=2,
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    x = tf.keras.layers.BatchNormalization()(x)

    x = tf.keras.layers.Conv2DTranspose(filters=1, kernel_size=3, strides=1, padding='same',
                                         activation='sigmoid', name="decode_final")(x)

    return x
```

```
def decoder_model(latent_dim, conv_shape):  
    inputs = tf.keras.layers.Input(shape=(latent_dim,))  
    outputs = decoder_layers(inputs, conv_shape)  
    model = tf.keras.Model(inputs, outputs)  
    return model
```

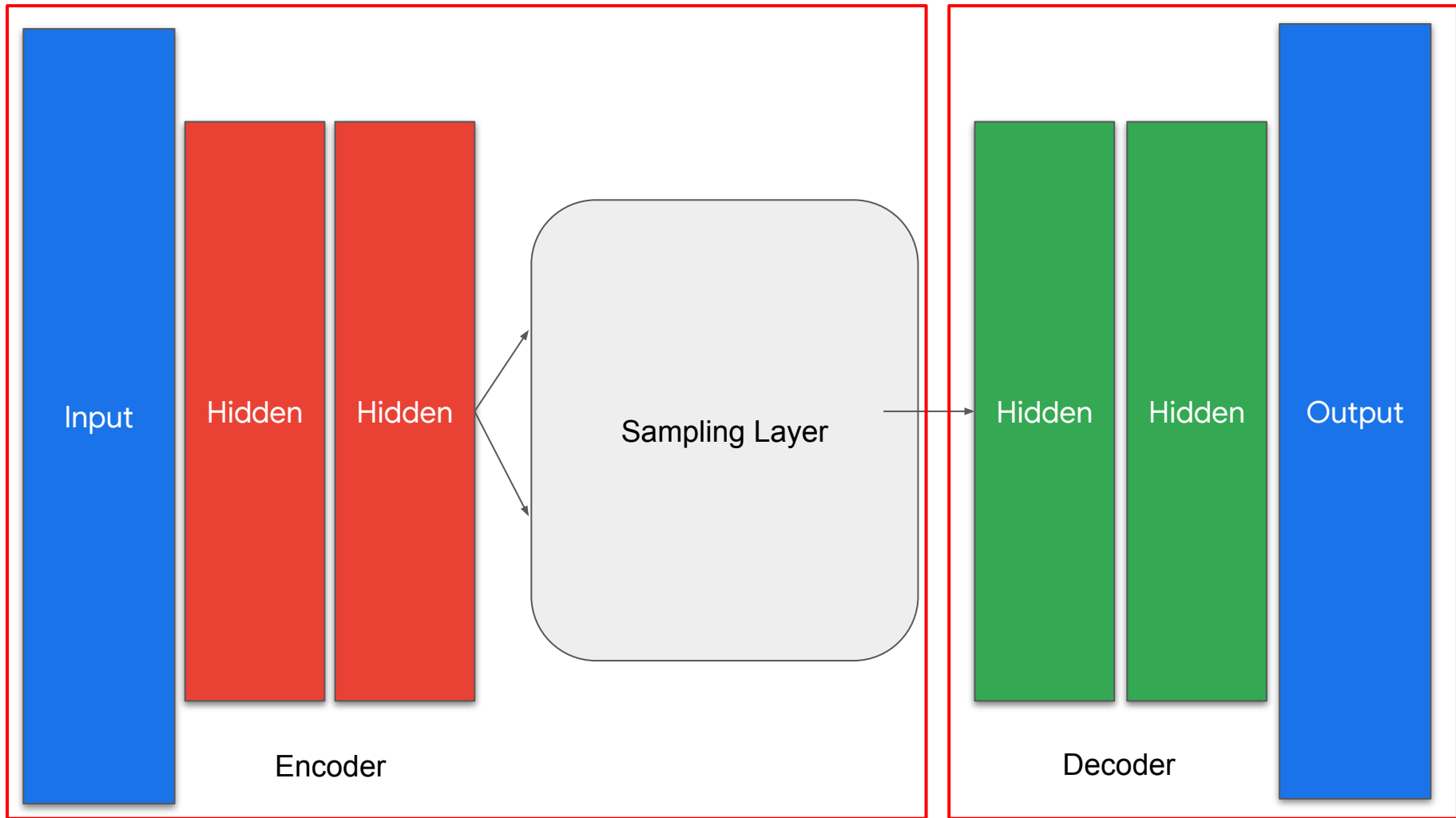
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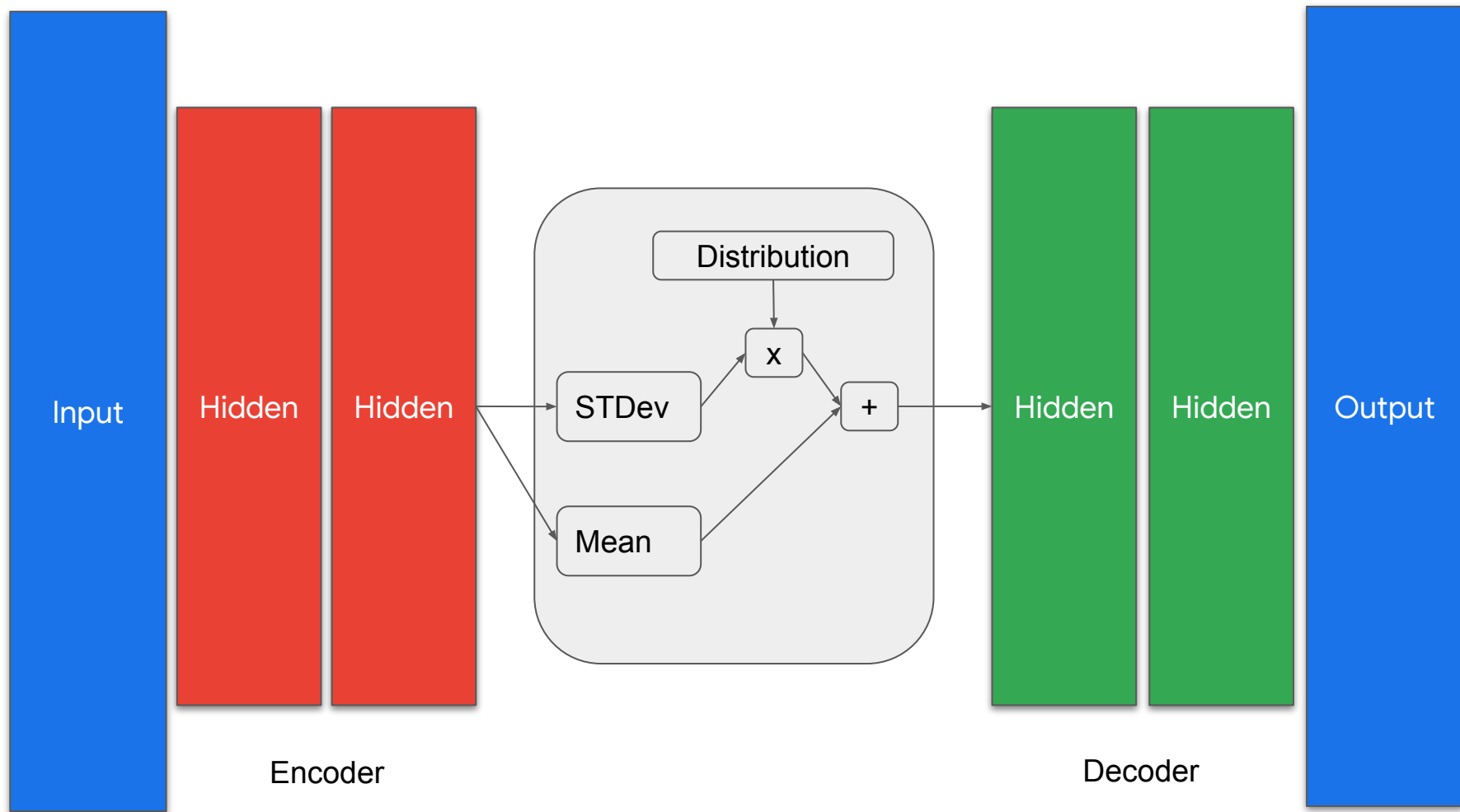


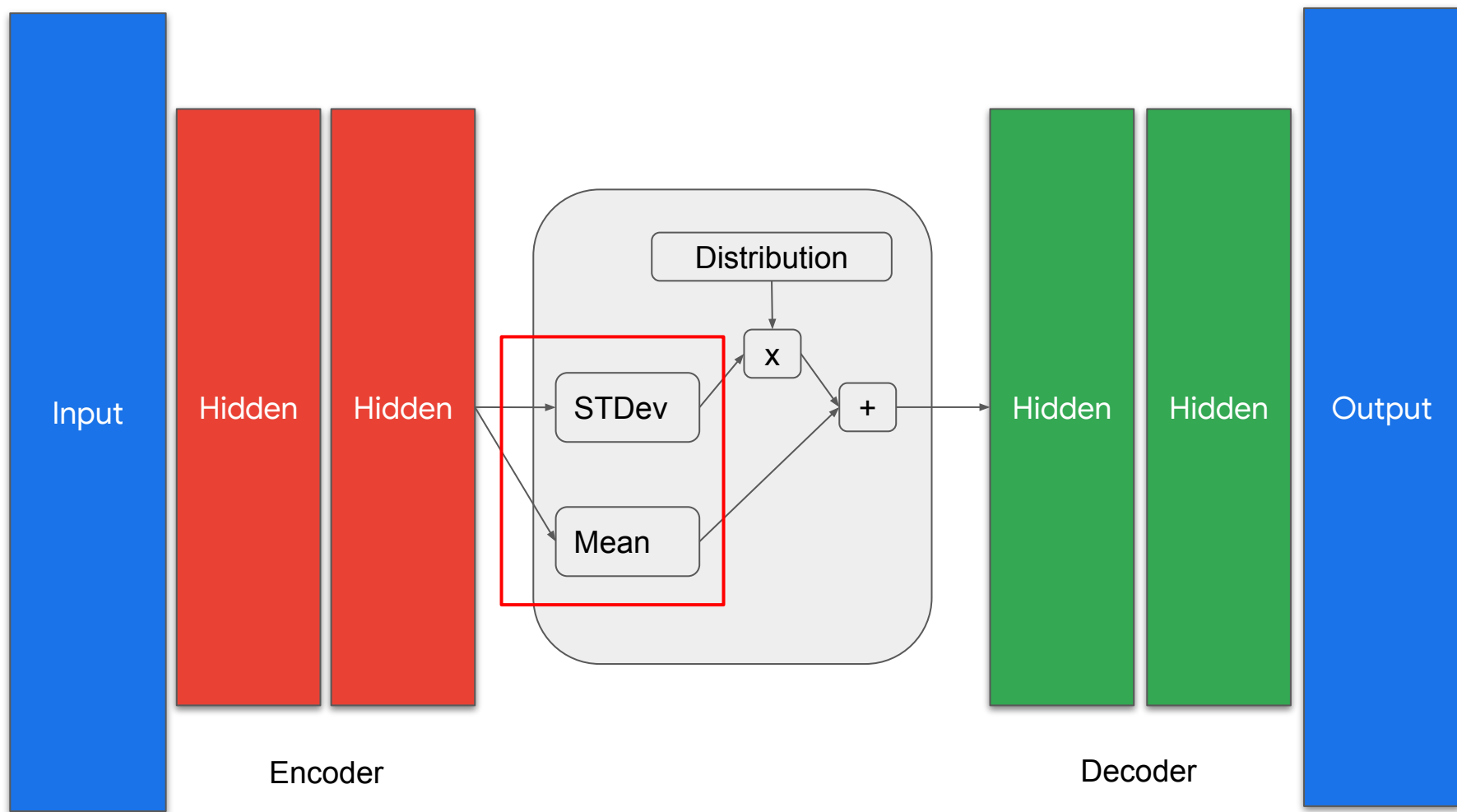
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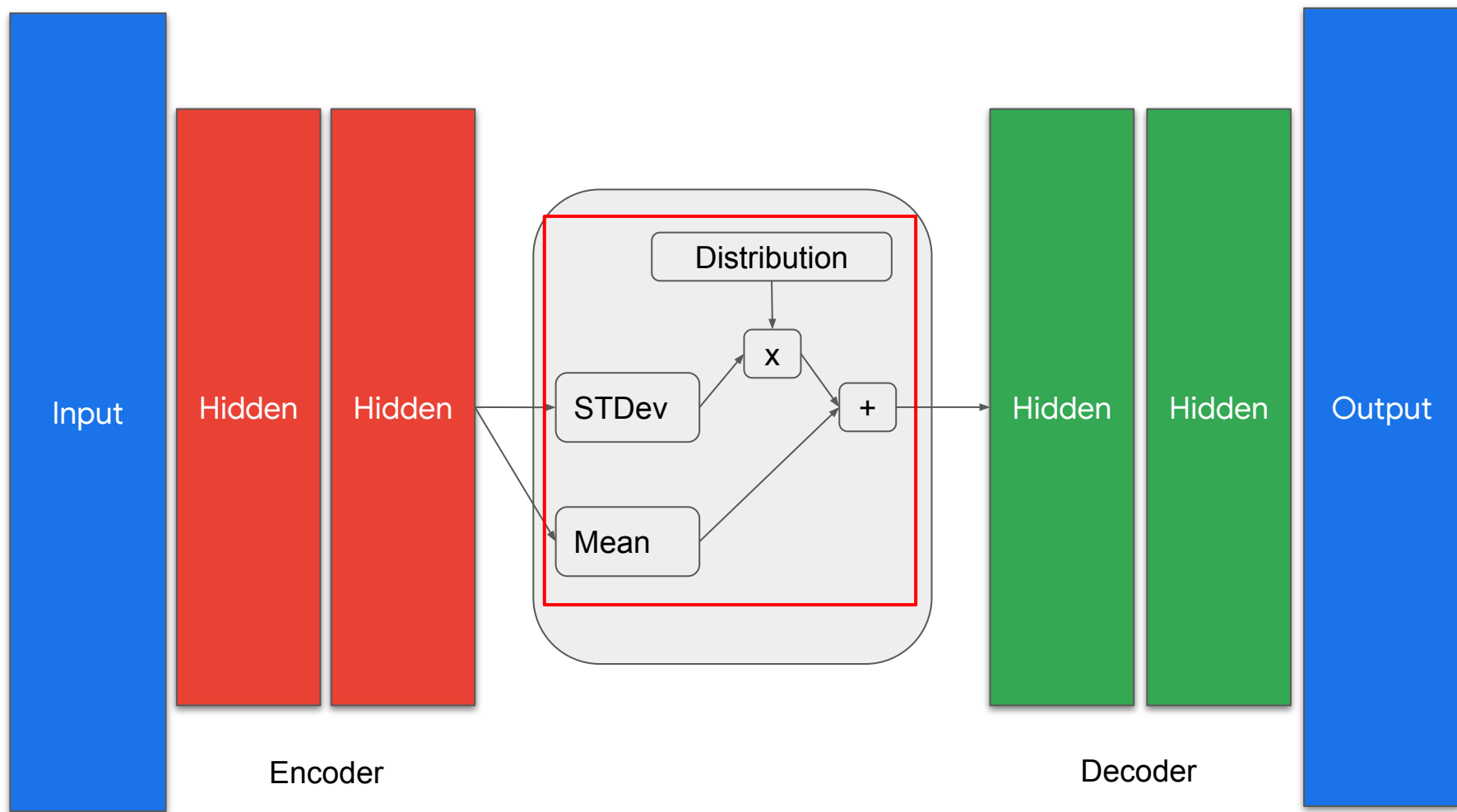
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```
# Define a kl reconstruction loss function
def kl_reconstruction_loss(inputs, outputs, mu, sigma):
    kl_loss = 1 + sigma - tf.square(mu) - tf.math.exp(sigma)
    return tf.reduce_mean(kl_loss) * -0.5
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https://en.wikipedia.org/wiki/Kullback%E2%80%93Leibler_divergence

<https://arxiv.org/abs/2002.07514>

```
def vae_model(encoder, decoder, input_shape):  
    inputs = tf.keras.layers.Input(shape=input_shape)  
    mu = encoder(inputs)[0]  
    sigma = encoder(inputs)[1]  
    z = encoder(inputs)[2]  
    reconstructed = decoder(z)  
    model = tf.keras.Model(inputs=inputs, outputs=reconstructed)  
    loss = kl_reconstruction_loss(inputs, z, mu, sigma)  
    model.add_loss(loss)  
    return model
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```

```
for epoch in range(epochs):  
    for step, x_batch_train in enumerate(train_dataset):  
        with tf.GradientTape() as tape:  
            reconstructed = vae(x_batch_train)  
            flattened_inputs = tf.reshape(x_batch_train, shape=[-1])  
            flattened_outputs = tf.reshape(reconstructed, shape=[-1])  
            loss = bce_loss(flattened_inputs, flattened_outputs) * 784  
            loss += sum(vae.losses) # Add KLD regularization loss  
  
        grads = tape.gradient(loss, vae.trainable_weights)  
        optimizer.apply_gradients(zip(grads, vae.trainable_weights))
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```

epoch: 99, step: 400

