```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.pipeline import Pipeline
from sklearn.metrics import accuracy_score, f1_score, classification_report
import joblib
```

```
!pip install --upgrade --force-reinstall transformers==4.44.2
!pip install --upgrade --force-reinstall accelerate datasets
```

```
import pandas as pd
from google.colab import files

# This opens a file picker in Colab → select your local CSV
uploaded = files.upload()

# Get the first uploaded filename
filename = list(uploaded.keys())[0]

# Read into pandas
df = pd.read_csv(filename)

# Quick sanity check
print("Rows:", len(df))
print("Columns:", df.columns.tolist())
print(df.head())
```

Choose files data.csv

data.csv(text/csv) - 79825 bytes, last modified: 23/09/2025 - 100% done

```
Saving data.csv to data.csv
Rows: 2129
Columns: ['reply', 'label']
                                    reply     label
0                  Can we discuss pricing??   NEUTRAL
1  Im excited to explore this further, plz send c...  POSITIVE
2                 We not looking for new solutions.  negative
3               Could u clarify features included?   neutral
4          lets,, schedule a meeting to dive deeper  positive
```

```
-none-any.whl.metadata (57 kB)
rom huggingface-hub<1.0,>=0.23.2->transformers==4.44.2)
  Using cached fsspec-2025.9.0-py3-none-any.whl.metadata (10 kB)
Collecting typing-extensions>=3.7.4.3 (from huggingface-hub<1.0,>=0.23.2->transformers==4.44.2)
  Using cached typing_extensions-4.15.0-py3-none-any.whl.metadata (3.3 kB)
Traceback (most recent call last):
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/cli/base_command.py", line 179, in exc_logging_wrapp
    status = run_func(*args)
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/cli/req_command.py", line 67, in wrapper
    return func(self, options, args)
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/commands/install.py", line 377, in run
```

```
df = df.rename(columns={"reply": "text", "label": "label"})
df["text"] = df["text"].astype(str).str.strip().str.replace(r"\s+", " ", regex=True)
df["label"] = df["label"].str.lower().str.strip()

# Quick check
print("Rows:", len(df))
print("Columns:", df.columns.tolist())
print("\nLabel distribution:\n", df["label"].value_counts())
df.head()
```

```
Rows: 2129
Columns: ['text', 'label']

Label distribution:
 label
positive    710
negative    710
neutral     709
Name: count, dtype: int64
```

	text	label
0	Can we discuss pricing??	neutral
1	Im excited to explore this further, plz send c...	positive
2	We not looking for new solutions.	negative
3	Could u clarify features included?	neutral
4	lets,, schedule a meeting to dive deeper	positive

```
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/resolvelib/resolvers.py", line 239, in _attempt_to_pin
    criteria = self._get_updated_criteria(candidate)
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/resolvelib/resolvers.py", line 230, in _get_updated_cr
    self._add_to_criteria(criteria, requirement, parent=candidate)
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/resolvelib/resolvers.py", line 173, in _add_to_criteri
    if not criterion.candidates:
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/resolvelib/structs.py", line 156, in __bool__
    return bool(self._sequence)
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/resolution/resolvelib/found_candidates.py", line 174
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/resolution/resolvelib/found_candidates.py", line 162
    in self._incompatible_ids)
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/resolution/resolvelib/found_candidates.py", line 49,
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/resolution/resolvelib/factory.py", line 300, in iter
    result = self._finder.find_best_candidate(
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/index/package_finder.py", line 884, in find_best_can
```

Next steps: Generate code with df | New interactive sheet

```
train_df, test_df = train_test_split(
    df, test_size=0.2, stratify=df['label'], random_state=42
)

print("Train size:", len(train_df), "Test size:", len(test_df))
```

```
Train size: 1703 Test size: 426
```

```
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/index/package_finder.py", line 792, in process_proje
    package_links = self.evaluate_links(
```

```
pipeline = Pipeline([
    ("tfidf", TfidfVectorizer(
        lowercase=True,
        strip_accents="unicode",
        stop_words="english",
        ngram_range=(1, 2),
        max_features=20000
    )),
    ("clf", LogisticRegression(
```

```
        max_iter=200,
        solver="liblinear",
        random_state=42
    ))
])

param_grid = {"clf__C": [0.25, 0.5, 1.0, 2.0, 4.0]}
grid = GridSearchCV(
    pipeline, param_grid=param_grid,
    cv=5, n_jobs=-1, scoring="f1_macro", verbose=1
)

grid.fit(train_df['text'], train_df['label'])
```

```
Fitting 5 folds for each of 5 candidates, totalling 25 fits
▸ GridSearchCV
▸ best_estimator_: Pipeline
    ▸ TfidfVectorizer
    ▸ LogisticRegression
```

```
/dist-packages/pip/_internal/cli/base_command.py", line 100, in main
/dist-packages/pip/_internal/cli/base_command.py", line 232, in _main
/dist-packages/pip/_internal/cli/base_command.py", line 215, in exc_logging_wrapp
ncelled by user")
ng/__init__.py", line 1586, in critical
, **kwargs)
  File "/usr/lib/python3.12/logging/__init__.py", line 1684, in _log
```

```
y_pred = grid.predict(test_df['text'])
print("Accuracy:", accuracy_score(test_df['label'], y_pred))
print("Macro F1:", f1_score(test_df['label'], y_pred, average="macro"))
print("\nClassification Report:\n", classification_report(test_df['label'], y_pred))
```

```
Accuracy: 0.9882629107981221
Macro F1: 0.9882781717888101

Classification Report:
              precision    recall  f1-score   support

    negative       0.99      0.98      0.99       142
     neutral       1.00      0.99      0.99       142
    positive       0.97      1.00      0.99       142

    accuracy                           0.99       426
   macro avg       0.99      0.99      0.99       426
weighted avg       0.99      0.99      0.99       426
```

```
  File "/usr/local/lib/python3.12/dist-packages/pip/_internal/utils/logging.py", line 177, in emit
    self.console.print(renderable, overflow="ignore", crop=False, style=style)
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/rich/console.py", line 1674, in print
    renderables = self._collect_renderables(
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/rich/console.py", line 1553, in _collect_renderables
    check_text()
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/rich/console.py", line 1531, in check_text
    append(sep_text.join(text))
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/rich/text.py", line 803, in join
    for text in iter_text():
  File "/usr/local/lib/python3.12/dist-packages/pip/_vendor/rich/text.py", line 790, in iter_text
    for last, line in loop_last(lines):
```

```
import joblib
joblib.dump(grid.best_estimator_, "baseline_model.joblib")
```

```
['baseline_model.joblib']
KeyboardInterrupt                         Traceback (most recent call last)
```

```
from google.colab import files
files.download("baseline_model.joblib")
```

```
11 frames
/usr/lib/python3.12/pathlib.py in stat(self, follow_symlinks)
    838         os.stat() does.
```

```
!pip install -q transformers datasets accelerate evaluate
```

```
    842     def lstat(self):
```

```
!pip install -q --upgrade transformers
```

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split

import torch
from datasets import Dataset
from transformers import (
    AutoTokenizer,
    AutoModelForSequenceClassification,
    TrainingArguments,
    Trainer,
    DataCollatorWithPadding,
)
```

```
# Ensure columns are "text" and "label" with lowercase labels
df["label"] = df["label"].str.lower().str.strip()

# Train / validation split
train_df, val_df = train_test_split(
    df, test_size=0.2, stratify=df["label"], random_state=42
)

# Hugging Face dataset objects
train_ds = Dataset.from_pandas(train_df)
val_ds = Dataset.from_pandas(val_df)

# Label mapping
labels = ["negative", "neutral", "positive"]
label2id = {l: i for i, l in enumerate(labels)}
id2label = {i: l for l, i in label2id.items()}

def encode_labels(example):
    example["labels"] = label2id[example["label"]]
    return example

train_ds = train_ds.map(encode_labels)
val_ds = val_ds.map(encode_labels)
```

```
Map: 100%  1703/1703 [00:00<00:00, 4436.51 examples/s]
Map: 100%  426/426 [00:00<00:00, 4525.41 examples/s]
```

```
model_name = "distilbert-base-uncased"
tokenizer = AutoTokenizer.from_pretrained(model_name)

def tokenize(examples):
    return tokenizer(examples["text"], truncation=True, padding=False, max_length=256)

train_ds = train_ds.map(tokenize, batched=True)
val_ds = val_ds.map(tokenize, batched=True)

data_collator = DataCollatorWithPadding(tokenizer=tokenizer)
```

```
/usr/local/lib/python3.12/dist-packages/transformers/tokenization_utils_base.py:1601: FutureWarning: `clean_up_toker
  warnings.warn(
Map: 100%  1703/1703 [00:00<00:00, 7200.24 examples/s]
Map: 100%  426/426 [00:00<00:00, 4602.28 examples/s]
```

```
model = AutoModelForSequenceClassification.from_pretrained(
    model_name,
    num_labels=len(labels),
    id2label=id2label,
    label2id=label2id,
)
```

```
Some weights of DistilBertForSequenceClassification were not initialized from the model checkpoint at distilbert-bas
You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.
```

```
from sklearn.metrics import accuracy_score, f1_score

def compute_metrics(eval_pred):
    logits, labels = eval_pred
    preds = np.argmax(logits, axis=-1)
    acc = accuracy_score(labels, preds)
    f1 = f1_score(labels, preds, average="macro")
    return {"accuracy": acc, "f1": f1}
```

```
from transformers import TrainingArguments
training_args = TrainingArguments(
    output_dir="./distilbert-reply-clf",
    evaluation_strategy="epoch",   # or "steps"
    save_strategy="epoch",
    learning_rate=2e-5,
    per_device_train_batch_size=16,
    per_device_eval_batch_size=16,
```

```
    num_train_epochs=4,
    weight_decay=0.01,
    load_best_model_at_end=True,
    metric_for_best_model="f1",
    greater_is_better=True,
    logging_dir="./logs",
    logging_steps=20,
)
```

```
/usr/local/lib/python3.12/dist-packages/transformers/training_args.py:1525: FutureWarning: `evaluation_strategy` is
  warnings.warn(
```

```
!pip install --upgrade transformers
```

```
import transformers
print(transformers.__version__)
```

```
from datasets import Dataset
from sklearn.model_selection import train_test_split

# 1. Normalize labels
df["label"] = df["label"].str.lower().str.strip()

# 2. Train/validation split
train_df, val_df = train_test_split(
    df, test_size=0.2, stratify=df["label"], random_state=42
)

# 3. Define mappings
labels = ["negative", "neutral", "positive"]
label2id = {l: i for i, l in enumerate(labels)}
id2label = {i: l for l, i in label2id.items()}

# 4. Convert pandas → Dataset, keeping only needed columns
train_ds = Dataset.from_pandas(train_df[["text", "label"]].reset_index(drop=True))
val_ds = Dataset.from_pandas(val_df[["text", "label"]].reset_index(drop=True))

# 5. Encode labels → integers
def encode_labels(example):
    return {"labels": label2id[example["label"]]}

train_ds = train_ds.map(encode_labels)
val_ds = val_ds.map(encode_labels)

# 6. Tokenizer
from transformers import AutoTokenizer
model_name = "distilbert-base-uncased"
tokenizer = AutoTokenizer.from_pretrained(model_name)

def tokenize(batch):
    return tokenizer(batch["text"], truncation=True, padding=True, max_length=256)

train_ds = train_ds.map(tokenize, batched=True)
val_ds = val_ds.map(tokenize, batched=True)

# 7. Remove original string label column
train_ds = train_ds.remove_columns(["label"])
val_ds = val_ds.remove_columns(["label"])

# 8. Verify
print(train_ds[0])
```

Map: 100% 1703/1703 [00:00<00:00, 9920.73 examples/s]

Map: 100% 426/426 [00:00<00:00, 4857.16 examples/s]

```
/usr/local/lib/python3.12/dist-packages/transformers/tokenization_utils_base.py:1601: FutureWarning: `clean_up_toker
  warnings.warn(
```

Map: 100% 1703/1703 [00:00<00:00, 6723.83 examples/s]

Map: 100% 426/426 [00:00<00:00, 3659.75 examples/s]

```
{'text': 'Please share the details, I’m interested.', 'labels': 2, 'input_ids': [101, 3531, 3745, 1996, 4751, 1010,
```

```
from transformers import AutoModelForSequenceClassification

model = AutoModelForSequenceClassification.from_pretrained(
    "distilbert-base-uncased",
```

```
    num_labels=len(labels),
    id2label=id2label,
    label2id=label2id,
)
```

```
Some weights of DistilBertForSequenceClassification were not initialized from the model checkpoint at distilbert-bas
You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.
```

```
from transformers import TrainingArguments

training_args = TrainingArguments(
    output_dir="./distilbert-reply-clf",
    eval_strategy="epoch",        # use "eval_strategy" (new name)
    save_strategy="epoch",
    learning_rate=2e-5,
    per_device_train_batch_size=16,
    per_device_eval_batch_size=16,
    num_train_epochs=4,
    weight_decay=0.01,
    load_best_model_at_end=True,
    metric_for_best_model="f1",
    greater_is_better=True,
    logging_dir="./logs",
    logging_steps=20,
    report_to="none",             # ✅ turn off wandb prompt
)
```

```
from sklearn.metrics import accuracy_score, f1_score

def compute_metrics(eval_pred):
    logits, labels = eval_pred
    preds = logits.argmax(axis=-1)
    acc = accuracy_score(labels, preds)
    f1 = f1_score(labels, preds, average="macro")
    return {"accuracy": acc, "f1": f1}
```

```
from transformers import Trainer, DataCollatorWithPadding

data_collator = DataCollatorWithPadding(tokenizer=tokenizer)

trainer = Trainer(
    model=model,
    args=training_args,
    train_dataset=train_ds,
    eval_dataset=val_ds,
    tokenizer=tokenizer,
    data_collator=data_collator,
    compute_metrics=compute_metrics,
)
```

```
trainer.train()
```

```
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:666: UserWarning: 'pin_memory' argument is se
  warnings.warn(warn_msg)
```

[428/428 18:11, Epoch 4/4]

Epoch	Training Loss	Validation Loss	Accuracy	F1
1	0.024900	0.025807	0.995305	0.995305
2	0.006000	0.004189	1.000000	1.000000
3	0.003800	0.002724	1.000000	1.000000
4	0.003100	0.002383	1.000000	1.000000

```
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:666: UserWarning: 'pin_memory' argument is se
  warnings.warn(warn_msg)
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:666: UserWarning: 'pin_memory' argument is se
  warnings.warn(warn_msg)
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:666: UserWarning: 'pin_memory' argument is se
  warnings.warn(warn_msg)
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:666: UserWarning: 'pin_memory' argument is se
  warnings.warn(warn_msg)
TrainOutput(global_step=428, training_loss=0.07845190939467366, metrics={'train_runtime': 1104.7997,
'train_samples_per_second': 6.166, 'train_steps_per_second': 0.387, 'total_flos': 24674562830736.0, 'train_loss':
```