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Meschede, 8th September 2023.

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# Math stuff for pyspark

## 1 Simulation of a Dataset

In order to generate a large dataset which fulfills the requirements ( $n \gg 10^9, k \gg 10^5$ ), the generation of the values needs to be done in a distributed fashion. At first, the following values need to be initialized:

- $n$  - number of rows/samples
- $k$  - number of columns/features
- $\vec{\beta}$  - beta, the coefficients of the function
- $cov$  - a covariance vector that determines the covariance to the first column for each column

In this implementation,  $n$  and  $k$  need to be set by the user while  $\vec{\beta}$  and  $cov$  are generated by numpy. For generating the actual dataset, `pyspark.mllib.random.RandomRDDs.normalVectorRDD(sc, n, k)` is used. This function creates an rdd containing  $n$  vectors, each containing  $k$  entries, where each entry is generated from a standard-normal distribution.

After generating this random noise matrix, the user-defined-function `createRow(noise)` is applied to the rdd, which returns two values,  $\vec{x}$  and  $y$ . With noise as  $\epsilon$  and  $cov$  as  $c$ :

$$\vec{x} = (\epsilon_0, \epsilon_0 c_1 + n_1, \dots, \epsilon_0 c_i + n_i)$$

$$y = \vec{x} \cdot \vec{\beta}$$

Applying this function yields an rdd where the first element is a vector with a moderate covariance structure and some added noise for a more realistic setting, while the second element is the target variable. An exemplary distribution might look like the following figure:

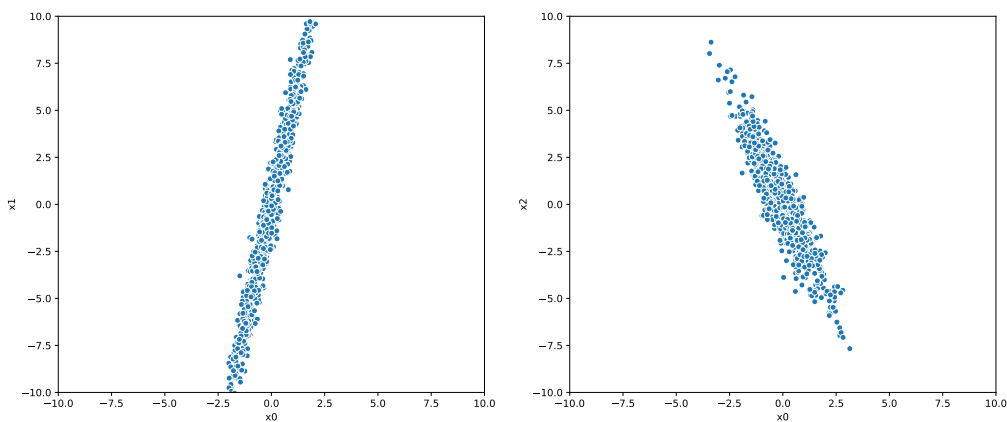


Figure 1: *exemplary generated dataset*