Internship Technical Assessment

Data Analyst

TASKS: Data Extraction, Cleaning, EDA, Visualization, Pipelining, Documentation

and Reporting

TIMELINE: 26/11/24 - 08/12/24

OVERVIEW:	Business Need: Smallholder Commercial Farmers are in need of live data updates on market prices on their commodities, to inform them on seasonal/monthly market price performance and ideal markets to sell to/in. Data Analyst's Objective: Analyse Market Price Data over the last 3 years (1st October 2021 – 30th September 2024), present these findings, insights and recommendations, along with a live weekly update on market prices, to adequately respond to the business case above.
DATA AND SCOPE:	Data Source: You will obtain data from the KAMIS website. Geographical Scope: Nairobi, Nyandarua, Nakuru, Meru and Kirinyaga Counties. Commodities Scope: Dry Maize, White Irish Potatoes, Dry Onions, Watermelons, Ripe and Unripe (Cooking) Bananas, Oranges, Mangoes, Kales and Regular Spinach.
QUESTIONS:	 1.What are the seasonal (quarterly) and monthly trends? 2.Which are the best and worst times of the year to sell these commodities and why? (Be commodity-specific) 3.Which are the top 3 priced markets per county? 4. What is today's wholesale and retail price in each county? (Live data pipeline, only for the dashboard).

DELIVERABLES:

TOOLS

EVALUATION CRITERIA:

- 1.A neat, coherent, well documented, titled and subtitled Jupyter notebook of your analysis exercise, with markdowns on findings, insights and recommendations for your audience (farmers) where applicable. (.ipynb)
- 2.A link to your PDF presentation **summary** of your analysis exercise: Introduction, Findings, Insights and Recommendations.
- 3.A link to your published dashboard of this analysis exercise, with a tile on live-updated price of the previous day. Only the price will be live, the rest is past, static data.

Data Analysis:

Pythonic notebooks, preferably Jupyter.

Presentation Report:

We highly recommend Canva, but you can use what you are most comfortable with to get the job done.

Dashboard:

Tableau or Power BI.

1. Jupyter Notebook (50%)

Correctness and completeness of data cleaning.

Quality and accuracy of the analysis.

Logical/Coherent flow of analysis.

Use of appropriate visualizations.

Clarity of documentation (comments and markdown cells).

2. Summary Report (20%)

Clarity and organization of insights.

Relevance and feasibility of recommendations.

Visual appeal and professional formatting.

3. Dashboard (30%)

Usability and interactivity.

Clarity in communicating key metrics and insights.

Aesthetic design and adherence to best practices.

Data Scientist / Machine Learning Engineer

TASKS: Data Extraction, Cleaning, EDA, Pre-processing, Model Building, Model

Evaluation and Iteration, Model Deployment

TIMELINE: 26/11/24 - 08/12/24

OVERVIEW:

Business Need:

Smallholder Commercial Wheat Farmers are in need a diagnostic tool to detect, diagnose and treat wheat crop pests and diseases to prevent yield reduction.

Data Scientist/Machine Learning Engineer's Objective:

Develop an image classification model to diagnose wheat crop pests and diseases, that can be deployed and used by smallholder farmers on a web platform.

DATA AND SCOPE:

Data Source:

You will obtain data from **Kaggle**.

Pest and Disease Classes:

Pests: Select at least 1 pest class

Diseases: Select at least 3 disease classes

NB: Do not forget to train and test with the 'Healthy' class as well.

Jupyter Notebook:

- Evidence of exploratory data analysis (EDA) on the provided dataset.
- Explanation of pre-processing steps.
- Model Development
- Model Evaluation

Deployed Model:

- Model Deployment Code/Files
- A Streamlit application script for deploying the trained model.
- Features to include: upload image button, prediction results and a display of confidence scores or probabilities for each class.

Please note that we require a working link to the deployed model (along with all necessary documentation and README files).

DELIVERABLES:

1. Data Handling & Pre-processing (20%)

Effective exploratory data analysis (Dataset characteristics like class distributions, insights into data quality).

Data Pre-processing: Use of appropriate techniques like resizing, normalization, and augmentation and handling of class imbalances.

2. Model Selection & Development (50%)

Model Selection process.

Appropriate use of libraries and model frameworks.

Appropriate model analysis and evaluation metrics.

Visualization of model performance.

Hyperparameter tuning to optimize parameters.

3. Deployment & Usability (20%)

A working deployment that can take an input image and provide predictions.

User-friendly interface.

4. Documentation & Communication (10%)

Clear, concise, and well-documented workflow in the notebooks. Logical flow from problem definition to results and conclusions. README files: Comprehensive setup instructions for running the solution.