8/10/25, 5:51 PM pipeline

```
In [3]: """
        Ejemplo mínimo de uso de scikit-learn Pipeline con pasos:
        1) Extracción de datos (desde CSV).
        2) Filtrado de registros fuera de rango.
        3) Tipado y preprocesamiento por tipo de variable (numéricas y categóricas).
        4) Separación del dataset para ML y entrenamiento de un modelo.
        Requisitos: pandas, scikit-learn.
        from pathlib import Path
        import pandas as pd
        from sklearn.model_selection import train_test_split
        from sklearn.pipeline import Pipeline
        from sklearn.compose import ColumnTransformer
        from sklearn.impute import SimpleImputer
        from sklearn.preprocessing import OneHotEncoder, StandardScaler
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import accuracy_score, classification_report
        # 1) EXTRACCIÓN DE DATOS
        # -----
        DATA PATH = "./synthetic customers.csv"
        df = pd.read_csv(DATA_PATH)
        # 2) FILTRADO
        # - Mantener edades 18-80
        # - income gtg > 0
        # - monthly_visits en [0, 30]
        mask = (
            (df["age"].between(18, 80)) &
            (df["income qtq"] > 0) &
            (df["monthly visits"].between(0, 30))
        df = df.loc[mask].reset_index(drop=True)
        # 3) TIPOS DE VARIABLES
        # - Definir columnas numéricas vs categóricas
             - Preprocesar: imputación, escalado y one-hot
        numeric_features = ["age", "income_gtq", "monthly_visits"]
        categorical_features = ["city", "has_kids"]
        numeric_pipeline = Pipeline(steps=[
            ("imputer", SimpleImputer(strategy="median")),
            ("scaler", StandardScaler())
        1)
        categorical pipeline = Pipeline(steps=[
            ("imputer", SimpleImputer(strategy="most_frequent")),
            ("onehot", OneHotEncoder(handle unknown="ignore"))
        1)
```

8/10/25, 5:51 PM pipeline

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preprocessor = ColumnTransformer(
   transformers=[
        ("num", numeric_pipeline, numeric_features),
        ("cat", categorical_pipeline, categorical_features),
   1
)
# 4) SEPARACIÓN DEL DATASET PARA ML
    - X, y
     - train/test split

    Pipeline completo con modelo final

X = df[numeric features + categorical features]
y = df["churn"]
X_train, X_test, y_train, y_test = train_test_split(
   X, y, test_size=0.25, random_state=42, stratify=y
clf = Pipeline(steps=[
    ("preprocess", preprocessor),
    ("model", LogisticRegression(max_iter=1000))
1)
# Entrenar
clf.fit(X_train, y_train)
# Evaluar
y pred = clf.predict(X test)
acc = accuracy_score(y_test, y_pred)
print(f"Accuracy: {acc:.3f}")
print(classification_report(y_test, y_pred, zero_division=0))
```

Accuracy: 0.947

```
precision recall f1-score
                                             support
                            0.92
                  1.00
                                      0.96
                                                  13
           1
                            1.00
                  0.86
                                      0.92
                                                   6
                                      0.95
                                                  19
    accuracy
   macro avg
                  0.93
                            0.96
                                       0.94
                                                  19
                                      0.95
weighted avg
                  0.95
                            0.95
                                                  19
```

```
In [8]: from sklearn import set_config
set_config(display='diagram')
clf
```

8/10/25, 5:51 PM pipeline

