Hotel Price Prediction Example Project



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Plan

- 1. Scraping Hotel Price Data using Python, BeautifulSoup, Selenium.
- 2. Cleaning Data with Power Query, Excel.
- 3. Managing and Analyzing Data with SQL, MySQL.
- 4. Visualizing Data with Looker Studio.
- 5. Machine Learning Regression Model in PyTorch.



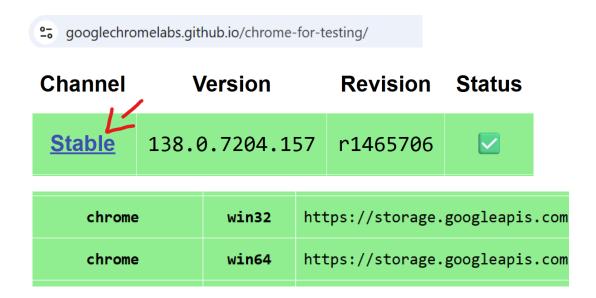
Concept

- 1. Comparing Hotel Prices in Taipei Metropolitain Area
- 2. Based on Month and Franchise Brand
- 3. Comparing chains of Marriott, Hilton & IHG
- 4. Analyze Trends and Extract a Lesson.
- 5. Build and Train Machine Learning Model to Predict Prices.



1A Download and Use Chromedriver

- Required for scraping through Chrome Browser
- Choosing location, as path is needed for further process.





1B Choosing Booking.com for reference

- Easier to use than several separate hotel websites, because the scraping does not need to prepare for different clickables, cookie walls and protection systems.
- From a statistical best practice point of view, using the same reference base allows more justified comparison. Third-party sites may add some fees, therefore using one source for all the different Hotels and Dates, makes the comparison fairer.



1C Creating Base Folder in Visual Studio

- For Scraping Booking.com, we opt for using Python code, with Selenium & BeautifulSoup.
- We create base folder for the project, with scraping_data.py for command.





1D Booking URL Logic

• After checking source code, the URL of Booking follows the below logic:

www.booking.com/hotel/tw/hilton-taipei-sinban.en-gb.html?checkin={checkin}&checkout={checkout}

- We select 6 hotels from Taipei Metropolitain Area, 2 each from the 3 franchise chains:
 - Marriott Taipei (Marriott)
 - Four Points by Sheraton Linkou (Marriott)
 - Hilton Taipei Sinban (Hilton)
 - Humble House Curio Collection (Hilton)
 - Holiday Inn Express Taoyuan (IHG)
 - Holiday Inn Express Taipei Main Station (IHG)
- We collect the right URL for each hotel, then add in {checkin} and {checkout}, as we plan to vary the dates.



1E Import scraping libraries

• We inquire Selenium, BeautifulSoup, and all scraping libraries, import them to scraping_data.py.

```
import time
import csv
from selenium import webdriver
from selenium.webdriver.chrome.service import Service
from selenium.webdriver.common.by import By
from selenium.webdriver.chrome.options import Options
from bs4 import BeautifulSoup
```



1F Complete Python Code for Scraping

Use Chromedriver path and indicate output file as hotel_prices.csv

```
CHROMEDRIVER_PATH = r"C:\chromedriver\d
OUTPUT_FILE = "hotel_prices.csv"
```

• Use previously collected hotel links.



1G Complete Python Code for Scraping

Precise the dates for which we want to collect data from each hotel.

```
DATES = [

("2025-08-06", "2025-08-07"),
("2025-08-07", "2025-08-08"),
("2025-09-02", "2025-09-03"),
("2025-09-03", "2025-09-04"),
("2025-10-02", "2025-10-03"),
("2025-10-15", "2025-10-16"),
("2025-11-02", "2025-11-03"),
("2025-11-15", "2025-11-16"),
("2025-12-02", "2025-12-03"),
("2025-12-15", "2025-12-16"),
]
```

Get driver.

```
def get_driver():
    service = Service(CHROMEDRIVER_PATH)
    options = Options()
    options.add_argument("--start-maximized")
    driver = webdriver.Chrome(service=service, options=options)
    return driver
```

1H Avoid clickables and cookie windows

- By defining accept_cookies function, which takes one parameter, driver.
- Tries to find and click a cookie acceptance button using Selenium.

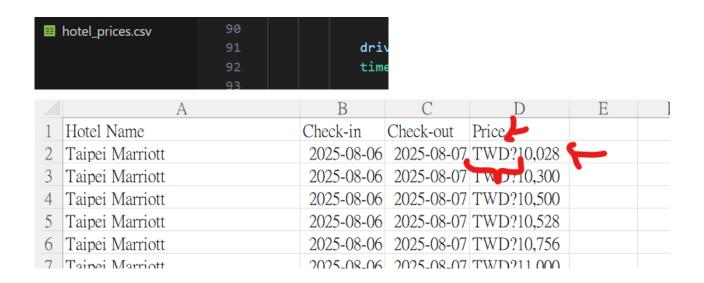
```
def accept_cookies(driver):
    try:
        time.sleep(3) # time to load
        button = driver.find_element(
            By.XPATH, "//button[contains(text(),'Accept')]")
        button.click()
        print("Cookie banner accepted")
    except Exception:
        print("No cookie banner found or clickable")
```

11 Double loop through hotels and dates

• With 5 seconds waiting time to give response time to booking.com

```
for hotel in HOTELS:
   for checkin, checkout in DATES:
       url = hotel["base_url"].format(checkin=checkin, checkout=checkout)
       print(f"Loading URL: {url}")
       driver.get(url)
       time.sleep(5) # initial load
       accept cookies(driver)
       prices = extract prices(driver)
       if prices:
           for price in prices:
                all_data.append([hotel["name"], checkin, checkout, price])
           print(
                f"Prices found for {hotel['name']} on {checkin} - {checkout}")
       else:
           print(
                f"No prices found for {hotel['name']} on {checkin} - {checkout}")
```

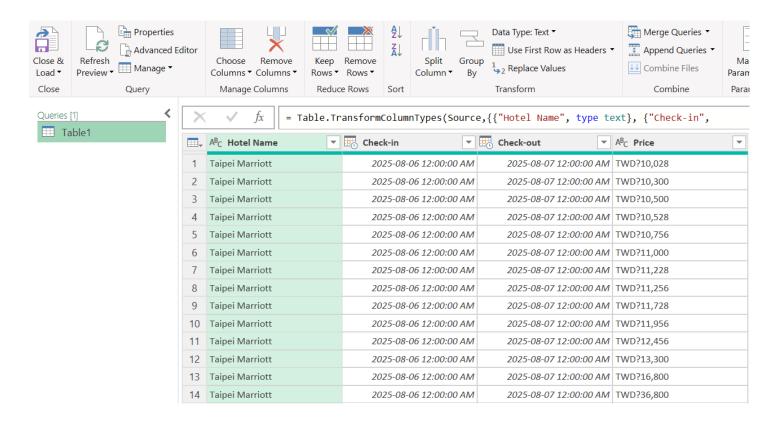
1J Save to CSV



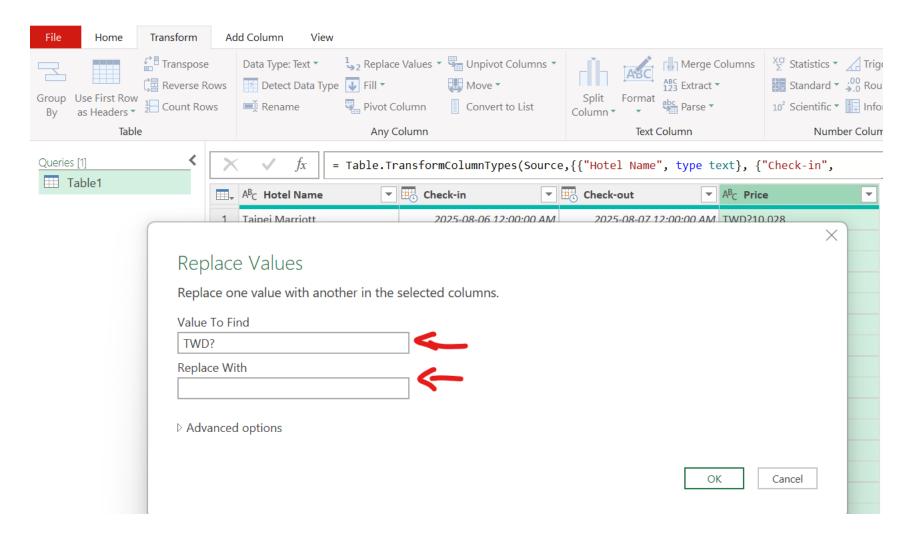
- In Excel, it is visible that Price is not in ideal format for analysis.
- We can use Power Query to clean the data.

2A Open Power Query Editor

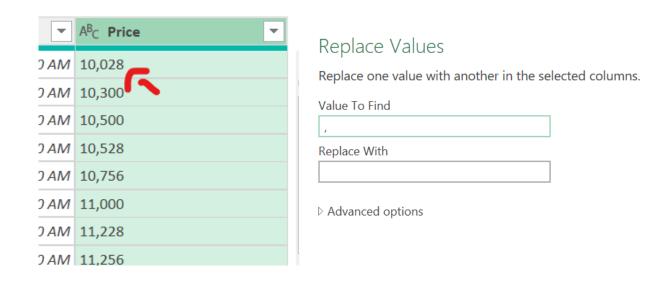
• Use get data from table / range.



2B Transform Price Column by Replace



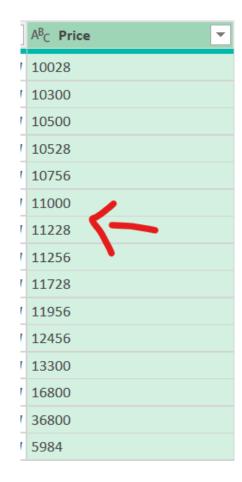
2C Remove commas and adjust dates







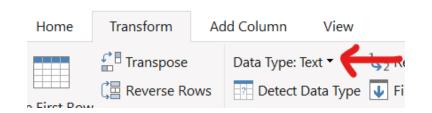
Date only, remove time, as it is not necessary for our analysis.

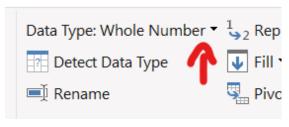




2D Change Data Type from Text to Number

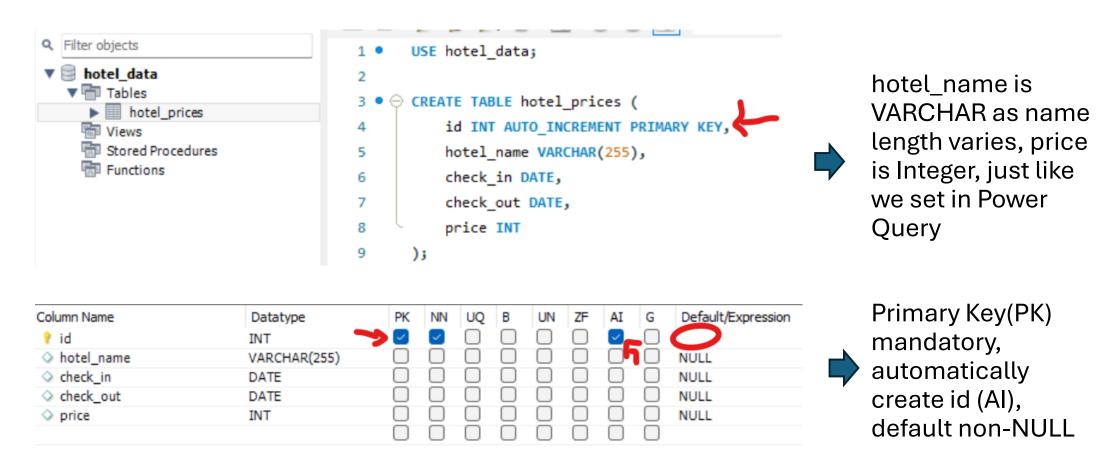
- The scrapped price data is shown as Text.
- We use Power Query Transform to change data type to Number, allowing future statistical analysis. Currency would also be an option.
- Once finished, we load the data.



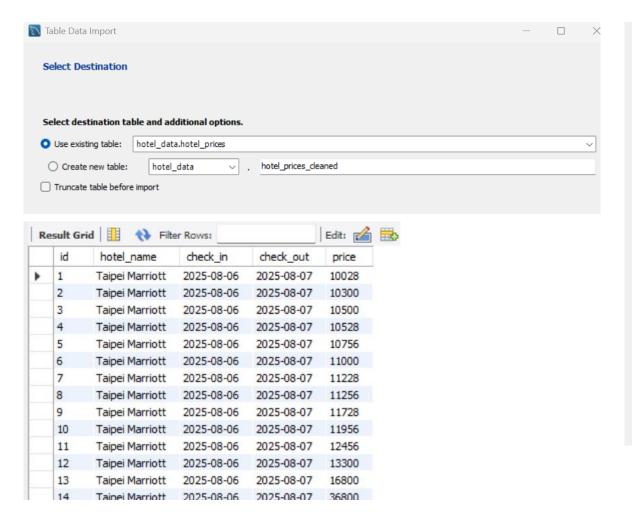


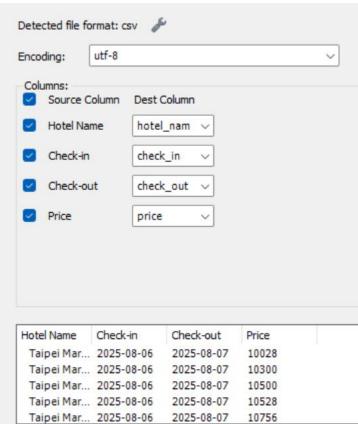


3A Create new MySQL Database and Table



3B Import Data from CSV into Table







3C Query to show nights above amount or in specific month

```
■ Imit to 1000 rows

1 • SELECT hotel_name, check_in, check_out, price
2 FROM hotel_prices
3 WHERE price > 15000;

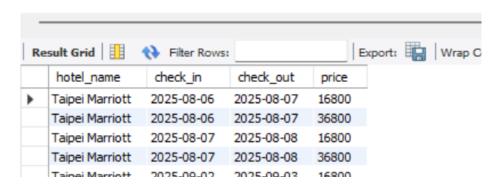
Limit to 1000 rows

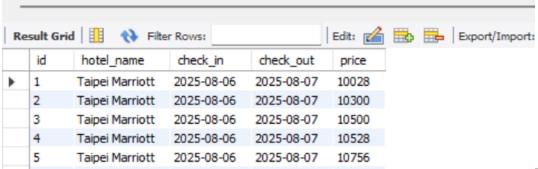
■ Imit to 1000 rows

■ SELECT *

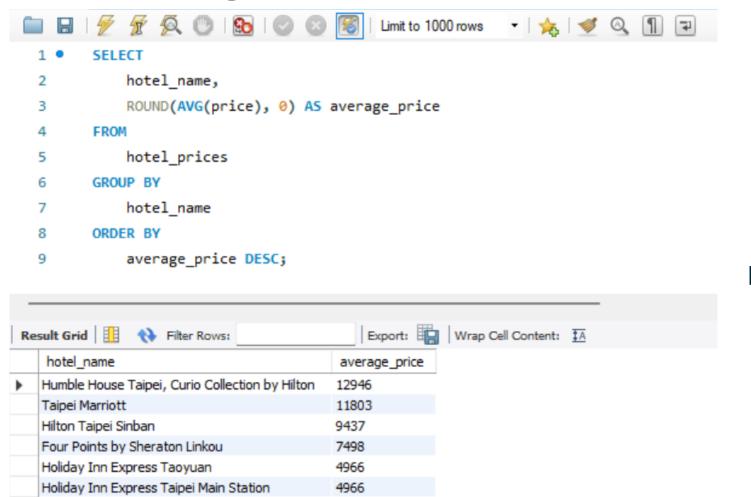
2 FROM hotel_prices
3 WHERE check_in BETWEEN '2025-08-01' AND '2025-08-31';

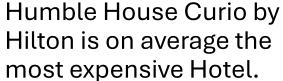
| Imit to 1000 rows | Imit to 1000 r
```





3D Average Price per Hotel

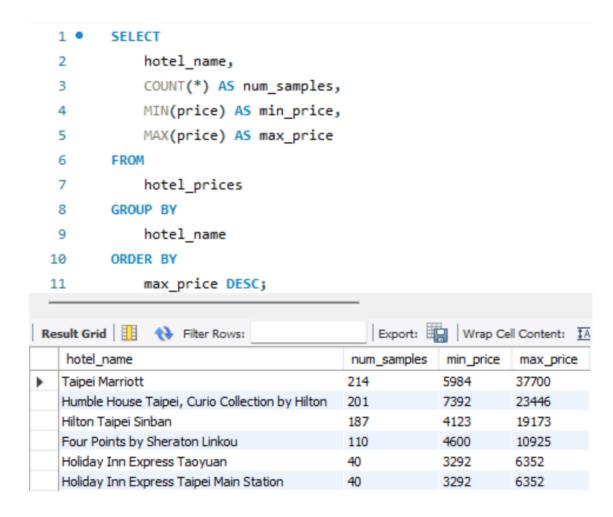




Holiday Inn Express are on average the cheapest Hotel(s)



3E Min Max Prices for each hotel





Allows us to decide if we get a good or a bad deal by booking at a given price point. e.g. if we book Taipei Marriott for 6100 NTD / night, that is a good deal.



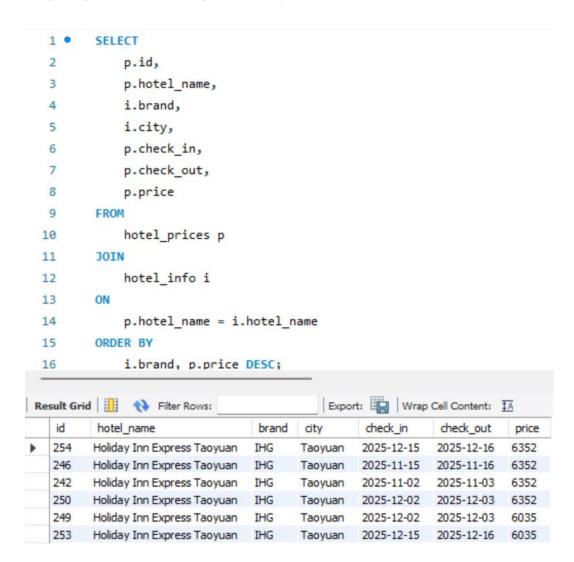
3F Create Hotel Info Table with SQL

```
USE hotel data;
2 • O CREATE TABLE hotel info (
           hotel id INT PRIMARY KEY,
           hotel_name VARCHAR(255),
4
5
           brand VARCHAR(100),
6
           city VARCHAR(100)
7
8
       INSERT INTO hotel info (hotel id, hotel name, brand, city) VALUES
       (1, 'Taipei Marriott', 'Marriott', 'Taipei'),
10
       (2, 'Holiday Inn Express Taoyuan', 'IHG', 'Taoyuan'),
11
       (3, 'Four Points by Sheraton Linkou', 'Marriott', 'Linkou'),
12
       (4, 'Humble House Taipei', 'Hilton', 'Taipei');
13
```



	hotel_id	hotel_name	brand	city
•	1	Taipei Marriott	Marriott	Taipei
	2	Holiday Inn Express Taoyuan	IHG	Taoyuan
	3	Four Points by Sheraton Linkou	Marriott	Linkou
	4	Humble House Taipei	Hilton	Taipei
	NULL	NULL	NULL	NULL

3G Inner Join



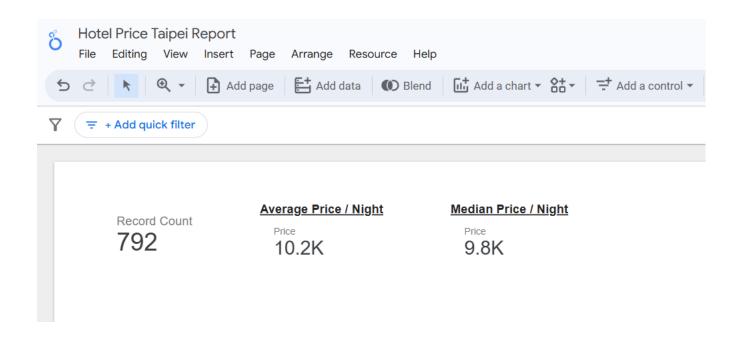


Performed an INNER JOIN to combine price data with hotel metadata (brand and city). This enriches analysis while ensuring only matched records are returned.

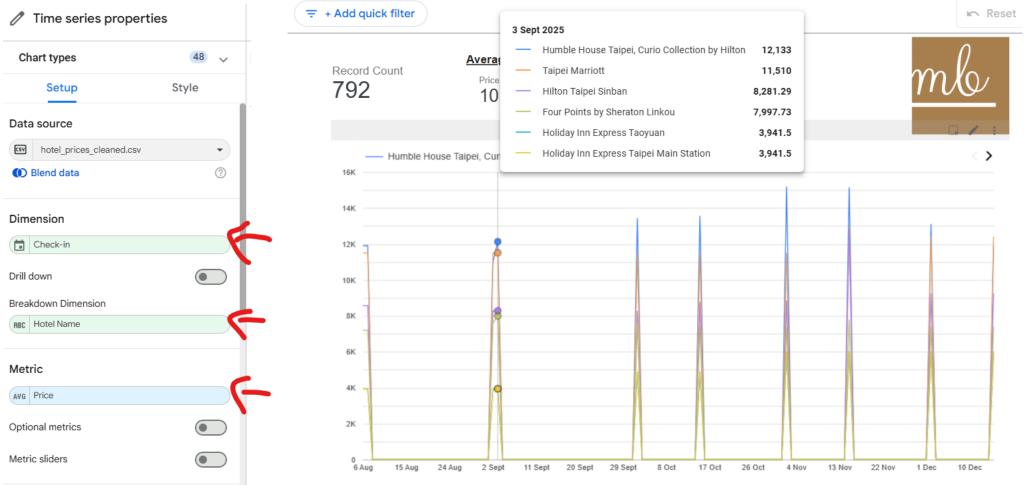


4A Load CSV to Looker Studio

- Added basic stat information, such as count of total number of rows, average price per night across all hotels and dates, as well as median price per night.
- Median < Mean \rightarrow The distribution is positively skewed.

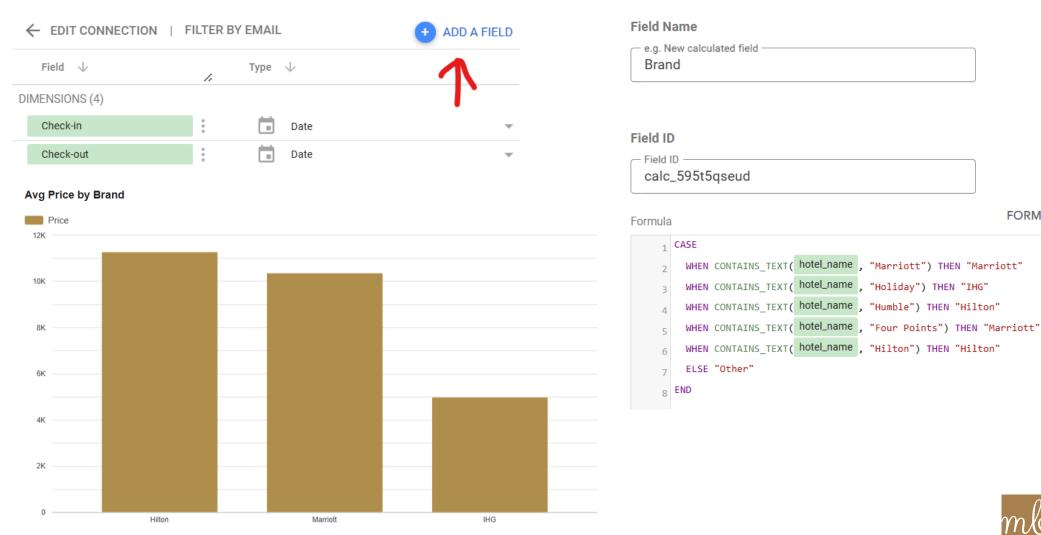


4B Time Series with Breakdown Dimension





4C Edit Data Source by Adding Brand



FORMAT F

5A Create price_prediction.py for ML

```
✓ Hotel Data♣ price_prediction.py♣ scraping_data.py
```

```
import pandas as pd
import numpy as np
import torch
import torch.nn as nn
import torch.optim as optim

data = pd.read_csv('hotel_prices_cleaned.csv')
```

```
Importing key tools: pandas, numpy, pytorch
```



5B Define Neural Network and Train it.

```
# Train Model
print("Training model...")
for epoch in range(200):
    optimizer.zero_grad()
    pred = model(X_tensor)
    loss = loss_fn(pred, y_tensor)
    loss.backward()
    optimizer.step()

if (epoch + 1) % 20 == 0:
    print(f"Epoch {epoch+1}, Loss: {loss.item():.2f}")

print("Training complete!\n")
```



Using Mean Squared Error loss and Adam optimizer to train for 200 epochs.



5C Create Price Prediction Function

```
def predict price(hotel name, checkin date, checkout date):
    hotel_idx = hotel_to_index.get(hotel_name)
    if hotel idx is None:
        raise ValueError(f"Unknown hotel name: {hotel name}")
    checkin_ord = pd.to_datetime(checkin_date).toordinal()
    checkout ord = pd.to datetime(checkout date).toordinal()
    features = np.array([[hotel idx, checkin ord, checkout ord]])
    features norm = (features - X mean) / X std
    features tensor = torch.tensor(features norm, dtype=torch.float32)
    prediction = model(features tensor).item()
    return prediction
```

Now it is possible to **predict** the price for any date, **even dates not in the original scrape.**



5D Prediction Examples

```
# Prediction Choice
new_price = predict_price(
   hotel_name="Taipei Marriott",
   checkin_date="2025-12-24",
   checkout_date="2025-12-25"
)
```



```
Epoch 160, Loss: 24877014.00
Epoch 180, Loss: 23904524.00
Epoch 200, Loss: 23226668.00
Training complete!

Predicted price for new date: TWD 15862
```

```
# Prediction Choice
new_price = predict_price(
    hotel_name="Hilton Taipei Sinban",
    checkin_date="2025-09-21",
    checkout_date="2025-09-22"
)
```



```
Epoch 120, Loss: 28956574.00
Epoch 140, Loss: 25161242.00
Epoch 160, Loss: 24007514.00
Epoch 180, Loss: 23246626.00
Epoch 200, Loss: 22690940.00
Training complete!

Predicted price for new date: TWD 5904
```

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