An introduction to Model Drift

with



Alyona Galyeva

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Agenda

- Intro and workshop setup
- Model performance
- Handling the drift
- Test-based monitoring
- Closing remarks







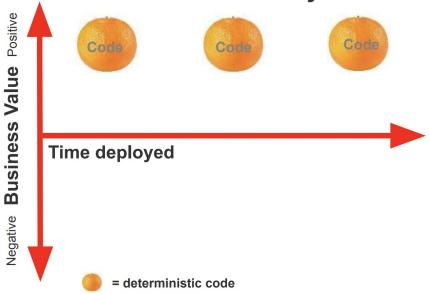


Workshop setup

Clone GitHub Repo and follow README.md instructions

FORRESTER

Code is deterministic and always runs as written



Traditional software

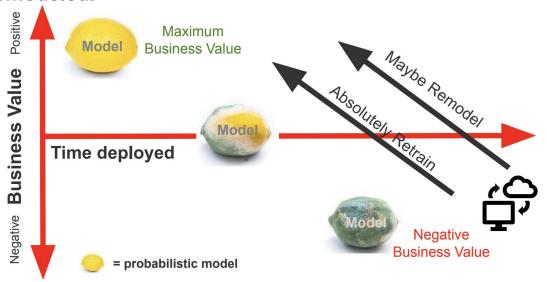
Software performance

Usually a DevOps team monitors and maintains the system in production.

https://neptune.ai/blog/how-to-monitor-your-models-in-production-guide

FORRESTER

ML models must be monitored, retrained, and often remodeled.



Al Software

Model performance

Machine learning models degrade over time as they are sensitive to changes in the real world.

https://neptune.ai/blog/how-to-monitor-your-models-in-production-guide

Why model monitoring is hard?

In your case monitoring and maintenance has to be a shared responsibility.

Data (Input)

Data Quality

Issues

Model

Predictions (Output)

Model Drift

Model Evaluation Metrics

Data/Feature Drift

Model Configuration

Prediction Drift

Outliers

Model Version

Concerted Adversaries

https://neptune.ai/blog/how-to-monitor-your-models-in-production-guide

Who Monitors What?







Data Scientists and/or Machine Learning Engineers

What is Evidently?

Evidently is an open-source Python library for data scientists and ML engineers.

It helps evaluate, test, and monitor the performance of ML models from validation to production.

Tests:

Batch model checks

Reports:

Interactive dashboards

Real-time
ML monitoring
(early dev)

Different types of drift

Data drift (input)

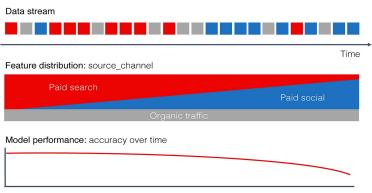
Describes how the (training) data changes over time for the model.

Model drift (in between)

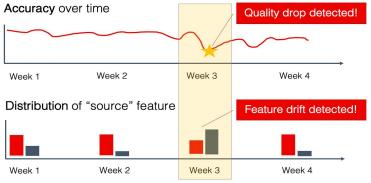
A model's performance might change over time. It can be caused by target drift, data drift, or simply a bug in the model's code.

Target drift (output)

Describes how the target variable changes.



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Test-based monitoring

Test-based monitoring in Evidently performs tests with a clear **pass** or **fail** result against predefined expectations.

This can be useful when

- You want to send an alert when there is a significant difference in the production data
- When the model's performance suddenly deteriorates
- Logging and monitoring of performance

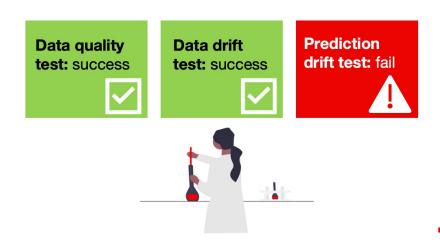


Image sources: Evidently.ai

Closing remarks

Interested in integrating Evidently with Grafana, Airflow, MLflow, Metaflow?

Interested in contributing to Evidently? Follow these guidelines



```
if questions:
    try:
        answer()
    except RuntimeError:
        pass
else:
    print("Thank You.")
```