

Leveraging Machine Learning to Generate Product Descriptions in E-Commerce

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Text-to-Text

(LLMs)



THE TEAM



BUSINESS PROBLEM FRAMING

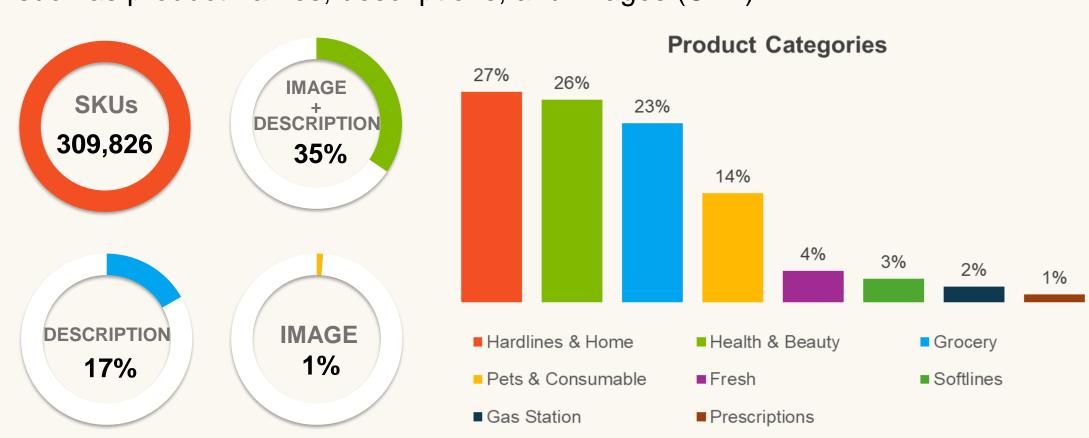
Our team collaborated with a nationwide grocery store chain in the US to tackle the challenge of online product listings. The chain mandates all vendors to provide product details and images for online sales. However, many fail to resulting in numerous unlisted products and loss. Besides, vendors struggle to verify if their descriptions meet standards. Only 35% of products meet these criteria, leading to inadequate submissions.



We propose a new solution that utilizes machine learning technology in product description generation and content scoring. In order to enhance the scalability and automation of the process, we also architected a pipeline in the Microsoft Azure platform to deploy the models. By ensuring that a high-quality description accompanies each product, we aim to reduce operation costs and foster a more satisfying online shopping experience.

DATA

The dataset is provided by a major grocery store in the U.S. It comprises 309,826 unique products spanning diverse categories and contain attributes such as product names, descriptions, and images (URL).



Our project uses state-of-the-art models for automating online retail product

- 1. Image-to-text task (including Optical Character Recognition & Captioning): To
- 2. Text-to-text task: To refine auto-generated product description

Based on practical trials with open-source models on Hugging Face, we're confident in Al's capability for these tasks. Moreover, if clients pursue extreme performance and are open to reasonable cost, we also offer Azure Al Vision as a premium, upgraded solution for the image-to-text task.

GPT 3.5 TURBO

Relevance

Completeness Readability

Our project presents a methodology for enhancing online

product descriptions through a ML pipeline in Azure cloud with

alternate paths for each product scenario. For the products with

no description or improper description, we extract information

The pooled contents from vendors and images are refined into

the desired format, and then the text-to-text LLM is used to

generate descriptions. Lastly, the output quality is assessed

from the product images through image-to-text model.

using a scoring algorithm powered by LLMs.

Confidence Score

Product information CSV format gets uploaded online

Learning for action

Machine Learning generates product

Generated content and scores are saved in CSV file

To deploy our solution, we developed an Azure pipeline that automatically

IMPACT

Our project demonstrates promising results in reducing operation costs, enhancing customer satisfaction, and boosting profitability for the retailer. It also has great potential for similar applications across various industries, such as consumer goods. We hope our work facilitates further research and development in this field for broader impacts.

Suppliers 300K+ **Products**

6139

Retailers Customers

100% high-quality product descriptions

Save 99% original labor time & costs

New product listing speed

ANALYTICS PROBLEM FRAMING

METHODOLOGY

With Description

Without Description

Without Image

With Description

Without Image

Without Description

MODEL BUILDING

descriptions, focusing on two tasks:

- capture information from product images

The generated text will be measured by a scoring model that focuses on 4 rubrics:

File triggers Functions to

DEPLOYMENT & LIFE CYCLE MANAGEMENT

descriptions and scores

processes product description creation and scoring. Architected with cloud services, the pipeline provides scalability, reliability, and cost-effectiveness. After deployment, we can seamlessly update our models without interrupting the service. Leveraging features such as model versioning, rolling updates, and A/B testing, we can iterate rapidly, adapt to user needs, and maximize business agility and competitiveness.

HOW

76% average quality score

ACKNOWLEDGEMENTS



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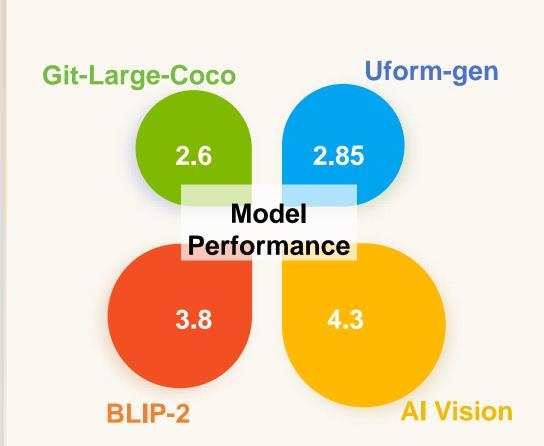
OCR Model Evaluating Image-to-Text Models:

Al Vision

Read Output

Image-to-text Model

(OCR & Captioning)

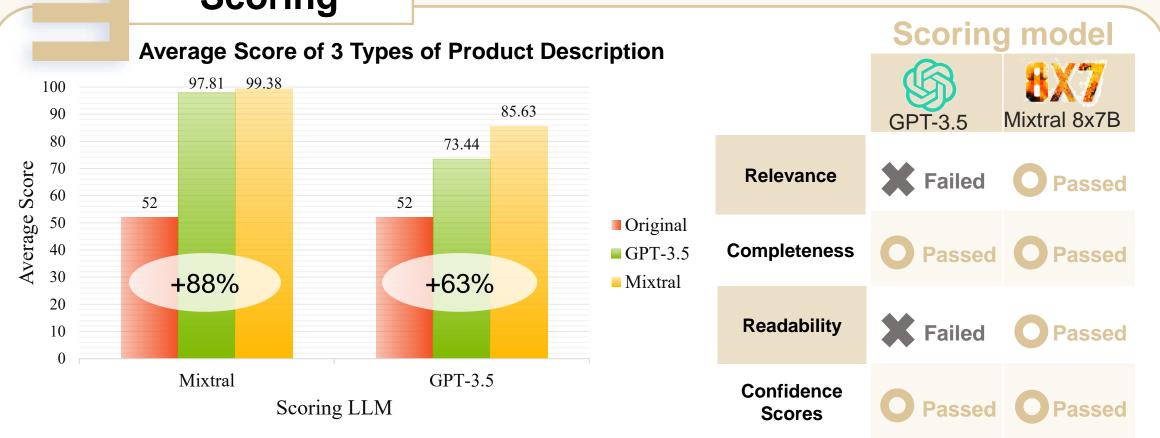


- Sample Selection: 80 diverse product images across 8 categories. 2. Model Testing: Human evaluators
- rate text outputs from different opensource models on a 1-5 scale (1 = irrelevant, 5 = highly accurate). 3. Final Choice: Select the model with the highest average score for

implementation.

NATURALLY first WHOLE CHOLESTEROL® AMPLY MADE Gluten Free Sweetened Whole Grain Cereal with Real Honey & Natural Flavor NET WT 1 LB 1 OZ (17 OZ) **Caption Output:** a box of cereal with a red bowl of oats Tags Output: breakfast cereal, box, cereal, convenience food, food, container

LLM Generation Natural Flavor Honey Nut Cheerios Generated Description: Introducing our Natural Flavor Honey Nut Cheerios, a delightful blend of wholesome ingredients and natural flavors crafted to start your day off right. Made with the goodness of whole oats as the first ingredient, this cereal is a hearty choice... Scoring



We utilize LLMs to score the generated results based on 4 rubrics. Compared to the vendor-provided product information, our solution can increase the average quality score by more than 63% regardless of which LLM is used.