IUM 25L - Projekt

Advanced model

```
In [1]: from sklearn.metrics import mean_absolute_error, median_absolute_error, accuracy
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import pprint

from nocarz.config import PROCESSED_DATA_DIR, MODELS_DIR, INPUT_COLUMNS, NUMERIC
from nocarz.src.advanced_model import AdvancedModel
```

Preparing the data

```
In [2]: train_data = pd.read_csv(PROCESSED_DATA_DIR / "train.csv")
    test_data = pd.read_csv(PROCESSED_DATA_DIR / "test.csv")

train_data = train_data.dropna(subset=NUMERICAL_TARGETS + CATEGORICAL_TARGETS).r

for col in INPUT_COLUMNS:
    train_data[col] = train_data[col].fillna("")
    test_data[col] = test_data[col].fillna("")
```

Training the model

```
In [3]: advanced_model = AdvancedModel()
   advanced_model.fit(train_data[INPUT_COLUMNS], train_data[NUMERICAL_TARGETS + CAT
   advanced_model.save(MODELS_DIR / "advanced_model.pkl")
```

Making predictions

```
In [4]:
    results = []

for idx, row in test_data.iterrows():
        predictions = advanced_model.predict(row[INPUT_COLUMNS])

        actual_values = row.to_dict()
        evaluation = AdvancedModel.evaluate_predictions(predictions, actual_values)

    result = {
        'listing_id': idx,
        'host_id': row['host_id'],
        'predictions': predictions,
        'evaluation': evaluation
    }
    results.append(result)
```

Calculating the metrics

```
In [5]: metrics = {}
        pred_numerical = {col: [] for col in NUMERICAL_TARGETS}
        true_numerical = {col: [] for col in NUMERICAL_TARGETS}
        pred_categorical = {col: [] for col in CATEGORICAL_TARGETS}
        true_categorical = {col: [] for col in CATEGORICAL_TARGETS}
        for result in results:
            predictions = result['predictions']
            evaluation = result['evaluation']
            for col in NUMERICAL_TARGETS:
                if col in evaluation and evaluation[col]['type'] == 'numerical':
                    pred_numerical[col].append(evaluation[col]['predicted'])
                    true_numerical[col].append(evaluation[col]['actual'])
            for col in CATEGORICAL_TARGETS:
                if col in evaluation and evaluation[col]['type'] == 'categorical':
                    pred_categorical[col].append(evaluation[col]['predicted'])
                    true_categorical[col].append(evaluation[col]['actual'])
        for col in NUMERICAL TARGETS:
            if len(pred_numerical[col]) > 0:
                y_true = np.array(true_numerical[col])
                y_pred = np.array(pred_numerical[col])
                metrics[col] = {
                    "mean_absolute_error": mean_absolute_error(y_true, y_pred),
                    "median_absolute_error": median_absolute_error(y_true, y_pred),
                    "count": len(y_true)
        for col in CATEGORICAL TARGETS:
            if len(pred_categorical[col]) > 0:
                y_true = true_categorical[col]
                y_pred = pred_categorical[col]
                metrics[col] = {
                     "accuracy": accuracy_score(y_true, y_pred),
                    "count": len(y true)
                }
        pprint.pprint(metrics)
```

```
{'accommodates': {'count': 552,
                          'mean_absolute_error': 0.8498042328042328,
                         'median_absolute_error': np.float64(0.66999999999999)},
        'bathrooms': {'count': 405,
                      'mean_absolute_error': 0.2344934646286498,
                      'median_absolute_error': np.float64(0.094999999999999)},
        'bathrooms_text': {'accuracy': 0.6557971014492754, 'count': 552},
        'bedrooms': {'count': 499,
                     'mean_absolute_error': 0.40312606164710374,
                     'median_absolute_error': np.float64(0.229999999999999)},
        'beds': {'count': 404,
                 'mean_absolute_error': 0.6265008840169731,
                 'median_absolute_error': np.float64(0.439999999999999)},
        'price': {'count': 405,
                  'mean_absolute_error': 60.36933454830491,
                  'median_absolute_error': np.float64(30.5900000000000)},
        'property_type': {'accuracy': 0.7590579710144928, 'count': 552},
        'room_type': {'accuracy': 0.907608695652174, 'count': 552}}
In [6]: RED = '#ff9999'
        BLUE = '#66b3ff'
        plt.figure(figsize=(14, 10))
        # 1. Create a subplot for price MAE
        plt.subplot(2, 2, 1)
        if 'price' in metrics:
            price_mae = metrics['price']['mean_absolute_error']
            price_med = metrics['price']['median_absolute_error']
            plt.bar(["Mean Absolute Error", "Median Absolute Error"], [price_mae, price_
                    color=[RED, BLUE])
            plt.ylabel("Error Value ($)")
            plt.title("Price Error Metrics")
            plt.text(0, price_mae + price_mae*0.02, f"${price_mae:.2f}", ha="center")
            plt.text(1, price_med + price_med*0.02, f"${price_med:.2f}", ha="center")
        # 2. Create a subplot for numerical columns MAE
        plt.subplot(2, 2, 2)
        non_price_cols = [col for col in NUMERICAL_TARGETS if col != "price" and col in
        non_price_maes = [metrics[col]['mean_absolute_error'] for col in non_price_cols]
        if non_price_maes:
            plt.bar(non_price_cols, non_price_maes, color=BLUE)
            plt.title("Numerical Columns Mean Absolute Error")
            plt.ylabel("Error Value")
            plt.xticks(rotation=45)
            for i, val in enumerate(non_price_maes):
                plt.text(i, val + val*0.02, f"{val:.2f}", ha="center")
        # 3. Create a subplot for categorical accuracy
        plt.subplot(2, 2, 3)
        categorical_cols = [col for col in CATEGORICAL_TARGETS if col in metrics]
        accuracies = [metrics[col]['accuracy'] for col in categorical_cols]
        if accuracies:
            plt.bar(categorical_cols, accuracies, color=RED)
            plt.ylabel("Accuracy")
            plt.ylim(0, 1)
            plt.title("Categorical Columns Accuracy")
            plt.xticks(rotation=45)
```

```
for i, val in enumerate(accuracies):
            plt.text(i, val + 0.015, f"{val:.1%}", ha="center")
  # 4. Number of predictions
  plt.subplot(2, 2, 4)
  num_predictions = []
  bar_colors = []
  labels = []
  for col in NUMERICAL_TARGETS:
       num_predictions.append(metrics[col]['count'])
      bar_colors.append(BLUE)
      labels.append(col)
  for col in CATEGORICAL_TARGETS:
      num_predictions.append(metrics[col]['count'])
      bar_colors.append(RED)
      labels.append(col)
  plt.bar(labels, num_predictions, color=bar_colors)
  plt.ylabel("Number of Predictions")
  plt.title("Number of Predictions per Target")
  plt.xticks(rotation=45)
  for i, val in enumerate(num_predictions):
       plt.text(i, val + 0.02 * max(num_predictions), str(val), ha="center")
  plt.tight_layout()
  plt.show()
                                                               Numerical Columns Mean Absolute Error
                    Price Error Metrics
             $60.37
 60
                                                   0.8
 50
                                                   0.7
                                                   0.6
€ 40
                                                 o.5
Value
00
                                   $30.59
                                                 0.4
                                                   0.3
                                                   0.2
 10
                                                   0.1
                                                   0.0
                               Median Absolute Error
         Mean Absolute Error
                                                                                          beds
                 Categorical Columns Accuracy
                                                                  Number of Predictions per Target
 1.0
                        90.8%
                                                  500
          75.9%
                                       65.6%
                                                  400
Accuracy
6.0
                                                  300
                                                 Number 200
 0.2
                                                  100
 0.0
```