



## Exercise: Predicting Customer Churn Using Time-Series Data

**Objective:** Build a **robust** classification model that predicts which customers will churn between **Jan. 1, 2024** and **Feb. 28, 2024**, using **time-series data** with a double index (**customer\_id** and **date**). Demonstrate modeling and feature engineering skills, as well as production-aware design practices.

**Deliverables:** Produce a GitHub repository named **customer-churn-classification** that contains:

1. A **README.md** file to shortly explain the purpose, setup, and usage of the repository.
2. The fully functional code in clearly commented **.py** scripts (no notebook).
3. The original **.csv** file, with an additional prediction column appended.
4. A **.json** file containing precision, recall, and at least one additional relevant metric.
5. A trained model artifact (e.g., a **.pkl** file).

**Estimated time to completion:** up to 120 minutes of focused work time. The emphasis is on **quality over completeness**. If you can't implement everything perfectly within that time, explain how you'd **prioritize** and **extend** your solution in a real-world scenario.

**Data Overview:** The dataset (**churn\_data.csv**) contains a double index (**customer\_id**, **date**), several features, and a **target variable** (**churn**, which is binary: **1** for churned, **0** otherwise). The data has been synthetically generated but should be treated as real-world data.

### Instructions:

1. **Data Preprocessing:**
  - Handle missing values.
  - Appropriately encode (1-hot, embeddings, etc.) categorical features.
2. **Data Engineering:**
  - Generate at least 3 relevant **date**-dependent features.
  - Enrich the data with one relevant external source (e.g., economic indicators).
3. **Modeling:**
  - Select an appropriate algorithm for **churn** classification, and implement it. Make sure to systematically tune and validate your model.
  - Be aware of data leakage, and provide a clear rationale for your choices.
4. **Model explanation:**
  - Use an appropriate method for **brief** model explanation (e.g., SHAP, PDP, LIME).
  - Shortly describe why the method you chose is technically relevant.