

DISTRIBUTION SHIFT

A Study on Their Effects on Statistical Models and Strategies for Mitigation

Andrea Spinelli, Giacomo Amerio, Giovanni Lucarelli, Tommaso Piscitelli

University of Trieste

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Introduction

Introduction

- Distribution shift is a common problem in machine learning.
- It occurs when the distribution of the training data differs from the distribution of the test data.
- This can lead to a decrease in the performance of the model.
- In this study, we analyze the effects of distribution shift on statistical models and propose strategies for its mitigation.

Data Generation

Training Dataset: Features

The dataset consists of $n = 10^4$ observations with 3 features and 1 target variable.

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The dataset consists of $n = 10^4$ observations with 3 features and 1 target variable. **Features:**

- · $X = (X_1, X_2, X_3) \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$
- $\mu_i \sim \mathcal{U}_{[0,1]}$ for i = 1, 2, 3
- $[\Sigma]_{i,j} \sim \mathcal{U}_{[-1,1]}$ for i, j = 1, 2, 3

Note: The Σ randomly generated has been transformed to a symmetric and positive semidefinite matrix by computing $\Sigma \Sigma^T$.

Training Dataset: Target Variable

Dicothomous target variable: $Y \in \{0,1\}$

$$z = \beta_0 + \sum_{i=1}^{3} \beta_i x_i + \sum_{i=1}^{3} \beta_{ii} x_i^2 + \sum_{i=1}^{2} \sum_{j=i+1}^{3} \beta_{ij} x_i x_j$$
$$Y \sim Be(p), \quad p = \frac{1}{1 + e^{-z}}$$

Testing Dataset

Performance Degradation

Performance Enhancement

Questions?

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