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        "from sklearn.model_selection import train_test_split\n",

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        "from datasets import Dataset\n",
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        "train_df, test_df = train_test_split(df, test_size=0.2,\nstratify=df['sentiment'], random_state=42) # Split data\n",
        "train_df.to_csv(\"/content/drive/MyDrive/LLM assignment\n2/train_data.csv\", index=False)\n",
        "test_df.to_csv(\"/content/drive/MyDrive/LLM assignment\n2/test_data.csv\", index=False)\n",
        "train_df, test_df = pd.read_csv(\"/content/drive/MyDrive/LLM\nassignment 2/train_data.csv\"), pd.read_csv(\"/content/drive/MyDrive/LLM\nassignment 2/test_data.csv\")\n",
        "print(f\"Train: {len(train_df)}, Test: {len(test_df)}\\n\\nTrain\ndist:\\n{train_df['sentiment'].value_counts()}\\n\\nTest\ndist:\\n{test_df['sentiment'].value_counts()}\\n\\nSamples:\\n{train_df[['\n'review','sentiment']].head(3)}\\n\")\n"
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        "
review
sentiment\n",
"0  I caught this little gem totally by accident b...
positive\n",
"1  I can't believe that I let myself into this mo...
negative\n",
"2  *spoiler alert!* it just gets to me the nerve ...
negative\n"
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        "print(\"Class labels:\", train_df['sentiment'].unique())\n",
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        "print(\"\\nTraining set label counts:\\n\\n\",
train_df['sentiment'].value_counts())\n",
        "print(\"\\nTest set label counts:\\n\\n\",
test_df['sentiment'].value_counts())\n",
        "class_labels = sorted(train_df['sentiment'].unique())\n",
        "print(\"Class label list:\", class_labels)\n"
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        "# Bar plot with counts\n",
        "plt.figure(figsize=(6,4))\n",
        "ax = sns.countplot(x='sentiment', data=df, palette='Set2')\n",
        "plt.title(\"Class Distribution (Full Dataset)\")\n",
        "for p in ax.patches:\n",
        "    ax.annotate(f'{p.get_height()}', (p.get_x()+0.3,
p.get_height()+500))\n",
        "plt.legend(title=\"Sentiment\",
labels=df['sentiment'].unique())\n",
        "plt.show()\n",
        "\n",
        "# Review length histogram\n",
        "plt.figure(figsize=(6,4))\n",
        "sns.histplot(df['length'], bins=30, kde=False,
color='skyblue')\n",
        "plt.title(\"Review Lengths (Full Dataset)\")\n",
        "plt.xlabel(\"Number of Characters\")\n",
        "plt.ylabel(\"Frequency\")\n",
        "plt.show()\n",
        "\n",
        "# Print one example per class\n",
        "for label in df['sentiment'].unique():\n",
        "    print(f\"\\n{label} sample:\\n\",
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" One of the other reviewers has mentioned that after watching just 1 Oz episode you'll be hooked. They are right, as this is exactly what happened with me.

The first thing that struck me about Oz was its brutality and unflinching scenes of violence, which set

in right from the word GO. Trust me, this is not a show for the faint hearted or timid. This show pulls no punches with regards to drugs, sex or violence. Its is hardcore, in the classic use of the word.

It is called OZ as that is the nickname given to the Oswald Maximum Security State Penitentiary. It focuses mainly on Emerald City, an experimental section of the prison where all the cells have glass fronts and face inwards, so privacy is not high on the agenda. Em City is home to many..Aryans, Muslims, gangstas, Latinos, Christians, Italians, Irish and more....so scuffles, death stares, dodgy dealings and shady agreements are never far away.

I would say the main appeal of the show is due to the fact that it goes where other shows wouldn't dare. Forget pretty pictures painted for mainstream audiences, forget charm, forget romance...OZ doesn't mess around. The first episode I ever saw struck me as so nasty it was surreal, I couldn't say I was ready for it, but as I watched more, I developed a taste for Oz, and got accustomed to the high levels of graphic violence. Not just violence, but injustice (crooked guards who'll be sold out for a nickel, inmates who'll kill on order and get away with it, well mannered, middle class inmates being turned into prison bitches due to their lack of street skills or prison experience) Watching Oz, you may become comfortable with what is uncomfortable viewing....thats if you can get in touch with your darker side.\n",

"\n",

"negative sample:\n",

" Basically there's a family where a little boy (Jake) thinks there's a zombie in his closet & his parents are fighting all the time.

This movie is slower than a soap opera... and suddenly, Jake decides to become Rambo and kill the zombie.

OK, first of all when you're going to make a film you must Decide if its a thriller or a drama! As a drama the movie is watchable. Parents are divorcing & arguing like in real life. And then we have Jake with his closet which totally ruins all the film! I expected to see a BOOGEYMAN similar movie, and instead i watched a drama with some meaningless thriller spots.

3 out of 10 just for the well playing parents & descent dialogs. As for the shots with Jake: just ignore them.\n"

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"from nltk.corpus import stopwords\n",

"\n",

"stop_words = set(stopwords.words('english'))\n",

"\n",

```

def preprocess_text(text):\n",
    text = text.lower() # Lower case\n",
    text = re.sub(r'<.*?>', '', text) # Remove HTML
tags\n",
    text = re.sub(r'^a-z\\s]', '', text) # Remove
punctuation & numbers\n",
    text = re.sub(r'\\s+', ' ', text).strip() # Remove extra
whitespace\n",
    text = ' '.join(word for word in text.split() if word not in
stop_words) # Remove stopwords\n",
    return text\n",
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    "\n",
    "print(df[['review', 'clean_review']].head(3))\n"
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            "1 A wonderful little production. <br /><br />The... \n",
            "2 I thought this was a wonderful way to spend ti... \n",
            "\n",
            "                                clean_review \n",
            "0 one reviewers mentioned watching oz episode yo... \n",
            "1 wonderful little production filming technique ... \n",
            "2 thought wonderful way spend time hot summer we... \n"
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        "# 1. Preprocess train and test reviews\n",
        "train_df['clean_review'] =
train_df['review'].apply(preprocess_text)\n",

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        "test_df['clean_review'] =
test_df['review'].apply(preprocess_text)\n",
        "\n",
        "# 2. Extract cleaned texts as lists\n",
        "train_texts = train_df['clean_review'].tolist()\n",
        "val_texts = test_df['clean_review'].tolist()\n"
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        "tokenizer = BertTokenizer.from_pretrained('bert-base-
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        "def tokenize_function(texts):\n",
        "    return tokenizer(texts, padding=\"max_length\",
truncation=True, max_length=256)\n",
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        "train_encodings = tokenize_function(train_texts)\n",
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secret in your Google Colab and restart your session.\n",
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notebooks.\n",
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optional to access public models or datasets.\n",
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    "    'labels': train_labels\n",
    "})\n",
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    "val_dataset = Dataset.from_dict({\n",
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"You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.\n"

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    "def compute_metrics(eval_pred):\n",
    "    logits, labels = eval_pred\n",
    "    predictions = np.argmax(logits, axis=-1)\n",
    "    acc = accuracy_score(labels, predictions)\n",
    "    f1 = f1_score(labels, predictions)\n",
    "    return {'accuracy': acc, 'f1': f1}\n",
    "\n",
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    "from datasets import Dataset\n",
    "\n",
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    "    'labels': test_labels\n",
    "})\n",
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    "pred_logits = predictions_output.predictions\n",
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    "\n",
    "# Optionally, print accuracy and F1 on test set\n",
    "from sklearn.metrics import accuracy_score, f1_score\n",
    "\n",
    "acc = accuracy_score(test_labels, pred_labels)\n",
    "f1 = f1_score(test_labels, pred_labels)\n",
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