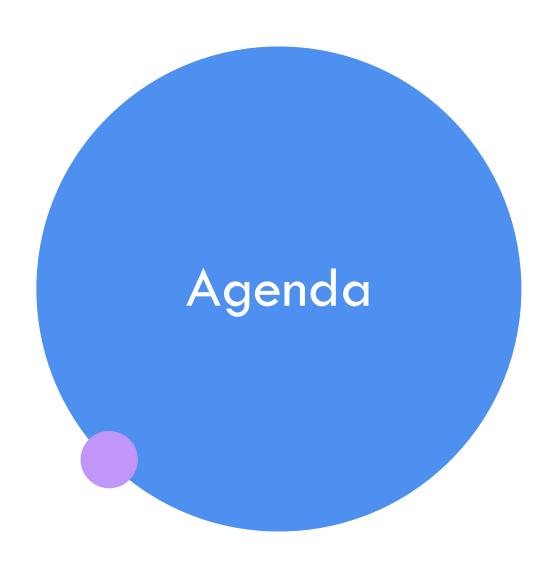


A Machine Learning Model for Diabetes

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- ⇒ Downloading the Pima Diabetes dataset from Kaggle
- ⇒ Preprocessing the data
- ⇒ Training a model using Hugging Face AutoTrain
- ⇒ Deploying the model with Gradio in Hugging Face Spaces

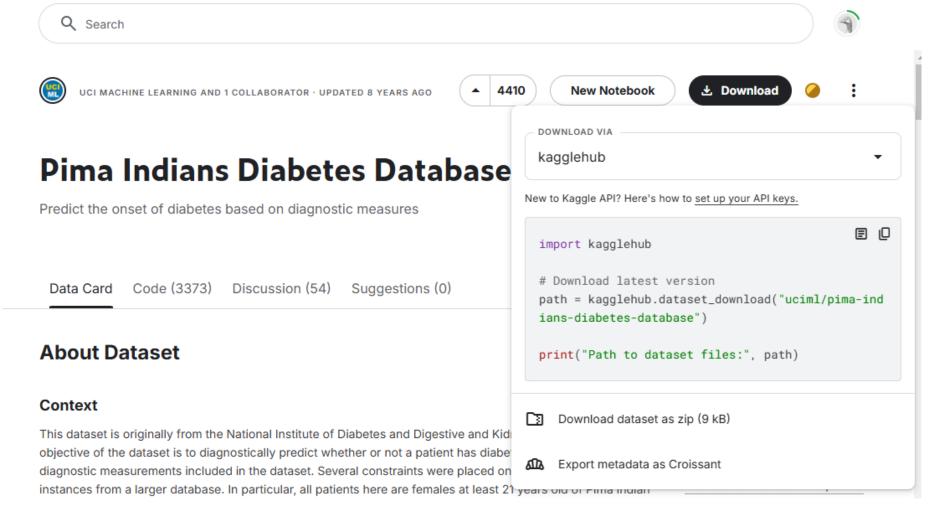
Pima Diabetes Dataset

- Predict the onset of diabetes based on diagnostic measures
- 768 patients are females at least 21 years old of Pima Indian heritage.
- Predictor variables includes the number of pregnancies the patient has had, their BMI, insulin level, age, and so on. (8 predictors)
- Target variable: Outcome (0,1)



https://www.kaggle.com/datasets/uciml/pima-indians-diabetes-database

Downloading the Dataset from Kaggle



https://www.kaggle.com/datasets/uciml/pima-indians-diabetes-database

Data Preprocessing and Visualization

 Replace zeros with NaN for columns where zero is invalid (Glucose, BloodPressure, SkinThickness, Insulin, BMI)

# Pregnancies = Number of times pregnant	# Glucose Plasma glucose concentration a 2 hours in an oral glucose tolerance test	# BloodPressure Diastolic blood pressure (mm Hg)	# SkinThickness = Triceps skin fold thickness (mm)	# Insulin = 2-Hour serum insulin (mu U/ml)	# BMI Body mass index in kg/(height in m)
0 17	0 199	0 122	0 99	0 846	0
6	148	72	35	0	33.6
1	85	66	29	0	26.6
8	183	64	0	0	23.3
1	00	44	22	0.4	20 1

Training with AutoTrain

- A no-code tool for training ML models
- Go to Hugging Face AutoTrain
- Upload the preprocessed dataset
- Choose the target column (diabetes outcome)
- Select model type and train
- Wait for training to complete!



Create powerful AI models without code

A new way to automatically train, evaluate and deploy state-of-the-art Machine Learning models.

Create new project

or read the documentation

https://huggingface.co/autotrain

Deploying on Hugging Face Spaces

- Deploy with Gradio: A Python library for building ML apps
- Create a new Hugging Face Space
- Select Gradio as the SDK
- Write a simple Gradio interface
- Upload the trained model
- Create requirements.txt
- Deploy!

Writing the Gradio App Code (app.py)

```
import gradio as gr
import numpy as np
import joblib
model = joblib.load("model4.joblib")
def myfunc(Pregnancies, Glucose, Blood Pressure, Skin Thickness, Insulin, BMI, Diabetes Pedigree Function, Age):
 data = np.array([[Pregnancies,Glucose,BloodPressure,SkinThickness,Insulin,BMI,DiabetesPedigreeFunction,Age]])
 p = model.predict_proba(data)
 return p[0][1]
demo = gr.Interface(
 fn=myfunc,
 inputs=[
   gr.Number(label="Pregnancies"),
   gr.Number(label="Glucose"),
   gr.Number(label="Blood Pressure"),
   gr.Number(label="Skin Thickness"),
   gr.Number(label="Insulin"),
   gr.Number(label="BMI"),
   gr.Number(label="Diabetes Pedigree Function"),
   gr.Number(label="Age")],
 outputs=gr.Textbox(label="Probability"),
 title="Diabetes prediction App",
 description="Enter patient info to predict diabetes risk")
demo.launch()
```

requirements.txt

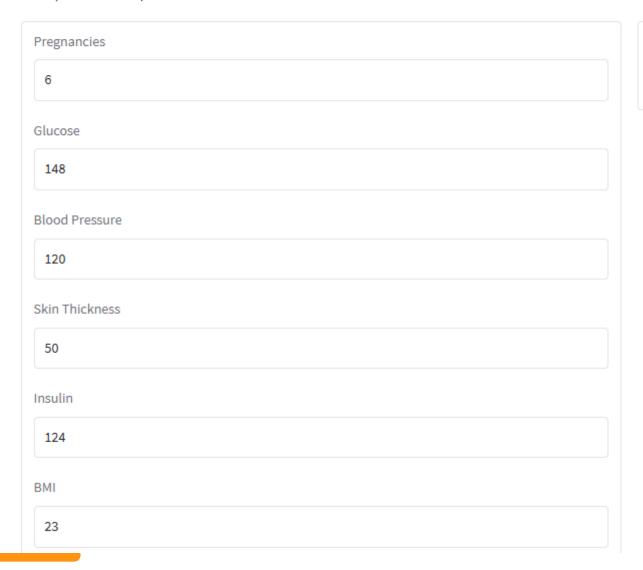
joblib

numpy

scikit-learn

Diabetes prediction App

Enter patient info to predict diabetes risk



Probability		
0.541666666666666		

Testing and Sharing Your Model

- Use the Gradio web UI to test the deployed model
- Enter patient data and get predictions
- Share the Hugging Face Space link with others

