# OPTIMIZING HIGH-PERFORMANCE DATA PROCESSING FOR LARGE-SCALE WEB CRAWLERS

#### PG Mall

Group D

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#### Introduction

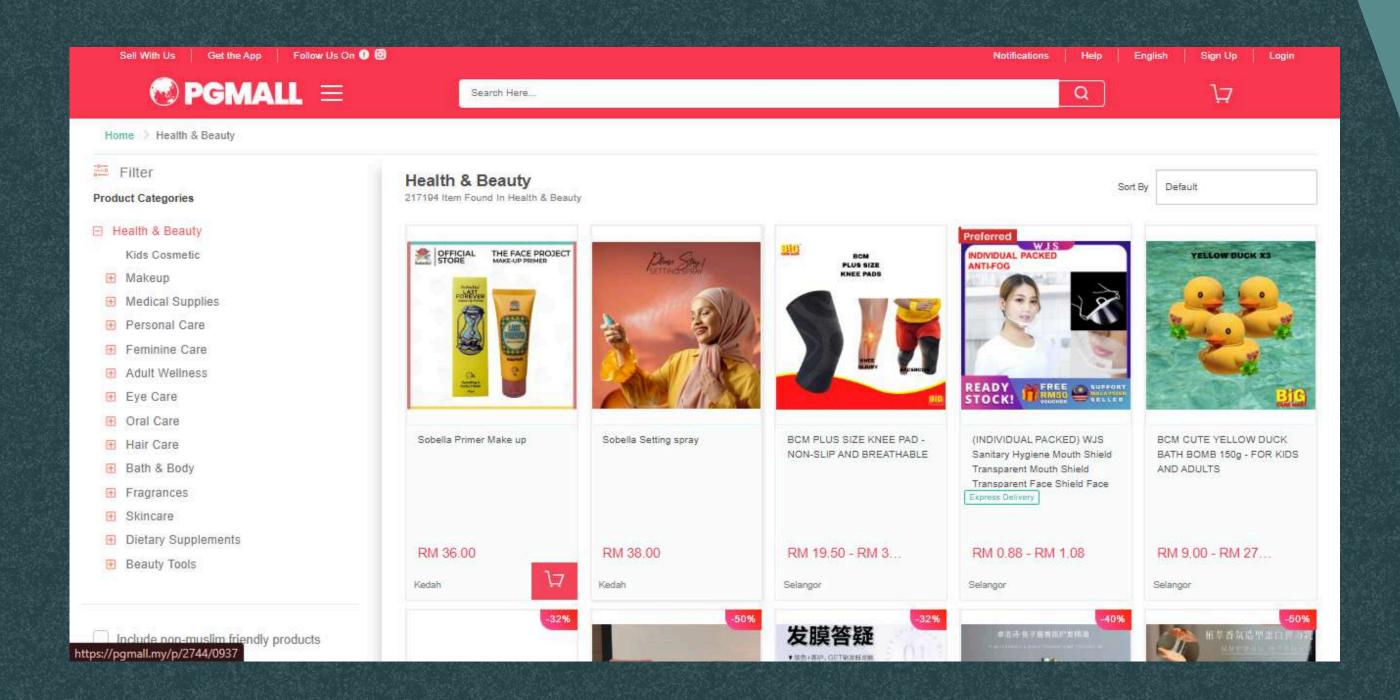
This project develops a scalable solution for extracting, cleaning, and analyzing real-world e-commerce data, focusing on web crawling, data preprocessing, and evaluating text-processing libraries under performance constraints.

#### **Objectives:**

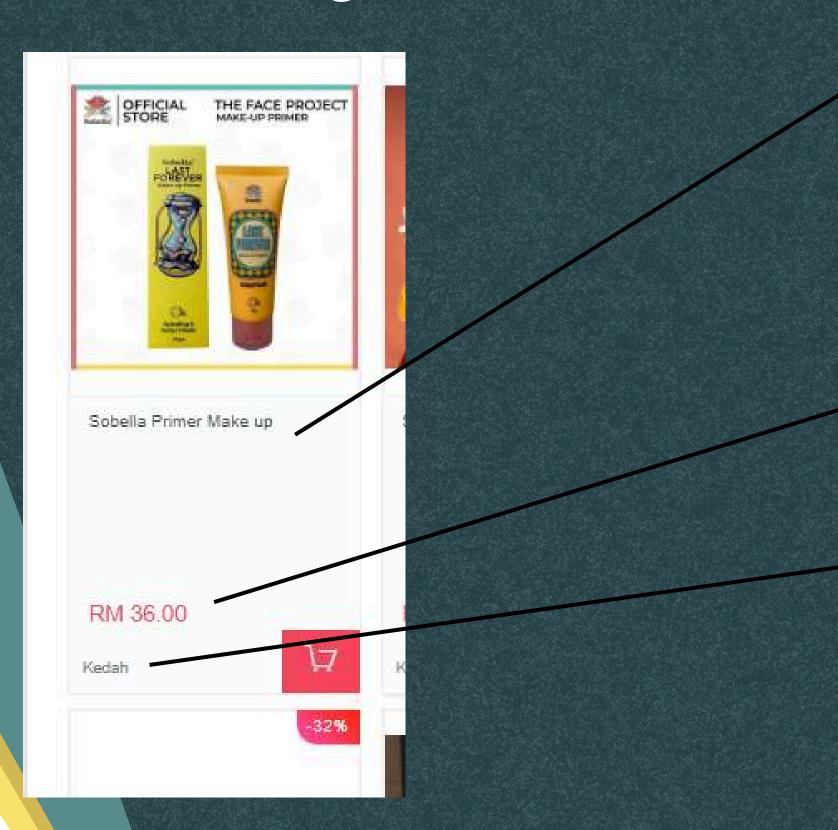
- Build a web crawler that extracts >=100,000 data from the PG Mall
- Perform text processing and data cleaning on the scraped data using text normalisation and noise removal techniques.
- Compare text processing libraries based on key performance indicators
- Apply high-performance computing techniques to optimise the text processing pipeline.
- Evaluate the system's performance, identifying bottlenecks and discussing improvements made.

#### Targeted Website & Data Field





#### Targeted Website & Data Field



Product Name

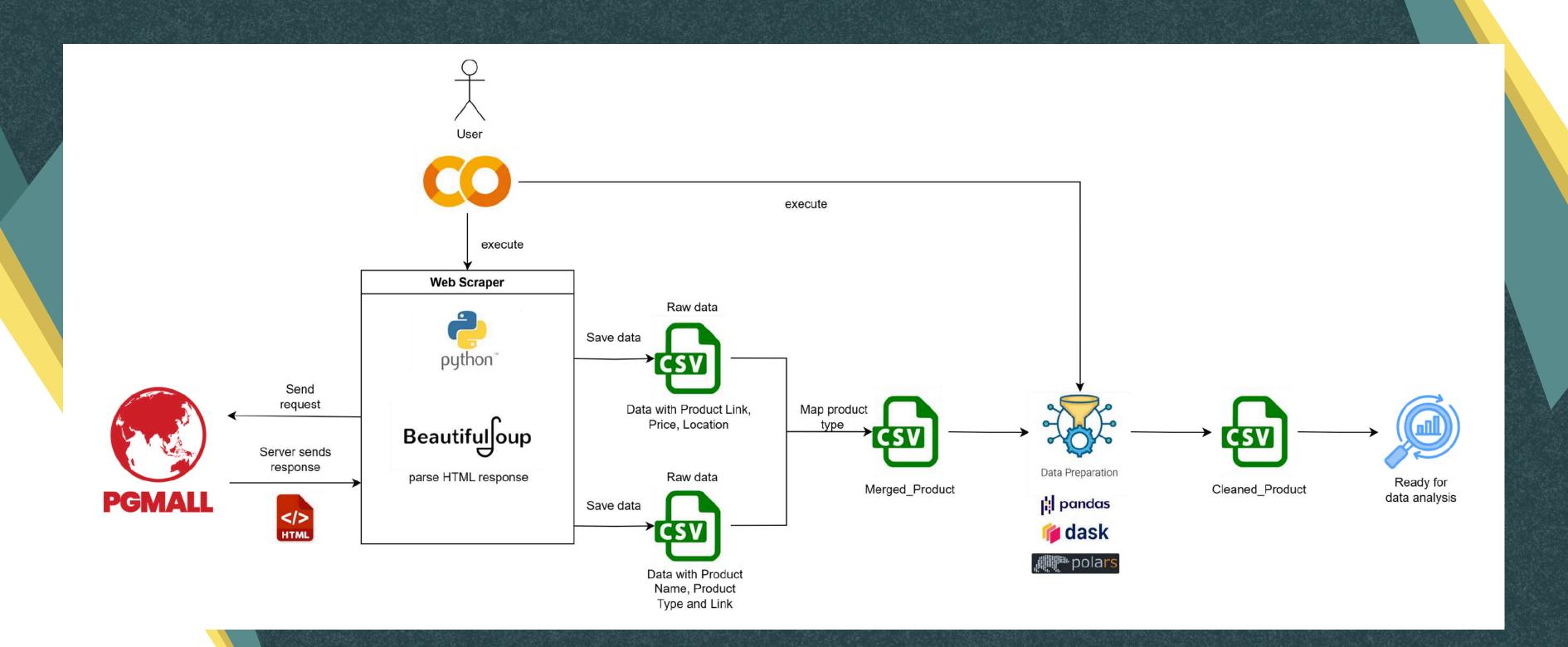
• Link \*extracted from HTML

Price

Location

• Product Type \*added during mapping phase

#### System Architecture



#### Tools & Frameworks















### Data Collection

#### Data Field and Method

Data field	Initial Status	
product_name	Contain characters outside of the normal ASCII	
link	No problem	
price	String type instead of float has a range of prices and string "RM" in cells	
location	Contain null cells	
product_type	Contain null cells	

- Sequential Pagination via While Loop
- Retry-Logic and Concurrency Control (max\_workers = 5)
- Multithreaded Implementation (concurrent.futures.ThreadPoolExecutor)

#### Ethical Consideration

- Respects Retry-After headers during rate limiting
- Uses a generic User-Agent string to mimic a browser
- Manually checked robot.txt and confirmed using code

```
def check robots txt():
                                                        Scraping allowed by robots.txt.
   """Check robots.txt before scraping"""
                                                   Scraping page 1...
   rp = urllib.robotparser.RobotFileParser()
   rp.set url("https://pgmall.my/robots.txt ")
                                                   Page 1 scraped and saved (50 items). Total: 50
   try:
       rp.read()
   except Exception as e:
       print(f"Error reading robots.txt: {e}")
       return False
   can fetch = rp.can fetch("*", "https://pgmall.my/category ")
   if not can fetch:
       print(" ▲ Scraping disallowed by robots.txt.")
   else:
       print(" Scraping allowed by robots.txt.")
   return can fetch
```

## Data Processing

#### MAPPING PHASE

product_name	link	price	location
MuscleRulz L-Carnit	https://pgma	RM128.00	Selangor
MuscleRulz Iso Rulz	https://pgma	RM314.00	Selangor
Kevin Levrone Gold	https://pgma	RM265.00	Selangor
CNI RJ Moisturizer	https://pgma	RM17.60	Selangor
CNI RJ Hair Cream	https://pgma	RM13.40	Selangor
CNI RJ Shower Crear	https://pgma	RM17.20	Selangor
CNI Siang-Siang (10	https://pgma	RM11.10	Selangor
CNI RJ Intimate Was	https://pgma	RM18.80	Selangor
ALHA ALFA ROYAL PF	https://pgma	RM59.90	Selangor



#### Item\_list.csv

product_name	link	product_type
Xiaomi Smart Scale S200	https://pgmall.my/p/	Bekalan Perubatan
Morilins Pain Relief Patch 1	https://pgmall.my/p/	Bekalan Perubatan
TYNOR KNEE CAP WITH PAT	https://pgmall.my/p/	Bekalan Perubatan
MEGA GAZGO 200MG 10 SO	https://pgmall.my/p/	Bekalan Perubatan
HmbG Borosilicate Glass B	https://pgmall.my/p/	Bekalan Perubatan
Potassium lodide AR / ACS,	https://pgmall.my/p/	Bekalan Perubatan
TOPSEAL Sterile Plain Gauz	https://pgmall.my/p/	Bekalan Perubatan
TOPSEAL Sterile Plain Gauz	https://pgmall.my/p/	Bekalan Perubatan



			용성하다 없었는데 사용으로 보고한	
product_name	link	price	location	product_type
ANZEN Intelligent	https://pgmall.my/	RM58.88	Selangor	Alat Kecantikan
Madeshow Akeme	https://pgmall.my/	RM140.00	Selangor	Alat Kecantikan
Madeshow Akeme	https://pgmall.my/	RM125.00	Selangor	Alat Kecantikan
WAHL Super Tape	https://pgmall.my/	RM288.00	Selangor	Alat Kecantikan
[ BeautyVault ] RE	https://pgmall.my/	RM6.00	Selangor	Alat Kecantikan
Pensonic Hair Dry	https://pgmall.my/	RM48.00	Kelantan	Alat Kecantikan
Hair Dryer 1400W/	https://pgmall.my/	RM 19.90 - RM 45	Selangor	Alat Kecantikan
(STONG SUCTION	https://pgmall.my/	RM14.99	Selangor	Alat Kecantikan
ANZEN Intelligent	https://pgmall.my/	RM58.88	Selangor	Alat Kecantikan

updated\_item\_list.csv

merged\_product\_list.csv

#### DATA CLEANING

#### **Load & Initial Check**

- Loaded updated\_item\_list.csv
- Checked for: Duplicate entries & Null values

#### **Duplicate & Null Handling**

- Dropped 7 duplicated records
- Replaced nulls with "Unknown" in location & product\_type

#### **Raw Data Issues**

- Product names had unreadable characters
- Prices stored as ranges with "RM" prefix

#### **Finalization**

- Renamed cleaned data fields
- Saved cleaned dataset as Item\_list\_cleaned.csv

#### **Product Name Cleaning**

 Removed unreadable characters

#### **Price Cleaning**

- Extracted numeric value from string
- Selected lowest value if price is a range

#### DATA STRUCTURE

Product Name	Link	Location	Product Type	Price
			••	
MuscleRulz L-Carnitine 3000mg (33 Servings) (READ DESC		Selangor	Unknown	128
MuscleRulz Iso Rulz (5LBS)	https://pgmall.my/p/2740/6329	Selangor	Unknown	314
Kevin Levrone Gold Whey (2kg)	https://pgmall.my/p/2740/6306	Selangor	Unknown	265
CNI RJ Moisturizer	https://pgmall.my/p/2740/6267	Selangor	Unknown	17.6
CNI RJ Hair Cream	https://pgmall.my/p/2740/6266	Selangor	Unknown	13.4
CNI RJ Shower Cream 300 ml	https://pgmall.my/p/2740/6258	Selangor	Unknown	17.2
CNI Siang-Siang (100g) - Body Talc Absorbs Perspiration	https://pgmall.my/p/2740/6256	Selangor	Unknown	11.1
CNI RJ Intimate Wash	https://pgmall.my/p/2740/6251	Selangor	Unknown	18.8
ALHA ALFA ROYAL PROPOLIS LUMINOUS SILK FOUNDATIO	https://pgmall.my/p/2740/1287	Selangor	Penjagaan mulut	59.9
ALHA ALFA FLAWMINOUS CUSHION FOUNDATION	https://pgmall.my/p/2740/0510	Selangor	Alat solek	79.9
Xiaomi Smart Scale S200   High-precision sensor   180 c	https://pgmall.my/p/2740/0500	Selangor	Bekalan Perubatan	59
[BeautyVault] READY STOCK   RHODE - Pocket Blush	https://pgmall.my/p/2230/3788	Selangor	Alat solek	196
Dettol Shower Gel Refill Pouch 800ml/ 850ml / Dettol Onz	https://pgmall.my/p/Y714/0413	Selangor	Mandian	10.9
Zen Basil Seeds   edible basil seeds usda organic, kosh	https://pgmall.my/p/2739/4542	New Jersey	Makanan Tambahan	259
[ BeautyVault ] READY STOCK   ARIANA GRANDE FRAGRAN	https://pgmall.my/p/J226/0608	Selangor	Wangian	350
Morilins Pain Relief Patch 1bag 5pcs	https://pgmall.my/p/2739/4336	Johor	Bekalan Perubatan	13.35
SKINTIFIC 3x Acid Intensive Acne Spot Gel (15ml)	https://pgmall.my/p/2739/4075	Selangor	Penjagaan kulit	59
HAUS MAGICKISS GLITTER LIPSTICK 4G	https://pgmall.my/p/2739/4071	Selangor	Alatsolek	27
HAUS POPSY SUPERSTAY LIPMATTE 3ML	https://pgmall.my/p/2739/4065	Selangor	Penjagaan kulit	27

Cleaned dataset with product name, link location, price and product type

# Optimization Techniques

#### Pandas

- Uses standard Pandas operations
- Processes data in-memory as a single chunk
- Applies functions row-by-row using .apply()
- Shows typical performance characteristics of unoptimized Pandas code

#### import pandas as pd

```
# Start performance timer and process monitor
start time = time.time()
process = psutil.Process(os.getpid())
# Drop duplicate rows
df = df.drop duplicates()
# Fill in null
df['location'].fillna('Unknown', inplace=True)
df['product_type'].fillna('Unknown', inplace=True)
# Standardize price format
df['cleaned_price'] = df['price'].apply(extract_lowest_price)
df['cleaned price'] = df['cleaned price'].round(2)
df.drop(columns=['price'], inplace=True)
# Clean unreadable characters from product name
df['product name'] = df['product name'].apply(lambda x: re.sub(r"[^\x00-\x7F]+", '', str(x)))
df.rename(columns={
    'product_name': 'Product Name',
    'cleaned price': 'Price',
    'location': 'Location',
    'link': 'Link',
    'product type': 'Product Type'
}, inplace=True)
# Save cleaned data
df.to csv("Item list cleaned.csv", index=False, float format='%.2f')
# Log performance metrics
end_time = time.time()
elapsed time = end time - start time
cpu percent = process.cpu percent(interval=1)
memory usage mb = process.memory info().rss / 1024 ** 2
throughput = df.shape[0] / elapsed_time
```

#### Dask

- Leverages Dask's parallel processing capabilities
- Uses lazy evaluation with explicit computation triggers
- Processes data in partitions for memory efficiency
- Demonstrates distributed computing benefits

#### import dask.dataframe as dd

```
# Rename columns
df = df.rename(columns={
    'product name': 'Product Name',
    'cleaned price': 'Price',
    'location': 'Location',
    'link': 'Link',
    'product type': 'Product Type'
# Drop the original 'price' column
df = df.drop('price', axis=1)
# Compute the result and round the price
result = df.compute()
result['Price'] = result['Price'].round(2)
# Save to CSV
result.to csv("Item list cleaned dask.csv", index=False, float format='%.2f')
# Performance metrics
end time = time.time()
elapsed time = end time - start time
cpu_percent = process.cpu_percent(interval=1)
memory_usage_mb = process.memory_info().rss / 1024 ** 2
throughput = result.shape[0] / elapsed_time
```

#### Polars

- Utilizes Polars' Rust-based, columnar data processing
- Employs vectorized operations and expression-based transformations
- Shows native string operations and regex optimizations
- Demonstrates high-performance data frame operations

#### import polars as pl

```
# Fill nulls with default values
df = df.with columns([
    pl.col("location").fill_null("Unknown"),
   pl.col("product type").fill null("Unknown")
# Extract cleaned versions first
df = df.with columns([
   pl.col("price").str.extract(r"(\d+(?:\.\d+)?)").cast(pl.Float64).round(2).alias("price"),
    pl.col("product name").str.replace all(r"[^\x00-\x7F]+", "").alias("product name")
1)
# Reorder columns if needed
df = df.select([
    "product_name", "price", "location", "link", "product_type"
# Rename columns
df = df.rename({
    "product name": "Product Name",
    "price": "Price",
    "location": "Location",
    "link": "Link",
    "product_type": "Product Type"
# Save cleaned CSV
df.write_csv("Item_list_cleaned_optimized.csv")
# End performance tracking
end time = time.time()
elapsed time = end time - start time
cpu percent = process.cpu percent(interval=1)
memory_usage_mb = process.memory_info().rss / 1024 ** 2
throughput = df.shape[0] / elapsed time
```

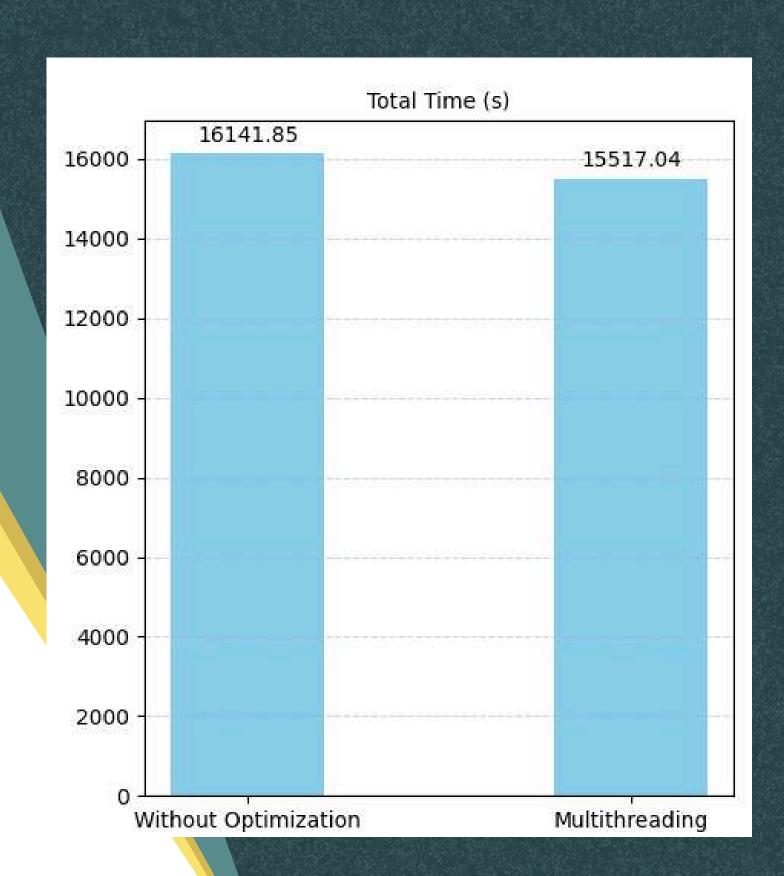
# Performance Evaluation

#### Web Scraping

Table 6.0.1 Performance Evaluation of Web Scraping

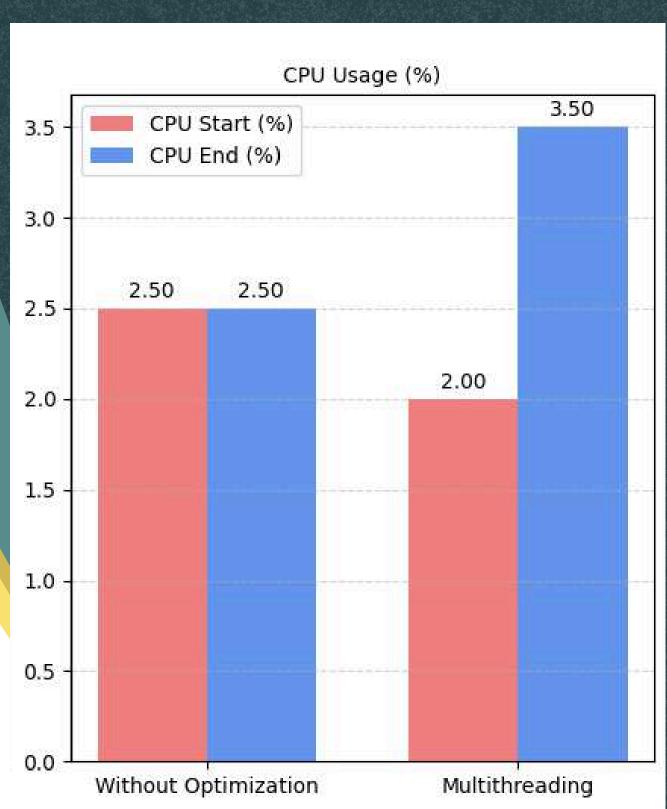
Metric	Without optimization	With optimization using multithreading
Total data scraped	200009 rows	200010 rows
Total time taken	16141.85 seconds (4 hours 28 minutes 58 seconds)	15517.04 seconds (4 hours 18 minutes 36 seconds)
Start CPU	2.5%	2.0%
End CPU	2.5%	3.5%
Start memory	166.88MB	167.01 MB
End memory	197.24 MB	274.76 MB
Used memory	30.36 MB	107.75 MB
Throughput	12.39 rec/s	12.89 rec/s

#### Web Scraping - Total Processing Time



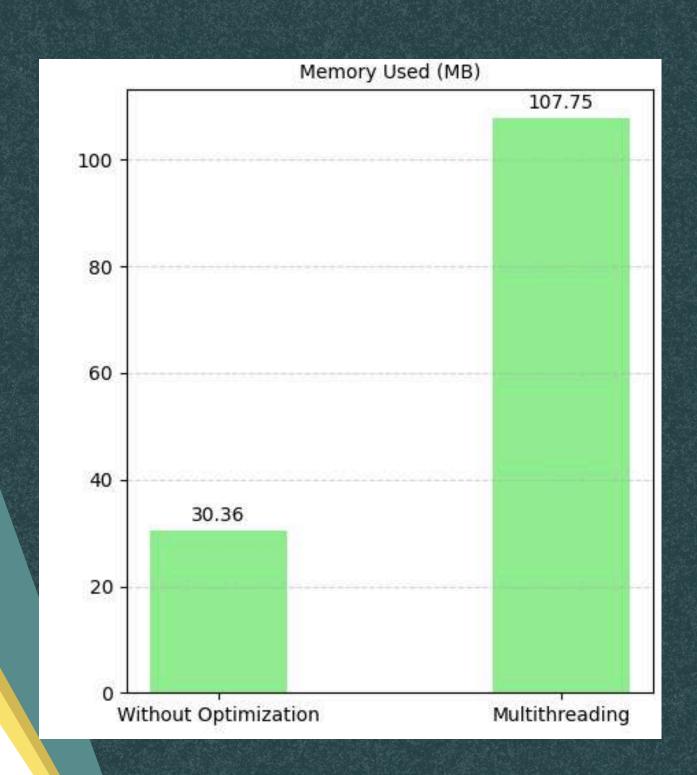
- Basic scraping: 16141.85 seconds
- Optimized scraping:
   15517.04 seconds
- 4% faster with multithreading

#### Web Scraping - CPU Usage



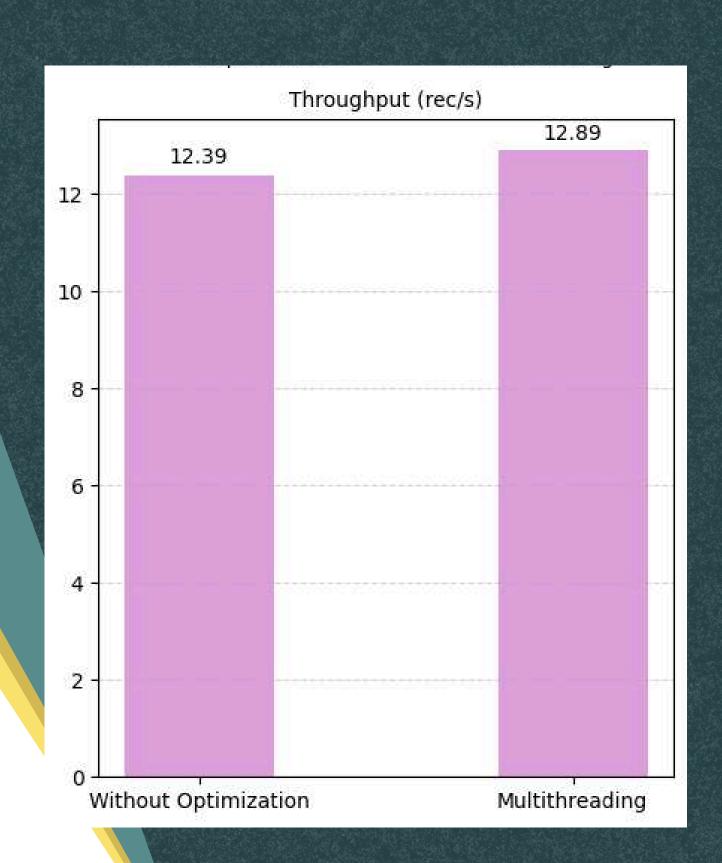
- Without optimization: Stable at 2.50%
- With optimization: Increased from 2.0% to 3.5%
- Multithreading uses more
   CPU for concurrent tasks

#### Web Scraping - Memory Usage



- Without optimization: 30.36
   MB
- With optimization: 107.75 MB
- Multithreading increases memory usage due to parallel threads

#### Web Scraping - Throughput



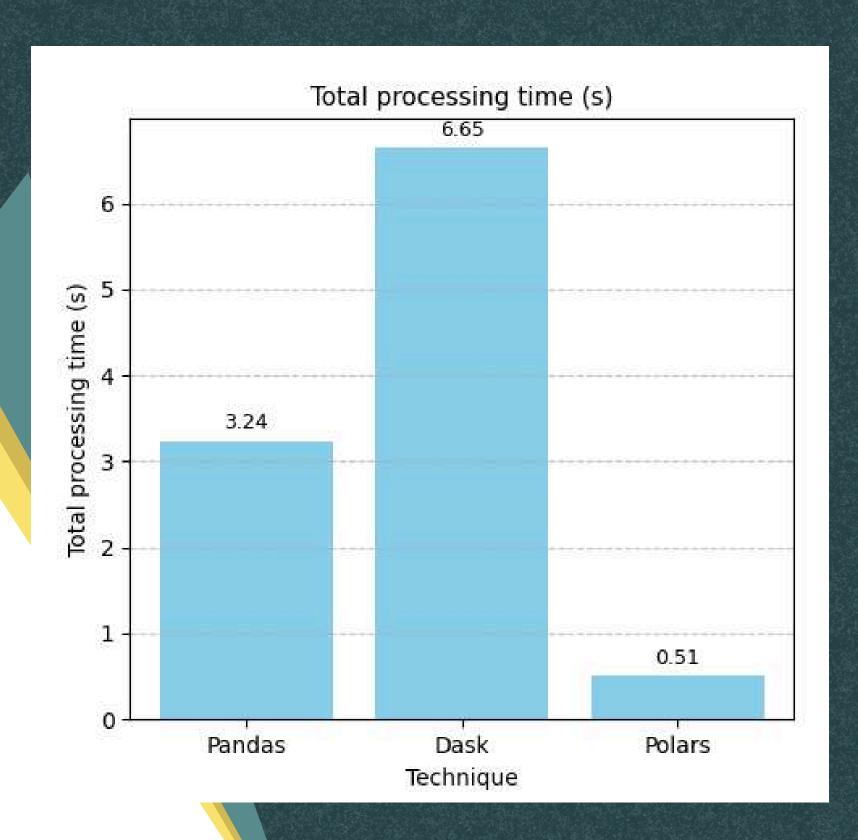
- Without optimization: 12.39 rec/s
- With optimization: 12.89 rec/s
- 4% higher throughput with optimization

#### Data Cleaning

Table 6.0.2 Performance Evaluation of Data Cleaning

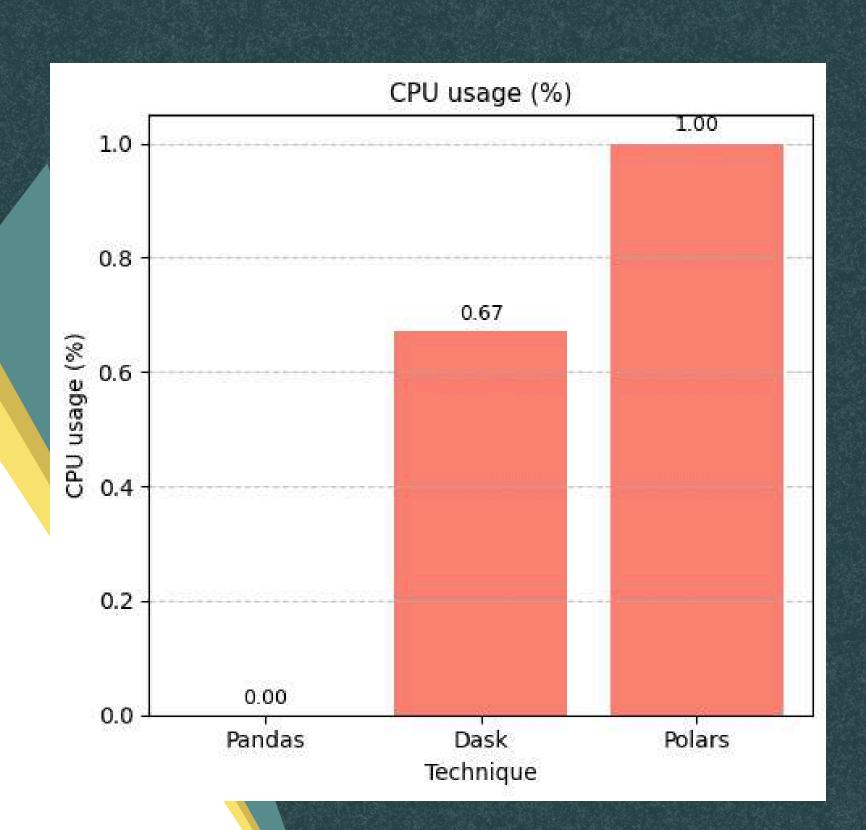
Techniques	Metric	Run 1	Run 2	Run 3	Avanaga
recumiques	Metric	Kunı	Kun 2	Kun 3	Average
Pandas	Total processing time	4.03 seconds	2.89 seconds	2.80 seconds	3.24 seconds
	Memory usage	269.84 MB	243 <u>.</u> 20 MB	218. <u>34MB</u>	243. <u>79MB</u>
	CPU usage	0.0 <u>%</u>	0.0%	0.0%	0.0%
	Throughput	51029.69 rec/s	71183.34 rec/s	51488.39 rec/s	57900.47 rec/s
Dask	Total processing time	6.70 seconds	6.86 seconds	6.38 seconds	6.65 seconds
	Memory usage	552 <u>.</u> 43 MB	447 <u>.</u> 21 MB	395.86 MB	465 <u>.</u> 17 MB
	CPU usage	1.0%	1.0%	0.0%	0.67 %
	Throughput	30719.88 rec/s	29994.33 rec/s	18149.05 rec/s	26287.75 rec/s
Polar	Total processing time	0.58 seconds	0.35 seconds	0.59 seconds	0.51 seconds
	Memory usage	319 <u>.</u> 55 MB	203.18 MB	240 <u>.</u> 30 MB	254 <u>.</u> 34 MB
	CPU usage	1.0%	1.0%	1.0%	1.0%
	Throughput	356363.62 rec/s	200205.37 rec/s	224245.89 rec/s	260271.63 rec/s

#### Data Cleaning - Total Processing Time



- Polars: Fastest at 0.51 seconds
- Dask: Slowest at 6.65 seconds
- Pandas: Moderate speed at 3.24 seconds
- Polars wins due to multithreaded, columnar processing

#### Data Cleaning - CPU Usage



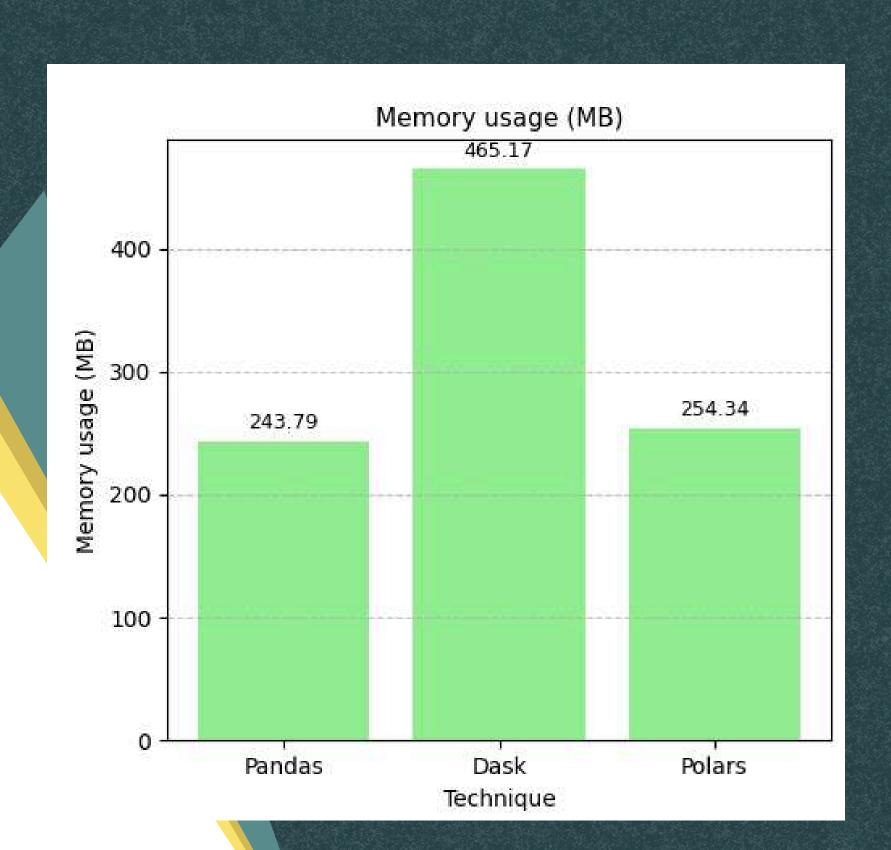
• Pandas: 0.0%

• Dask: 0.67%

• Polars: 1.0% (highest – fully

utilizes CPU cores)

#### Data Cleaning - Memory Usage



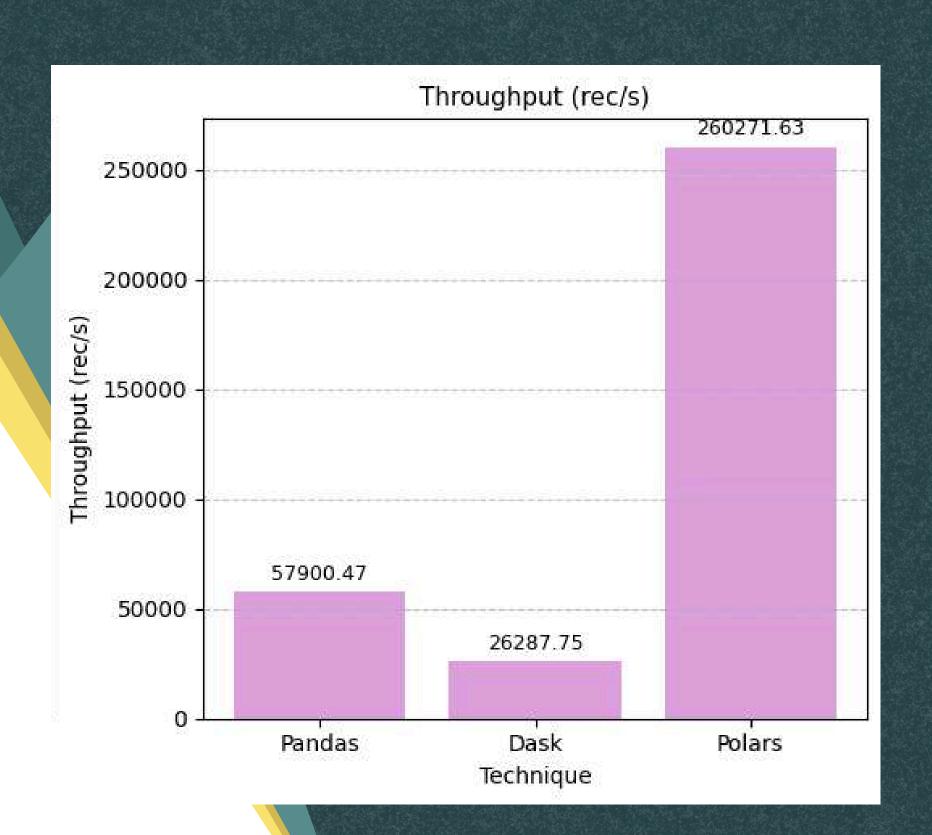
 Pandas: 243.79 MB – least memory used

• Polars: 254.34

• Dask: 465.17 MB

 Dask uses more memory due to parallel partition handling

#### Data Cleaning - Throughput



Polars: 260271.63 rec/sec Highest throughput

Pandas: 57900.47rec/s

• Dask: 26287.75 rec/sec

Polars processed data most efficiently

# Challenges & Limitations

Challenges	Descriptions
Website Restrictions	Some websites block scraping via CAPTCHAs, rate limiting and dynamic content
Limited Data Availability	Many Malaysian websites have small datasets and some sites paginate data poorly, making large-scale scraping difficult.
Inconsistent Data Structure	Websites change layout, breaking scrapers and some data is hidden behind login walls

Limitations	Descriptions
Hardware Dependencies	Scraping speed varies by CPU/RAM and low-end devices struggle with large-scale scraping
Slow Scraping Process	Polar/Dask clean data fast, but scraping itself bottlenecked by network latency and rate-limiting delays
Maintenance Overload	Scarpers need constant updates if websites change HTML structure and proxy/IP rotation may be needed to avoid bans

#### Conclusion

#### **Data Collection**

 Over 200,000 cleaned records collected from PG Mall's "Health & Beauty" category.

#### **Data Cleaning Framework**

- Polars: Most efficient and scalable
- Pandas: Suitable for simplicity and lowmemory environments.
- Dask: Best for very large or distributed datasets.

#### **Performance Comparison**

- Multithreading offers faster data collection and higher throughput.
- Trade-off: higher CPU and memory usage, but acceptable for large data volumes.

#### Improvement

- Use proxy rotation and headless browsers
- Expand to more categories or websites
- Use asynchronous requests
- Deploy scraping on cloud platforms (e.g., GCP, AWS)
- Use modular code or public APIs.

# THANK YOU