

Frontend: Automated Review Rating System

Frontend Overview: Streamlit Application

The frontend is implemented using **Streamlit**, providing an interactive web interface for predicting product review ratings. Users can input any textual review, and the app compares predictions from two trained machine learning models:

- **Model A:** Trained on a balanced dataset
- **Model B:** Trained on an imbalanced dataset

Predictions are displayed side by side along with confidence scores, enabling easy comparison of model behavior.

Key Components

1. Model & Vectorizer Loading:

Pre-trained Logistic Regression models (`Model_Balanced.pkl` and `Model_Imbalanced.pkl`) and their corresponding TF-IDF vectorizers (`tfidf_Balanced.pkl` and `tfidf_Imbalanced.pkl`) are loaded using `joblib`. Models are ready to perform inference on new reviews.

2. Text Preprocessing:

Function `preprocess_review()` handles text cleaning:

- Converts text to lowercase.
- Removes URLs, HTML tags, special characters, and extra whitespace.
- Lemmatizes words using `SpaCy`.
- Removes stopwords to retain meaningful words.

This ensures that input text aligns with the format used during model training.

3. Prediction Function:

Function `predict_review()` takes a raw review, preprocesses it, applies TF-IDF vectorization, and generates the predicted rating and confidence score from the selected model.

4. User Interface:

- Page title and subtitle with custom CSS styling.
- Input area for typing or pasting a review.
- Predict button triggers rating prediction for both models.
- Results displayed in cards with distinct borders and color themes:
 - Balanced Model: Green border
 - Imbalanced Model: Orange border
- Confidence is visualized using Streamlit progress bars.

5. Additional Features:

- Spinner animation provides feedback while processing.
- Footer credits the use of Streamlit and SpaCy.

Workflow

1. User enters a review in the text area.
2. Upon clicking Predict Rating button:
 - The review is preprocessed.
 - Predictions are generated using both balanced and imbalanced models.
 - Results are displayed side by side:
 - Predicted rating (1–5 stars)
 - Model confidence
3. Users can visually compare how different training data distributions affect predictions.

Technologies Used

- **Streamlit:** Frontend web application framework for Python.
- **SpaCy:** NLP preprocessing (lemmatization, stopword removal).
- **Joblib:** Model and vectorizer serialization/deserialization.
- **Python Libraries:** `re` for regex cleaning, `numpy` for numeric operations, `time` for spinner simulation.

Python Code Used

```
import streamlit as st
import joblib
import re
import spacy
import numpy as np
import time

# --- Load NLP Model ---
nlp = spacy.load("en_core_web_sm")

# --- Load Models and Vectorizers ---
model_dir = "../models"
model_A = joblib.load(f"{model_dir}/Model_Balanced.pkl")
model_B = joblib.load(f"{model_dir}/Model_Imbalanced.pkl")
vectorizer_A = joblib.load(f"{model_dir}/tfidf_Balanced.pkl")
vectorizer_B = joblib.load(f"{model_dir}/tfidf_Imbalanced.pkl")

# --- Helper Functions ---
```

```

def preprocess_review(text):
    """Clean, remove stopwords, and lemmatize text."""
    text = text.lower()
    text = re.sub(r"http\S+|www\S+|https\S+", "", text)
    text = re.sub(r"<.*?>", "", text)
    text = re.sub(r"^[a-zA-Z0-9\s]", "", text)
    text = re.sub(r"\s+", " ", text).strip()
    doc = nlp(text)
    return " ".join([token.lemma_ for token in doc if token.text not in
        nlp.Defaults.stop_words])

def predict_review(review, model, vectorizer):
    processed = preprocess_review(review)
    tfidf_review = vectorizer.transform([processed])
    prediction = model.predict(tfidf_review)[0]
    confidence = model.predict_proba(tfidf_review).max()
    return prediction, confidence

# --- Streamlit Page Setup ---
st.set_page_config(page_title="Review Rating Predictor", page_icon="",
    layout="wide")

# --- Custom CSS for Styling ---
st.markdown("""
<style>
    body {
        background-color: #f8f9fa;
    }
    .main-title {
        text-align: center;
        color: #2E8B57;
        font-size: 40px;
        font-weight: 700;
        margin-bottom: 10px;
    }
    .sub-title {
        text-align: center;
        color: #555;
        font-size: 18px;
        margin-bottom: 40px;
    }
    .card {
        padding: 20px;
        border-radius: 15px;
        box-shadow: 0 4px 10px rgba(0,0,0,0.1);
        background-color: white;
        transition: transform 0.3s ease;
    }
    .card:hover {
        transform: scale(1.03);
    }
</style>
""")

```

```

        .balanced {
            border-left: 8px solid #4CAF50;
        }
        .imbalanced {
            border-left: 8px solid #FF9800;
        }
        .confidence-bar {
            height: 10px;
            border-radius: 10px;
            margin-top: 5px;
        }
        .footer {
            text-align:center;
            color:gray;
            margin-top:50px;
            font-size:14px;
        }
    </style>
""" , unsafe_allow_html=True)

# --- Title Section ---
st.markdown("<div class='main-title'> Review Rating Predictor</div>",
            unsafe_allow_html=True)
st.markdown("<div class='sub-title'>Compare predictions from Balanced
            vs. Imbalanced Models</div>", unsafe_allow_html=True)

# --- Input Section ---
review_input = st.text_area(" Enter a product review below:",
                             placeholder="Type your review here...")

if st.button(" Predict Rating"):
    if review_input.strip() == "":
        st.warning("Please enter a review first.")
    else:
        with st.spinner(". Loading Analyzing review....."):
            time.sleep(1.2)
            pred_A, conf_A = predict_review(review_input, model_A,
                                             vectorizer_A)
            pred_B, conf_B = predict_review(review_input, model_B,
                                             vectorizer_B)

        st.success(" Prediction completed successfully")

# --- Display Results Side by Side ---
col1, col2 = st.columns(2, gap="large")

with col1:
    st.markdown("<div class='card balanced'>", unsafe_allow_html=
                True)
    st.markdown("### Model A (Balanced Data)")
    st.write(f"***Predicted Rating:** {pred_A}")

```

```

st.progress(float(conf_A))
st.write(f"**Confidence:** {conf_A:.2f}")
st.markdown("</div>", unsafe_allow_html=True)

with col2:
    st.markdown("<div class='card imbalanced'>", unsafe_allow_html=True)
    st.markdown("### Model B (Imbalanced Data)")
    st.write(f"**Predicted Rating:** {pred_B}")
    st.progress(float(conf_B))
    st.write(f"**Confidence:** {conf_B:.2f}")
    st.markdown("</div>", unsafe_allow_html=True)

st.markdown("<div class='footer'>Made with using Streamlit & SpaCy</div>", unsafe_allow_html=True)

```


Outputs:







Review Rating Predictor

Compare predictions from Balanced vs. Imbalanced Models

 Enter a product review below:

free read know not bother time waste reading

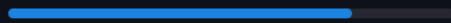
  Predict Rating

Prediction completed successfully



Model A (Balanced Data)

Predicted Rating: ★ 1

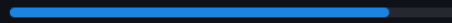


Confidence: 0.78




Model B (Imbalanced Data)

Predicted Rating: ★ 1



Confidence: 0.85

Made with  using Streamlit & SpaCy