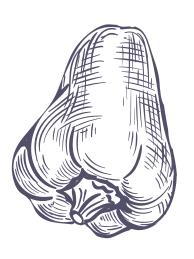




Byte BITES

Generating **nutritious**, **personalized** recipes that best **utilize user ingredients** through a user-friendly web interface



01

Problem Statement

Target Audience



44%

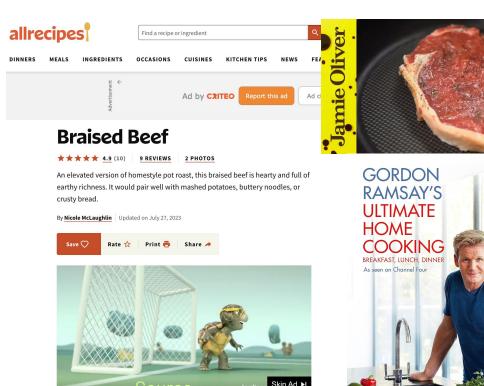
of Americans meal prep regularly

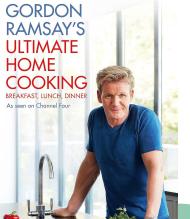
26.9%

22.2%

of Americans meal prep to save time of Americans meal prep to eat he<mark>althier</mark>

Current cooking resources overlook user needs





inconvenient, time consuming Little to no nutritional information





an Average American spends

\$1300

on food that end up being unused / discarded



Problems that Byte**BEIGS** addresses



Convenience

Our user can conveniently generate recipes with our user-friendly web app.



Nutrition

Recipes generated by ByteBites provide detailed nutritional facts to users.



Waste

ByteBites generates recipes that optimally utilize ingredients that the user has bought from their grocery store runs.

Target Audience





- Looking for efficient, nutritious ways to prep meals
- Looking to optimally utilize ingredients to save costs
- High acceptance for Al-based applications



02

Unique Value Proposition

Our Uniqueness

01

Automated Ingredient Recognition

Use OCR model to identify ingredients from receipt

03

Health and Wellness integration

Monitor nutrients intake and provide insights on user's diet

02



Personalized Meal Planning

Dietary restrictions, time constraints & meal types

04

User-Driven Recipe Refinement

Incorporate user ratings with RLHF to improve recipe recommendation

Competitive Position



High) Yumnly: evel of Personalization (Low dishgen SuperCook



Level of Automation (Low → High)



03

Bytebites <u>Demo</u>



04

Scalability and Efficiency

Model & Optimization



Data

We use recipe data from food.com containing **180K recipes** and **700K reviews**, reflecting 18 years of user interactions and uploads.



Model

We finetune **facebook-opt-125m** model on the recipe dataset and we plan to use the **Llama3-8B** model in our final application.



Infrastructure Considerations







Model Optimization & Scalability

Fine-tuning



RLHF







Model Optimization & Scalability

Fine-tuning

Task specific

PEFT (LoRA)

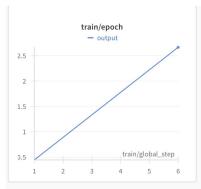
Efficient Training

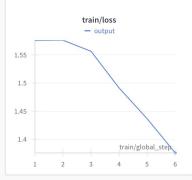
RLHF

Tailored for user

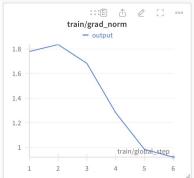
Performance Monitoring

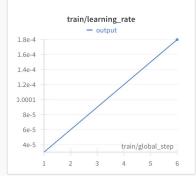




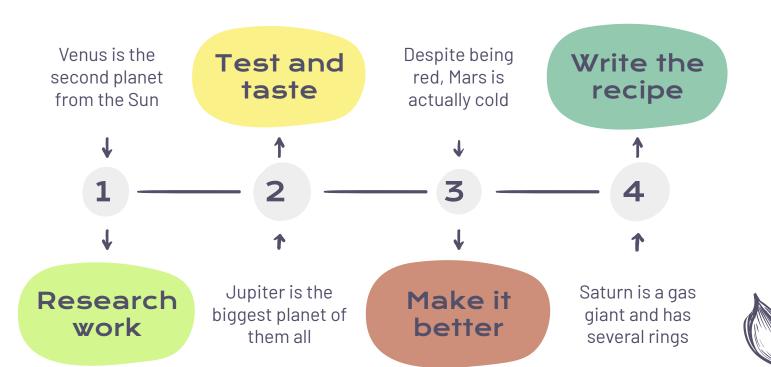








HOW TO DEVELOP RECIPES



Technical Scalability

Discuss how your application can scale to meet growing demands.

- 1. GCP infrastructure: dynamically adjust our resources
- 2. architecture including cloud storage and computing solutions support data-intensive processes like OCR and large language model (LLM) operations
- **3.** Containerized deployment via docker.
- **4.** Vertex AI: automatically scales for real-time ML workloads, ensuring high availability for our users



Performance Optimization

Explain any optimizations implemented for efficiency (e.g., fine-tuning, infrastructure choices).

To reduce training costs, we select a medium-sized ML model as our base model. For optimized performance, we have fine-tuned the language model to specialize in ecipe generation. We use PEFT (LoRA) to optimize for efficient fine-tuning. We save the fine-tuned model weights so that the model can be loaded and directly deployed on Vertex AI for our application. This targeted fine-tuning enhances model accuracy and reduces latency, making interactions faster and more relevant to user needs.

Additionally, the application uses RAG (Retrieval-Augmented Generation) to handle complex queries and provide highly contextualized responses, significantly improving efficiency in generating personalized meal plans. We're also implementing caching mechanisms for frequently requested data and pre-computing certain steps in the recipe design pipeline to minimize runtime and improve response times.



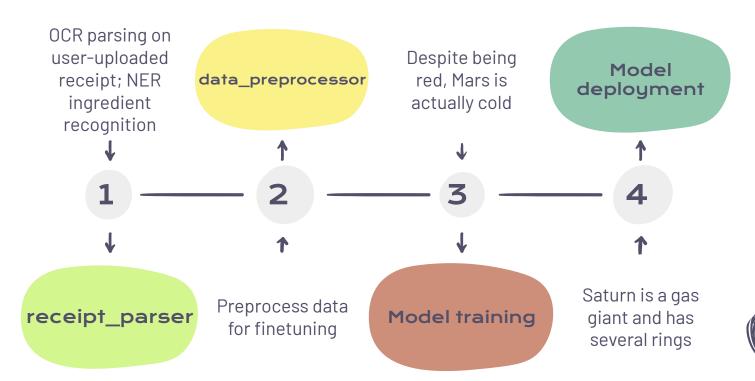
Infrastructure considerations

We will be using GCP services for robust cloud computing, storage, and ML model hosting.

- GCP Cloud Storage for secure data handling and version control
- Vertex AI for streamlined machine learning deployment, and Vertex AI Pipelines for efficient ML workflow management
- Docker for containerization, ensuring consistent and rapid deployment across environments



ByteBites pipeline





04

Future Development & Growth Potential

05

Future Development & Growth Potential



Personalization

Improve recipe suggestions based on individual tastes, dietary restrictions, and past behavior

Community

Foster a community of cooking enthusiasts and encourage social interaction.

Integration

Seamless interaction between the app and smart kitchen appliances.

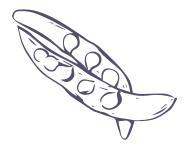
FAMILY

Saturn is a gas giant and has several rings

Personalization

Improve recipe suggestions based on individual tastes, dietary restrictions, and past behavior

Community



Personalization

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Integration

Seamless interaction between the app and smart kitchen appliances.

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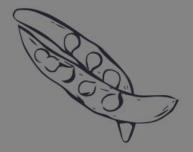
Personalization

Improve recipe suggestions based on individual tastes, dietary restrictions, and past behavior

Integration

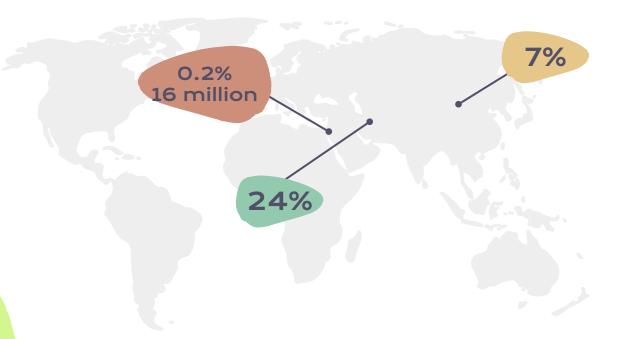
Seamless interaction between the app and smart kitchen appliances.

Community



Expanding Globally Through Localization

Objective: To attract international users by tailoring the app's content to various languages and regional culinary preferences.



Islam

Halal food: no alcohol, cblood, pork

Buddhism

Vegetarianism

Judaism

Kosher food: no port and shellfish

Retail Partnership

Objective: Integrate with grocery retailers to enable direct ordering from the app, making shopping seamless and convenient.





THANKS

DO YOU HAVE ANY QUESTIONS?

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