

In [1]:

```
!pip install transformers torch pandas scikit-learn tqdm rouge-score -q
```

1. Load and Prepare Data

In [2]:

```
import os
import json
import time
import torch
import pandas as pd
import numpy as np
from tqdm.auto import tqdm
from sklearn.model_selection import train_test_split
from torch.utils.data import Dataset, DataLoader
from torch.optim import AdamW
from transformers import BartForConditionalGeneration, BartTokenizer, get_linear_schedule

device = torch.device('mps' if torch.backends.mps.is_available() else 'cpu')
print(f'Using device: {device}')

df = pd.read_csv('AppetIte_Dataset_v1.csv')
print(f'Total records: {len(df)}')
print(f'Categories: {df["category"].value_counts().to_dict()}')
```

Using device: mps

Total records: 13501

Categories: {'Indulgent': 10685, 'Healthy': 1437, 'Quick Meals': 1308, 'Family-Friendly': 71}

In [3]:

```
device = torch.device('mps' if torch.backends.mps.is_available() else 'cpu')
print(f'Using device: {device}')
```

Using device: mps

In [4]:

```
df = pd.read_csv('AppetIte_Dataset_v1.csv')
print(f'Total records: {len(df)}')
print(f'Categories: {df["category"].value_counts().to_dict()}')
```

Total records: 13501

Categories: {'Indulgent': 10685, 'Healthy': 1437, 'Quick Meals': 1308, 'Family-Friendly': 71}

In [5]:

df

Out[5]:

	recipe_id	recipe_name	ingredients	instructions	image_path	category	storage_tips	nutrition
0	1	0	['1 (3½-4-lb.) whole chicken', '2¾ tsp. kosher...	pat chicken dry with paper towels, season all ...	miso-butter-roast-chicken-acorn-squash-panzanella	Indulgent	Store ingredients in airtight containers; refr...	

	recipe_id	recipe_name	ingredients	instructions	image_path	category	storage_tips	nutrition
1	2	1	['2 large egg whites', '1 pound new potatoes (...]	preheat oven to 400°f and line a rimmed baking...	crispy-salt-and-pepper-potatoes-dan-kluger	Indulgent	Store ingredients in airtight containers; refr...	
2	3	2	['1 cup evaporated milk', '1 cup whole milk', ...]	place a rack in middle of oven; preheat to 400...	thanksgiving-mac-and-cheese-erick-williams	Indulgent	Store ingredients in airtight containers; refr...	
3	4	3	['1 (¾- to 1-pound) round italian loaf, cut in...	preheat oven to 350°f with rack in middle. gen...	italian-sausage-and-bread-stuffing-240559	Healthy	Store ingredients in airtight containers; refr...	
4	5	4	['1 teaspoon dark brown sugar', '1 teaspoon ho...	stir together brown sugar and hot water in a c...	newtons-law-apple-bourbon-cocktail	Quick Meals	Store ingredients in airtight containers; refr...	
...
13496	13497	13496	['1 cup all-purpose flour', '2/3 cup unsweeten...	preheat the oven to 350°f. into a bowl sift to...	brownie-pudding-cake-14408	Indulgent	Store ingredients in airtight containers; refr...	
13497	13498	13497	['1 preserved lemon', '1 1/2 pound butternut s...	preheat oven to 475°f. halve lemons and scoop ...	israeli-couscous-with-roasted-butternut-squash...	Indulgent	Store ingredients in airtight containers; refr...	
13498	13499	13498	['leftover katsuo bushi (dried bonito flakes) ...]	if using katsuo bushi flakes from package, moi...	rice-with-soy-glazed-bonito-flakes-and-sesame-...	Indulgent	Store ingredients in airtight containers; refr...	
13499	13500	13499	['1 stick (1/2 cup) plus 1 tablespoon unsalted...	melt 1 tablespoon butter in a 12-inch heavy sk...	spanakopita-107344	Indulgent	Store ingredients in airtight containers; refr...	
13500	13501	13500	['12 medium to large fresh poblano chiles (2 1...	lay 4 chiles on their sides on racks of gas bu...	mexican-poblano-spinach-and-black-bean-lasagne...	Indulgent	Store ingredients in airtight containers; refr...	

13501 rows × 8 columns

In [6]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 13501 entries, 0 to 13500
Data columns (total 8 columns):
#   Column                Non-Null Count  Dtype
---  -
0   recipe_id             13501 non-null  int64
1   recipe_name           13501 non-null  int64
2   ingredients            13501 non-null  object
3   instructions          13493 non-null  object
4   image_path            13501 non-null  object
5   category              13501 non-null  object
6   storage_tips          13501 non-null  object
7   nutrition_score       13501 non-null  float64
dtypes: float64(1), int64(2), object(5)
memory usage: 843.9+ KB
```

In [7]:

```
def prepare_input_text(row):
    ingredients = row['ingredients'] if pd.notna(row['ingredients']) else 'no ingredient'
    category = row['category'] if pd.notna(row['category']) else 'general'
    return f"Generate a {category} recipe using: {ingredients}"
```

In [8]:

```
def prepare_target_text(row):
    recipe_name = row['recipe_name'] if pd.notna(row['recipe_name']) else 'Delicious Rec
    instructions = row['instructions'] if pd.notna(row['instructions']) else 'Instructio
    return f"Recipe: {recipe_name}. Instructions: {instructions}"
```

In [9]:

```
df['input_text'] = df.apply(prepare_input_text, axis=1)
df['target_text'] = df.apply(prepare_target_text, axis=1)
```

In [10]:

```
train_df, val_df = train_test_split(df, test_size=0.15, random_state=42, stratify=df['ca

train_df = train_df.head(100).reset_index(drop=True)
val_df = val_df.head(25).reset_index(drop=True)
```

In [11]:

```
print(f'Training samples: {len(train_df)}')
print(f'Validation samples: {len(val_df)}')
```

Training samples: 100
Validation samples: 25

2. Load Pre-trained BART Model (Smaller Version)

In [12]:

```
model_name = 'facebook/bart-base'
tokenizer = BartTokenizer.from_pretrained(model_name)
model = BartForConditionalGeneration.from_pretrained(model_name).to(device)
```

In [13]:

```
print(f'Model loaded: {model_name}')
print(f'Number of parameters: {sum(p.numel() for p in model.parameters()),}')
```

Model loaded: facebook/bart-base
Number of parameters: 139,420,416

3. Test Pre-trained Model (Before Fine-tuning)

In [14]:

```
sample_inputs = [  
    "Generate a Healthy recipe using: chicken, rice, broccoli, garlic",  
    "Generate a Quick Meals recipe using: pasta, tomato sauce, basil",  
    "Generate an Indulgent recipe using: chocolate, cream, butter"  
]
```

In [15]:

```
print('Testing pre-trained model:\n')  
model.eval()  
  
for text in sample_inputs:  
    inputs = tokenizer(text, return_tensors='pt', max_length=128, truncation=True).to(device)  
  
    with torch.no_grad():  
        generated_ids = model.generate(  
            **inputs,  
            max_length=100,  
            num_beams=4,  
            early_stopping=True,  
            temperature=0.8,  
            do_sample=False # Deterministic for testing  
        )  
  
        output = tokenizer.decode(generated_ids[0], skip_special_tokens=True)  
        print(f'Input: {text}')  
        print(f'Output: {output}\n')
```

The following generation flags are not valid and may be ignored: ['temperature']. Set `TRANSFORMERS_VERBOSITY=info` for more details.

Testing pre-trained model:

Input: Generate a Healthy recipe using: chicken, rice, broccoli, garlic
Output: Generate a Healthy recipe using: chicken, rice, broccoli, garlic

Input: Generate a Quick Meals recipe using: pasta, tomato sauce, basil
Output: Generate a Quick Meals recipe using: pasta, tomato sauce, basil

Input: Generate an Indulgent recipe using: chocolate, cream, butter
Output: Generate an Indulgent recipe using: chocolate, cream, butter

4. Create PyTorch Dataset

In [16]:

```
class RecipeDataset(Dataset):  
    def __init__(self, dataframe, tokenizer, max_input_length=128, max_target_length=200):  
        self.data = dataframe  
        self.tokenizer = tokenizer  
        self.max_input_length = max_input_length  
        self.max_target_length = max_target_length
```

```

def __len__(self):
    return len(self.data)

def __getitem__(self, idx):
    input_text = self.data.loc[idx, 'input_text']
    target_text = self.data.loc[idx, 'target_text']

    input_encoding = self.tokenizer(
        input_text,
        max_length=self.max_input_length,
        padding='max_length',
        truncation=True,
        return_tensors='pt'
    )

    target_encoding = self.tokenizer(
        target_text,
        max_length=self.max_target_length,
        padding='max_length',
        truncation=True,
        return_tensors='pt'
    )

    labels = target_encoding['input_ids'].squeeze()
    labels[labels == self.tokenizer.pad_token_id] = -100
    return {
        'input_ids': input_encoding['input_ids'].squeeze(),
        'attention_mask': input_encoding['attention_mask'].squeeze(),
        'labels': labels
    }

train_dataset = RecipeDataset(train_df, tokenizer)
val_dataset = RecipeDataset(val_df, tokenizer)

```

In [17]:

```

print(f'Train dataset size: {len(train_dataset)}')
print(f'Validation dataset size: {len(val_dataset)}')

```

Train dataset size: 100

Validation dataset size: 25

5. Fine-tune Model (Optimized for Speed)

In [18]:

```

BATCH_SIZE = 2
GRADIENT_ACCUMULATION_STEPS = 4
EPOCHS = 3
LEARNING_RATE = 5e-5
WARMUP_STEPS = 10

```

In [19]:

```

train_loader = DataLoader(train_dataset, batch_size=BATCH_SIZE, shuffle=True, num_worker
val_loader = DataLoader(val_dataset, batch_size=BATCH_SIZE, shuffle=False, num_workers=0

```

In [20]:

```

optimizer = AdamW(model.parameters(), lr=LEARNING_RATE, weight_decay=0.01)
total_steps = (len(train_loader) // GRADIENT_ACCUMULATION_STEPS) * EPOCHS

```

```
scheduler = get_linear_schedule_with_warmup(optimizer, num_warmup_steps=WARMUP_STEPS, nu
```

In [21]:

```
print(f'Total training steps: {total_steps}')
print(f'Effective batch size: {BATCH_SIZE * GRADIENT_ACCUMULATION_STEPS}')
```

Total training steps: 36

Effective batch size: 8

In [22]:

```
def train_epoch(model, train_loader, optimizer, scheduler, device, epoch):
    model.train()
    total_loss = 0
    progress_bar = tqdm(train_loader, desc=f'Epoch {epoch} Training')

    for batch_idx, batch in enumerate(progress_bar):
        input_ids = batch['input_ids'].to(device)
        attention_mask = batch['attention_mask'].to(device)
        labels = batch['labels'].to(device)

        outputs = model(input_ids=input_ids, attention_mask=attention_mask, labels=labels)
        loss = outputs.loss / GRADIENT_ACCUMULATION_STEPS
        total_loss += loss.item() * GRADIENT_ACCUMULATION_STEPS
        loss.backward()

        if (batch_idx + 1) % GRADIENT_ACCUMULATION_STEPS == 0:
            optimizer.step()
            scheduler.step()
            optimizer.zero_grad()

    avg_loss = total_loss / (batch_idx + 1)
    progress_bar.set_postfix({'Loss': f'{avg_loss:.4f}'})

    if hasattr(torch, 'mps') and torch.backends.mps.is_available():
        if batch_idx % 10 == 0:
            torch.mps.empty_cache()

    return total_loss / len(train_loader)
```

In [23]:

```
def validate(model, val_loader, device):
    model.eval()
    total_loss = 0
    progress_bar = tqdm(val_loader, desc='Validating')

    with torch.no_grad():
        for batch_idx, batch in enumerate(progress_bar):
            input_ids = batch['input_ids'].to(device)
            attention_mask = batch['attention_mask'].to(device)
            labels = batch['labels'].to(device)

            outputs = model(input_ids=input_ids, attention_mask=attention_mask, labels=labels)
            total_loss += outputs.loss.item()

            progress_bar.set_postfix({'Val Loss': f'{outputs.loss.item():.4f}'})

    return total_loss / len(val_loader)
```

In [24]:

```

os.makedirs('models', exist_ok=True)
os.makedirs('training_logs', exist_ok=True)

print('Training started...\n')

best_val_loss = float('inf')
training_history = {'train_loss': [], 'val_loss': [], 'epoch_times': []}

for epoch in range(1, EPOCHS + 1):
    start_time = time.time()

    print(f'\nEpoch {epoch}/{EPOCHS}')
    train_loss = train_epoch(model, train_loader, optimizer, scheduler, device, epoch)
    val_loss = validate(model, val_loader, device)

    training_history['train_loss'].append(train_loss)
    training_history['val_loss'].append(val_loss)
    epoch_time = time.time() - start_time
    training_history['epoch_times'].append(epoch_time)

    print(f'Train Loss: {train_loss:.4f}')
    print(f'Validation Loss: {val_loss:.4f}')
    print(f'Epoch Duration: {epoch_time/60:.2f} minutes')

    if val_loss < best_val_loss:
        best_val_loss = val_loss
        print('New best model - saving...')
        model.save_pretrained('models/appetite_bart_best')
        tokenizer.save_pretrained('models/appetite_bart_best')

        model_info = {
            'epoch': epoch,
            'train_loss': train_loss,
            'val_loss': val_loss,
            'training_time': sum(training_history['epoch_times']),
            'date_saved': time.strftime('%Y-%m-%d %H:%M:%S')
        }

        with open('models/appetite_bart_best/training_info.json', 'w') as f:
            json.dump(model_info, f, indent=2)

    with open('training_logs/training_history.json', 'w') as f:
        json.dump(training_history, f, indent=2)

print('\nTraining complete!')

```

Training started...

Epoch 1/3

Epoch 1 Training: 0% | 0/50 [00:00<?, ?it/s]

Validating: 0% | 0/13 [00:00<?, ?it/s]

Train Loss: 4.4627

Validation Loss: 3.1086

Epoch Duration: 0.36 minutes

New best model - saving...

/opt/anaconda3/envs/aml/lib/python3.12/site-packages/transformers/modeling_utils.py:391
8: UserWarning: Moving the following attributes in the config to the generation config:

```
{'early_stopping': True, 'num_beams': 4, 'no_repeat_ngram_size': 3, 'forced_bos_token_id': 0}. You are seeing this warning because you've set generation parameters in the model config, as opposed to in the generation config.
```

```
warnings.warn(
```

```
Epoch 2/3
```

```
Epoch 2 Training: 0%|          | 0/50 [00:00<?, ?it/s]
```

```
Validating: 0%|          | 0/13 [00:00<?, ?it/s]
```

```
Train Loss: 3.5052
```

```
Validation Loss: 2.9424
```

```
Epoch Duration: 0.34 minutes
```

```
New best model - saving...
```

```
Epoch 3/3
```

```
Epoch 3 Training: 0%|          | 0/50 [00:00<?, ?it/s]
```

```
Validating: 0%|          | 0/13 [00:00<?, ?it/s]
```

```
Train Loss: 3.2148
```

```
Validation Loss: 2.8462
```

```
Epoch Duration: 0.33 minutes
```

```
New best model - saving...
```

```
Training complete!
```

6. Load Fine-tuned Model and Test

```
In [25]:
```

```
finetuned_model = BartForConditionalGeneration.from_pretrained('models/appetite_bart_best')
finetuned_tokenizer = BartTokenizer.from_pretrained('models/appetite_bart_best')
finetuned_model = finetuned_model.to(device)
finetuned_model.eval()
```

```
print('Fine-tuned model loaded successfully!')
```

```
try:
```

```
    with open('models/appetite_bart_best/training_info.json', 'r') as f:
```

```
        training_info = json.load(f)
```

```
    print(f'Best epoch: {training_info["epoch"]}') 
```

```
    print(f'Validation loss: {training_info["val_loss"]:.4f}')
```

```
    print(f'Training time: {training_info["training_time"]/60:.1f} minutes')
```

```
except:
```

```
    pass
```

```
Fine-tuned model loaded successfully!
```

```
Best epoch: 3
```

```
Validation loss: 2.8462
```

```
Training time: 1.0 minutes
```

```
In [26]:
```

```
test_recipes = [
```

```
    "Generate a Healthy recipe using: chicken breast, broccoli, olive oil, garlic, lemon"
```

```
    "Generate a Quick Meals recipe using: pasta, tomato sauce, basil, mozzarella cheese"
```

```
    "Generate an Indulgent recipe using: chocolate, cream, butter, vanilla, eggs",
```

```
]
```

```
print('Testing fine-tuned model:\n')
```

```
for i, test_input in enumerate(test_recipes, 1):
```

```
    print(f'Test {i}')
```

```
    print(f'Input: {test_input}')
```



```

inputs = finetuned_tokenizer(test_input, return_tensors='pt', max_length=128, trunca

with torch.no_grad():
    generated_ids = finetuned_model.generate(
        **inputs,
        max_length=150,
        num_beams=5,
        no_repeat_ngram_size=3,
        early_stopping=True,
        temperature=0.8,
        do_sample=True
    )

ai_recipe = finetuned_tokenizer.decode(generated_ids[0], skip_special_tokens=True)
print(f'Generated Recipe: {ai_recipe}')
print('-' * 80)
print()

```

Testing fine-tuned model:

Test 1

Input: Generate a Healthy recipe using: chicken breast, broccoli, olive oil, garlic, lemon

Generated Recipe: Recipe: 1. Instructions: preheat oven to 350°F. season chicken breast with salt and pepper. add broccoli, broccoli, and broccoli to a large bowl. season with garlic and season with salt, pepper, and garlic. cook until tender, about 2 minutes. remove chicken breast from the breast and let cool slightly. stir in broccoli and broccoli. let cool, stirring occasionally, until the broccoli is tender and tender. add chicken breast and broccoli and cook until browned, about 3 minutes. add the broccoli and chicken breast to the chicken breast, then season with olive oil. serve chicken breast in a small bowl over medium-high heat, stirring often, until cooked through, about 10 minutes. transfer to a plate and

Test 2

Input: Generate a Quick Meals recipe using: pasta, tomato sauce, basil, mozzarella cheese

Generated Recipe: Recipe: 1125. Instructions: preheat oven to 400°F. season with salt and pepper. add mozzarella cheese, basil, and basil to the pasta. toss to coat. add the basil and basil. cover with tomato sauce and season with basil. serve in a large saucepan over medium-high heat, stirring occasionally, until the sauce is golden brown, about 3 minutes. serve on a large plate. toss with remaining basil and tomato sauce. serve with basil and toss with basil, tomato sauce, salt, and pepper, tossing occasionally. add basil and pepper to the saucepan. serve. garnish with 1/2 cup of basil and sprinkle with salt, pepper, and 1/4 cup of mo

Test 3

Input: Generate an Indulgent recipe using: chocolate, cream, butter, vanilla, eggs

Generated Recipe: Recipe: Instructions: preheat oven to 350°F. whisk together chocolate, butter, vanilla, and eggs in a large bowl. whisk until smooth, about 2 minutes. add eggs, sugar, and vanilla. stir until smooth. whisk in vanilla, eggs, and chocolate mixture. add vanilla and eggs and whisk until combined. pour in the chocolate mixture, stirring often, until smooth and smooth. let cool, about 5 minutes. stir in eggs and vanilla, stirring occasionally, until the mixture is smooth. add milk and vanilla and stir to combine. add chocolate and vanilla mixture and stir until blended. whisk again, stirring constantly, until combined, about 10 minutes. transfer chocolate mixture to a bowl and pour over the remaining chocolate mixture

7. Evaluation with ROUGE Scores

In [27]:

```
from rouge_score import rouge_scorer

def simple_evaluation(model, tokenizer, val_df, device, num_samples=10):
    scorer = rouge_scorer.RougeScorer(['rouge1', 'rouge2', 'rougeL'], use_stemmer=True)
    rouge1_scores = []
    rouge2_scores = []
    rougeL_scores = []

    model.eval()
    sample_indices = np.random.choice(len(val_df), min(num_samples, len(val_df)), replace=True)

    print(f'Evaluating on {len(sample_indices)} samples...\n')

    for idx in sample_indices:
        input_text = val_df.iloc[idx]['input_text']
        reference_text = val_df.iloc[idx]['target_text']

        inputs = tokenizer(input_text, return_tensors='pt', max_length=128, truncation=True)

        with torch.no_grad():
            generated_ids = model.generate(**inputs, max_length=150, num_beams=4, early_stopping=True)

        prediction = tokenizer.decode(generated_ids[0], skip_special_tokens=True)

        try:
            scores = scorer.score(reference_text, prediction)
            rouge1_scores.append(scores['rouge1'].fmeasure)
            rouge2_scores.append(scores['rouge2'].fmeasure)
            rougeL_scores.append(scores['rougeL'].fmeasure)
        except:
            pass

    results = {
        'ROUGE-1': np.mean(rouge1_scores),
        'ROUGE-2': np.mean(rouge2_scores),
        'ROUGE-L': np.mean(rougeL_scores)
    }

    return results

eval_results = simple_evaluation(finetuned_model, finetuned_tokenizer, val_df, device, num_samples)

print('Evaluation Results:')
print(f'ROUGE-1: {eval_results["ROUGE-1"]:.4f}')
print(f'ROUGE-2: {eval_results["ROUGE-2"]:.4f}')
print(f'ROUGE-L: {eval_results["ROUGE-L"]:.4f}')

overall_rouge = (eval_results['ROUGE-1'] + eval_results['ROUGE-2'] + eval_results['ROUGE-L']) / 3
print(f'\nOverall ROUGE: {overall_rouge:.4f}')
```

Evaluating on 10 samples...

Evaluation Results:

ROUGE-1: 0.3340

ROUGE-2: 0.0986

ROUGE-L: 0.1897

Overall ROUGE: 0.2074

8. Safety Features and Model Card

Allergen Detection

In [28]:

```
import re
from datetime import datetime

ALLERGENS = ['peanut', 'milk', 'egg', 'soy', 'fish', 'shellfish', 'wheat', 'gluten', 'se
```

In [29]:

```
def detect_allergens(text):
    found = [a for a in ALLERGENS if re.search(rf'\b{a}', text.lower())]
    return found
```

In [30]:

```
def safety_check(recipe_text):
    allergens = detect_allergens(recipe_text)
    if allergens:
        print(f'Warning: Contains allergens: {allergens}')

    if any(bad in recipe_text.lower() for bad in ['kill', 'poison', 'suicide']):
        print('Unsafe content detected! Review required.')

    return allergens
```

In [31]:

```
sample_recipe = "Recipe: Peanut Butter Cookies. Instructions: Mix peanut butter, eggs, s
allergens = safety_check(sample_recipe)
print(f'Detected allergens: {allergens}')
```

Warning: Contains allergens: ['peanut', 'egg']

Detected allergens: ['peanut', 'egg']

Prediction Logging

In [32]:

```
def log_prediction(input_text, output_text, allergens=None):
    os.makedirs('logs', exist_ok=True)
    entry = {
        'timestamp': datetime.now().isoformat(),
        'input': input_text,
        'output': output_text,
        'allergens': allergens or []
    }

    with open('logs/predictions.jsonl', 'a') as f:
        f.write(json.dumps(entry) + '\n')
    print('Logged prediction for monitoring.')
```

```
log_prediction(
    "Generate a Healthy recipe using: chicken, broccoli",
    "Recipe: Grilled Chicken with Broccoli...",
    allergens=[]
)
```

Logged prediction for monitoring.

Model Card

In [33]:

```
def create_model_card(name='Appetite-BART', version='v1.0'):
    card = f"""# Model Card: {name}

## Model Information
- **Version:** {version}
- **Base Model:** facebook/bart-base
- **Purpose:** Generate recipes from ingredients
- **Training Data:** Curated AppetiteDataset.csv (100 samples)

## Intended Use
- Input: List of ingredients and desired category (Healthy, Quick Meals, Indulgent, Fami
- Output: Recipe name and cooking instructions

## Limitations & Risks
- May generate recipes containing common allergens
- Limited to patterns seen in training data
- Not a substitute for professional dietary advice

## Mitigation Strategies
- Allergen detection filter implemented
- User feedback collection for improvement
- Manual review for edge cases

## Contact
Project Maintainer: Sharath

## License
Educational use only
"""

    with open('MODEL_CARD.md', 'w') as f:
        f.write(card)
    print('Model Card created: MODEL_CARD.md')

create_model_card()
```

Model Card created: MODEL_CARD.md

9. Complete Pipeline Function

This function combines everything for easy recipe generation with safety checks:

In [34]:

```
def generate_recipe_with_safety(ingredients, category='Healthy', model=None, tokenizer=N
    if model is None:
```

```

        model = finetuned_model
    if tokenizer is None:
        tokenizer = finetuned_tokenizer

    input_text = f"Generate a {category} recipe using: {ingredients}"

    inputs = tokenizer(input_text, return_tensors='pt', max_length=128, truncation=True)

    model.eval()
    with torch.no_grad():
        generated_ids = model.generate(
            **inputs,
            max_length=150,
            num_beams=5,
            no_repeat_ngram_size=3,
            early_stopping=True,
            temperature=0.8,
            do_sample=True
        )

    recipe_text = tokenizer.decode(generated_ids[0], skip_special_tokens=True)

    allergens = detect_allergens(recipe_text)

    log_prediction(input_text, recipe_text, allergens)

    result = {
        'input': input_text,
        'recipe': recipe_text,
        'category': category,
        'allergens': allergens,
        'timestamp': datetime.now().isoformat()
    }

    return result

print('\n' + '='*80)
print('COMPLETE RECIPE GENERATION PIPELINE')
print('='*80 + '\n')

test_cases = [
    ('chicken, rice, vegetables', 'Healthy'),
    ('pasta, cheese, tomato sauce', 'Quick Meals'),
    ('chocolate, cream, strawberries', 'Indulgent')
]

for ingredients, category in test_cases:
    result = generate_recipe_with_safety(ingredients, category)

    print(f"Input: {ingredients}")
    print(f"Category: {category}")
    print(f"Generated Recipe: {result['recipe']}")
    if result['allergens']:
        print(f"Allergens: {result['allergens']}")
    print('-' * 80)
    print()

```

=====

COMPLETE RECIPE GENERATION PIPELINE

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Logged prediction for monitoring.

Input: chicken, rice, vegetables

Category: Healthy

Generated Recipe: Recipe: 1. Instructions: preheat oven to 350°f. add chicken, rice, and rice to a large bowl and stir until cooked through. add rice, chicken, and chicken to the bowl. stir until rice is tender, about 1/2 cup. cover with foil and let cool, stirring occasionally, until the chicken is tender. transfer to a plate and serve with rice and chicken. refrigerate until ready to serve. transfer chicken to a bowl and serve on a rimmed baking sheet. place chicken on a baking sheet and cover with plastic wrap. serve chicken and rice in a small bowl over medium-high heat. let cool slightly, stirring often, until chicken is golden brown, about 2 to 3 minutes.

Logged prediction for monitoring.

Input: pasta, cheese, tomato sauce

Category: Quick Meals

Generated Recipe: Recipe: 1. Instructions: preheat oven to 350°f. add pasta, cheese, and tomato sauce to a large bowl and mix until smooth. add cheese, tomato sauce, and salt. season with salt and pepper and season with pepper and pepper. serve on a large plate. refrigerate until cheese is tender, about 15 minutes. serve in a small bowl over medium-high heat. cover with plastic wrap and refrigerate for at least 30 minutes. add the pasta and cheese mixture to the pasta mixture. toss to coat with salt, pepper, salt, and pepper, stirring often. serve with cheese and pepper in a large saucepan over medium heat, stirring occasionally, until the pasta is golden brown. serve immediately.

Logged prediction for monitoring.

Input: chocolate, cream, strawberries

Category: Indulgent

Generated Recipe: Recipe: 1. Instructions: preheat oven to 350°f. whisk together chocolate, strawberries, vanilla, and vanilla until smooth, about 2 minutes. add strawberries, cream, and strawberries to a bowl and mix until smooth. transfer to a large bowl. cover with plastic wrap and refrigerate until firm, about 10 minutes. let cool, stirring occasionally, to remove excess. stir in strawberries and vanilla. add chocolate and vanilla, stirring often, and stir until smooth and smooth. add the strawberries and strawberries. pour in the chocolate mixture. stir until the strawberries are golden brown, about 3 minutes. transfer the strawberries to the bowl and toss in the remaining strawberries. whisk in the cream and vanilla mixture. pour the strawberries over the
