

















Figure 12. Deployment of the model

The performance of the model was verified against the baseline performance, with a sharp increment on the precision of the model, gaining full acceptability by the teams in charge of its development and its exploitation. To illustrate the gains, the FPR decreased by a 27%, meanwhile the precision increased by a 22%.

The full deployment of these technologies can pave the way for an effective and trustable deployment of *Safety Intelligence*. For example, the identification of trends for eliciting recommendations in a large basis can only be supported by data in the origin with a very high level of integrity. In addition, these studies support SESAR funded project such as FARO [13].

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