EDA Notebook - Sec Filings 10-K Dataset.

Doc Owner - Joel Markapudi.

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Plan for notebook EDA: (potentially)

- 1. Schema of dataset ovevrview, Column overview, etc.
- 2. Company stats, Temporal coverage/stats (yearly filings, etc.), Sentence Distribution stats.
- 3. Section Mapping & Distribution, Section breakdown, Text by Section stats, Common N-Grams (Top 20) ??
- 4. Missing data, Missing data patterns, Cardinality analysis. (heatmaps, etc.)
- 5. Data Quality flags, Edge cases.

High level plan for now.

```
In [17]: # Notebook bootstrap
    from pathlib import Path
    import sys, os

# point Python to your ./src so `import` works
```

```
PROJECT ROOT = Path.cwd().resolve()
         SRC = PROJECT ROOT / "src"
         if str(SRC) not in sys.path:
             sys.path.insert(0, str(SRC))
         # env vars (reads assets/config.env you already have)
         from dotenv import load dotenv
         load_dotenv(PROJECT_ROOT / "assets" / "config.env")
         # autoreload so edits in src/ reflect without restarting kernel
         %load ext autoreload
         %autoreload 2
       The autoreload extension is already loaded. To reload it, use:
         %reload_ext autoreload
In [18]: import nltk
         nltk.download('punkt')
                                   # Sentence tokenizer
         nltk.download('stopwords') # English stopwords
         nltk.download('averaged_perceptron_tagger') # POS tagging (optional)
        [nltk_data] Downloading package punkt to
        [nltk_data]
                      C:\Users\joems\AppData\Roaming\nltk_data...
       [nltk_data] Package punkt is already up-to-date!
       [nltk_data] Downloading package stopwords to
        [nltk_data]
                     C:\Users\joems\AppData\Roaming\nltk_data...
        [nltk_data] Package stopwords is already up-to-date!
        [nltk_data] Downloading package averaged_perceptron_tagger to
                       C:\Users\joems\AppData\Roaming\nltk_data...
        [nltk_data]
        [nltk_data] Package averaged_perceptron_tagger is already up-to-
       [nltk_data]
                          date!
Out[18]: True
In [19]: ## Starting all EDA:
         # Section 1: Data Understanding - SEC 10-K Filings Dataset
         import pandas as pd
         import polars as pl
         import numpy as np
         import seaborn as sns
```

```
import matplotlib.pyplot as plt
from pathlib import Path
from itables import init_notebook mode, show
import warnings
from IPython.display import display, HTML
# warnings.filterwarnings('ignore')
# init notebook mode(all interactive=True)
# sns.set style("whitegrid")
# plt.rcParams['figure.figsize'] = (12, 6)
# plt.rcParams['font.size'] = 11
# # Polars display settings (show ALL columns, no truncation)
# pl.Config.set_tbl_width_chars(1000)
# pl.Config.set_tbl_cols(-1) # -1 means show ALL columns
# pl.Config.set_tbl_rows(100) # Show up to 100 rows when printing
# # Pandas display settings (for later use)
# pd.set option('display.max columns', None)
# pd.set option('display.max colwidth', 100)
# pd.set option('display.precision', 2)
def display table with html(df, title=""):
    """Display pandas DataFrame as styled HTML table"""
    display(HTML(f"<h3>{title}</h3>"))
   html_str = df.to_html(classes='table table-striped table-hover', border=0)
    display(HTML(html str))
print("Environment ready")
# Load dataset
DATA_PATH = Path("../data/exports/sec_filings_small_full.parquet")
df = pl.read parquet(DATA PATH)
print(f"Dataset loaded: {df.shape[0]:,} rows x {df.shape[1]} columns")
print(f"Memory usage: {df.estimated_size('mb'):.1f} MB")
```

```
Environment ready
Dataset loaded: 200,000 rows x 19 columns
Memory usage: 144.1 MB

In [21]: display_table_with_html(
          df.head(3).to_pandas(),
          title="Dataset Preview (First 3 Rows)"
)
```

Dataset Preview (First 3 Rows)

ITEM 1.BUSINESS General AAR CORP. and its subsidiaries {'1d': are referred to 0, herein '30d': 2020-07-AAR 0000001750_10-0000001750_10collectively as **0** 0000001750 0 [AIR] 1 0, 21 CORP K_2020 K_2020_section_1_0 "AAR," '5d': "Company," 1} "we," "us," and "our" unless the context indicates otherwise.

1 0000001750 AAR was founded in 1951, organized in	0 {'1d': 0, '30d': 0,	AAR 0000001750_10- CORP K_2020	0000001750_10- K_2020_section_1_1	2	[AIR]
1955 and	'5d':				
reincorporated in Delaware in 1966.	1}				

	cik	sentence	section	labels	filingDate	name	docil) sentencelD	sentenceCount	tickers	excl
2		We are a diversified provider of products and ervices to the worldwide aviation and government and defense markets.	0	{'1d': 0, '30d': 0, '5d': 1}	2020-07- 21		0000001750_10 K_202			[AIR]	

docID

```
In [22]: # Table 1: Complete schema with statistics
         schema_info = []
         for col in df.columns:
             dtype = str(df[col].dtype)
             null_count = df[col].null_count()
             null_pct = (null_count / len(df)) * 100
             unique_count = df[col].n_unique()
             try:
                 if df[col].dtype == pl.List(pl.Utf8):
                     sample = str(df[col].drop_nulls().head(1).to_list()[0][:2]) + "..."
                 else:
                     sample = str(df[col].drop_nulls().head(1).to_list()[0])
                     if len(sample) > 50:
                         sample = sample[:47] + "..."
             except:
                 sample = "N/A"
             schema_info.append({
                 'Column': col,
                 'Type': dtype,
                 'Nulls': f"{null_count:,}",
                 'Null %': f"{null_pct:.1f}%",
                 'Unique': f"{unique_count:,}",
                 'Sample': sample
             })
```

```
schema_df = pd.DataFrame(schema_info)
display_table_with_html( schema_df, title="Table 1: Full Schema Overview (19 Columns)" )
```

Table 1: Full Schema Overview (19 Columns)

	Column	Туре	Nulls	Null %	Unique	Sample
0	cik	String	0	0.0%	10	0000001750
1	sentence	String	0	0.0%	96,465	ITEM 1.BUSINESS General AAR CORP. and its subsi
2	section	Int64	0	0.0%	20	0
3	labels	Struct({'1d': Int64, '30d': Int64, '5d': Int64})	0	0.0%	8	{'1d': 0, '30d': 0, '5d': 1}
4	filingDate	String	0	0.0%	181	2020-07-21
5	name	String	0	0.0%	10	AAR CORP
6	docID	String	0	0.0%	188	0000001750_10-K_2020
7	sentenceID	String	0	0.0%	200,000	0000001750_10- K_2020_section_1_0
8	sentenceCount	Int64	0	0.0%	200,000	1
9	tickers	List(String)	0	0.0%	10	['AIR']
10	exchanges	List(String)	0	0.0%	3	['NYSE']
11	entityType	String	0	0.0%	1	operating
12	sic	String	0	0.0%	10	3720
13	stateOfIncorporation	String	0	0.0%	5	DE
14	tickerCount	Int32	0	0.0%	1	1
15	acceptanceDateTime	String	0	0.0%	188	2020-07- 21T17:19:15.000Z
16	form	String	0	0.0%	1	10-K
17	reportDate	String	0	0.0%	91	2020-05-31
18	returns	Struct({'1d': Struct({'closePriceEndDate': Float64, 'closePriceStartDate': Float64, 'endDate': String, 'ret': Float64, 'startDate': String}), '30d':	0	0.0%	188	{'1d': {'closePriceEndDate': 19.010000228881836

Column Type Nulls Null Unique Sample

```
In [24]: # Table 2: Logical grouping of columns by purpose
         categories = {
             '/ Identifiers': ['cik', 'docID', 'sentenceID', 'name', 'tickers'],
             '    Text Content': ['sentence'],
             'ill Document Metadata': ['section', 'filingDate', 'reportDate', 'period', 'form', 'acceptanceDateTime'],
             ' Company Info': ['exchanges', 'entityType', 'sic', 'stateOfIncorporation', 'tickerCount'],
             '  ML Targets': ['labels', 'returns'],
             '    Derived/Counters': ['sentenceCount']
         # Build summary table
         cat_summary = []
         for category, cols in categories.items():
             for col in cols:
                 if col in df.columns:
                     dtype = str(df[col].dtype)
                     null_pct = (df[col].null_count() / len(df)) * 100
                     unique = df[col].n_unique()
                     cat_summary.append({
                          'Category': category,
                          'Column': col,
                          'Type': dtype,
                          'Null %': f"{null_pct:.1f}%",
                          'Unique': f"{unique:,}"
                     })
         cat_df = pd.DataFrame(cat_summary)
         display_table_with_html(
             cat_df,
```

```
title="Table 2: Column Categorization by Purpose"
)
```

Table 2: Column Categorization by Purpose

	Category	Column	Туре	Null %	Unique
0	Identifiers	cik	String	0.0%	10
1	Identifiers	docID	String	0.0%	188
2	Identifiers	sentenceID	String	0.0%	200,000
3	Identifiers	name	String	0.0%	10
4	Identifiers	tickers	List(String)	0.0%	10
5	Text Content	sentence	String	0.0%	96,465
6	Document Metadata	section	Int64	0.0%	20
7	Document Metadata	filingDate	String	0.0%	181
8	Document Metadata	reportDate	String	0.0%	91
9	Document Metadata	form	String	0.0%	1
10	Document Metadata	acceptanceDateTime	String	0.0%	188
11	Company Info	exchanges	List(String)	0.0%	3
12	Company Info	entityType	String	0.0%	1
13	Company Info	sic	String	0.0%	10
14	Company Info	stateOfIncorporation	String	0.0%	5
15	Company Info	tickerCount	Int32	0.0%	1
16	◎ ML Targets	labels	Struct({'1d': Int64, '30d': Int64, '5d': Int64})	0.0%	8
17	◎ ML Targets	returns	Struct({'1d': Struct({'closePriceEndDate': Float64, 'closePriceStartDate': Float64, 'endDate': String, 'ret': Float64, 'startDate': String}), '30d': Struct({'closePriceEndDate': Float64, 'closePriceStartDate': Float64, 'endDate': String, 'ret': Float64, 'startDate': String}), '5d':	0.0%	188

Category Column Type Null Unique

Struct({'closePriceEndDate': Float64, 'closePriceStartDate': Float64, 'endDate': String, 'ret': Float64, 'startDate': String})})

18 Derived/Counters

sentenceCount

Int64 0.0% 200.000

```
In [25]: # Answer Q2: Company distribution balance
         company_stats = (
             df.group_by("cik", "name")
             .agg([
                 pl.count("sentence").alias("sentence_count"),
                 pl.col("tickers").first().alias("tickers"),
                 pl.col("sic").first().alias("sic"),
                 pl.n_unique("docID").alias("num_filings"),
                 pl.col("reportDate").min().alias("earliest_report"),
                 pl.col("reportDate").max().alias("latest_report"),
             .sort("sentence_count", descending=True)
         # Convert to pandas for display
         company_stats_pd = company_stats.to_pandas()
         # Extract first ticker from list (clean display)
         company_stats_pd['ticker'] = company_stats_pd['tickers'].apply(lambda x: x[0] if isinstance(x, list) and len(x) > 0 els
         company_stats_pd = company_stats_pd.drop(columns=['tickers'])
         # Reorder columns for clarity
         company_stats_pd = company_stats_pd[['cik', 'name', 'ticker', 'sic', 'sentence_count', 'num_filings', 'earliest_report'
         display_table_with_html(
             company_stats_pd,
             title="Table 3: Company Statistics (Answers Q2: Distribution Balance)"
         # Quick summary
```

```
total_sentences = company_stats_pd['sentence_count'].sum()
avg_sentences = company_stats_pd['sentence_count'].mean()
std_sentences = company_stats_pd['sentence_count'].std()

print(f"\ni Distribution Summary:")
print(f" Total sentences: {total_sentences:,}")
print(f" Average per company: {avg_sentences:,.0f}")
print(f" Std deviation: {std_sentences:,.0f}")
print(f" Imbalance ratio: {company_stats_pd['sentence_count'].max() / company_stats_pd['sentence_count'].min():.2f}x
```

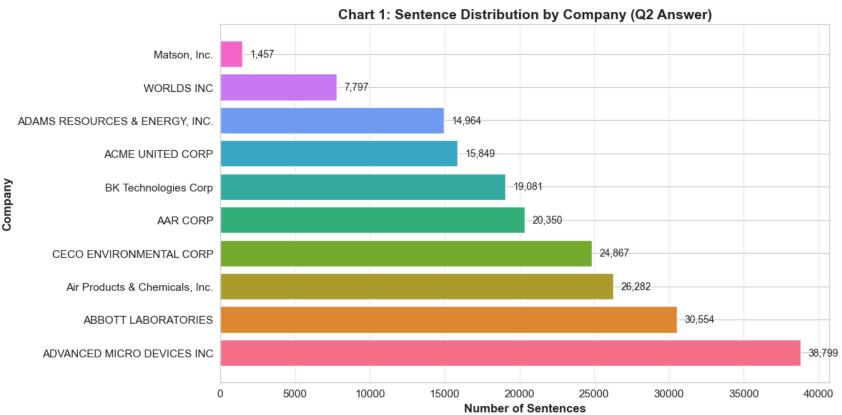
Table 3: Company Statistics (Answers Q2: Distribution Balance)

	cik	name	ticker	sic	sentence_count	num_filings	earliest_report	latest_report
0	0000002488	ADVANCED MICRO DEVICES INC	N/A	3674	38799	24	1993-12-26	2020-12-26
1	0000001800	ABBOTT LABORATORIES	N/A	2834	30554	25	1993-12-31	2020-12-31
2	0000002969	Air Products & Chemicals, Inc.	N/A	2810	26282	20	2001-09-30	2020-09-30
3	0000003197	CECO ENVIRONMENTAL CORP	N/A	3564	24867	17	2004-12-31	2020-12-31
4	0000001750	AAR CORP	N/A	3720	20350	21	1994-05-31	2020-05-31
5	0000002186	BK Technologies Corp	N/A	3663	19081	21	1995-12-31	2020-12-31
6	0000002098	ACME UNITED CORP	N/A	3420	15849	26	1995-12-31	2020-12-31
7	0000002178	ADAMS RESOURCES & ENERGY, INC.	N/A	5172	14964	19	2002-12-31	2020-12-31
8	0000001961	WORLDS INC	N/A	7372	7797	13	2008-12-31	2020-12-31
9	0000003453	Matson, Inc.	N/A	4400	1457	2	2019-12-31	2020-12-31

Distribution Summary:
Total sentences: 200,000
Average per company: 20,000
Std deviation: 10,874

Imbalance ratio: 26.63x (max/min)

```
In [26]: # Visualize company balance (bar chart)
fig, ax = plt.subplots(figsize=(12, 6))
companies = company_stats_pd['name'].tolist()
```



```
In [27]: # Answer Q3: Temporal distribution
         # Extract year from reportDate
         df_temporal = df.with_columns([
             pl.col("reportDate").str.strptime(pl.Date, "%Y-%m-%d").alias("report_date_parsed")
         ]).with_columns([
             pl.col("report_date_parsed").dt.year().alias("year")
         ])
         # Sentences per year
         year_stats = (
             df_temporal.group_by("year")
             .agg([
                 pl.count("sentence").alias("sentence_count"),
                 pl.n_unique("docID").alias("num_filings")
             1)
             .sort("year")
         year_stats_pd = year_stats.to_pandas()
         display_table_with_html(
             year_stats_pd,
             title="Table 4: Temporal Distribution by Year (Answers Q3: Date Range)"
         # Summary stats
         print(f" Earliest year: {year_stats_pd['year'].min()}")
                   Latest year: {year_stats_pd['year'].max()}")
         print(f"
                   Total span: {year_stats_pd['year'].max() - year_stats_pd['year'].min() + 1} years")
         print(f"
                   Average sentences/year: {year_stats_pd['sentence_count'].mean():,.0f}")
         print(f"
```

Table 4: Temporal Distribution by Year (Answers Q3: Date Range)

	year	sentence_count	num_filings
0	1993	853	2
1	1994	1257	3
2	1995	1135	3
3	1996	1211	3
4	1997	1510	3
5	1998	1365	3
6	1999	370	1
7	2000	1181	2
8	2001	1298	3
9	2002	6361	7
10	2003	6631	7
11	2004	7878	8
12	2005	8290	8
13	2006	9637	8
14	2007	9916	8
15	2008	9929	9
16	2009	10473	9
17	2010	10511	9
18	2011	10879	9
19	2012	9976	9
20	2013	9713	9
21	2014	10228	9

	year	sentence_count	num_filings
22	2015	10572	9
23	2016	11251	9
24	2017	11990	9
25	2018	11956	9
26	2019	11034	10
27	2020	12595	10

Temporal Coverage:
Earliest year: 1993
Latest year: 2020
Total span: 28 years

Average sentences/year: 7,143

```
In [28]: # Chart 2: Sentences over time
         fig, ax = plt.subplots(figsize=(14, 6))
         years = year_stats_pd['year'].tolist()
         counts = year_stats_pd['sentence_count'].tolist()
         ax.plot(years, counts, marker='o', linewidth=2, markersize=6, color='#2E86AB')
         ax.fill between(years, counts, alpha=0.3, color='#2E86AB')
         ax.set xlabel('Year', fontsize=12, fontweight='bold')
         ax.set ylabel('Number of Sentences', fontsize=12, fontweight='bold')
         ax.set title('Chart 2: Temporal Coverage (Q3 Answer: Are filings evenly spread?)', fontsize=14, fontweight='bold')
         ax.grid(True, alpha=0.3)
         # Add annotations for min/max years
         \max idx = counts.index(\max(counts))
         min idx = counts.index(min(counts))
         ax.annotate(f'Peak: {max(counts):,}', xy=(years[max_idx], counts[max_idx]),
                     xytext=(10, 20), textcoords='offset points',
                     bbox=dict(boxstyle='round', fc='yellow', alpha=0.7),
                     arrowprops=dict(arrowstyle='->', connectionstyle='arc3,rad=0'))
```

```
plt.tight_layout()
plt.show()
```

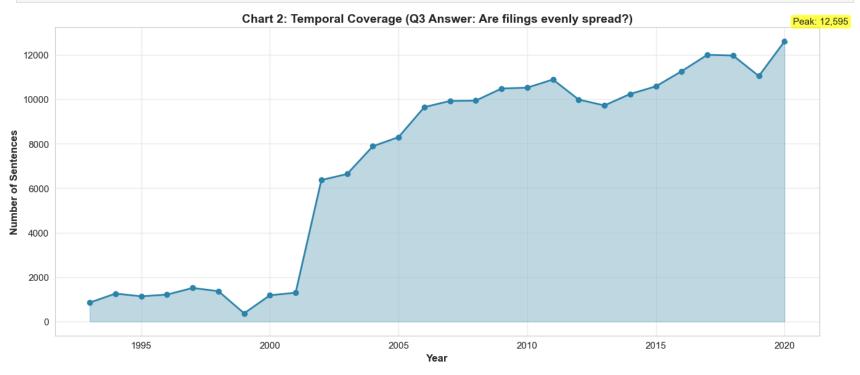


Table 5: Section Code Distribution (Answers Q1: What are the 20 codes?)

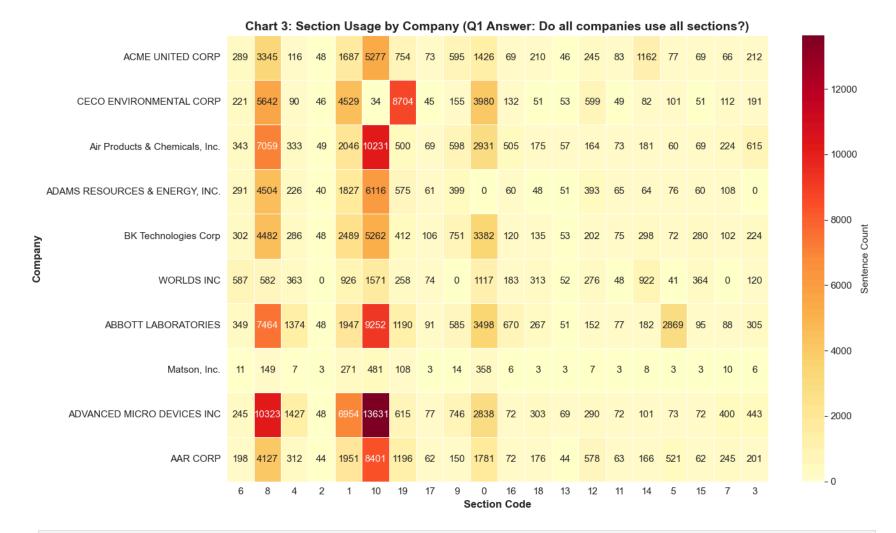
	section	sentence_count	num_companies	percentage
0	0	21311	9	10.66
1	1	24627	10	12.31
2	2	374	9	0.19
3	3	2317	9	1.16
4	4	4534	10	2.27
5	5	3893	10	1.95
6	6	2836	10	1.42
7	7	1355	9	0.68
8	8	47677	10	23.84
9	9	3993	9	2.00
10	10	60256	10	30.13
11	11	608	10	0.30
12	12	2906	10	1.45
13	13	479	10	0.24
14	14	3166	10	1.58
15	15	1125	10	0.56
16	16	1889	10	0.94
17	17	661	10	0.33
18	18	1681	10	0.84
19	19	14312	10	7.16

Section Analysis:

Expected sections: 0-9 (10 codes)
Actual sections: 0 to 19 (20 codes)

Section coverage: 9-10 companies per section

```
In [30]: # Cross-tab: Which companies use which sections?
         section_company_cross = (
             df.group_by(["section", "name"])
             .agg(pl.count("sentence").alias("count"))
             .pivot(index="name", columns="section", values="count")
             .fill_null(0)
         section_company_pd = section_company_cross.to_pandas().set_index('name')
         # Heatmap
         fig, ax = plt.subplots(figsize=(14, 8))
         sns.heatmap(section_company_pd, annot=True, fmt='.0f', cmap='YlOrRd',
                     linewidths=0.5, cbar_kws={'label': 'Sentence Count'}, ax=ax)
         ax.set_title('Chart 3: Section Usage by Company (Q1 Answer: Do all companies use all sections?)',
                      fontsize=14, fontweight='bold')
         ax.set_xlabel('Section Code', fontsize=12, fontweight='bold')
         ax.set_ylabel('Company', fontsize=12, fontweight='bold')
         plt.tight_layout()
         plt.show()
```



```
])
# Statistics
token stats = token counts.select([
    pl.col("token count").mean().alias("mean"),
    pl.col("token_count").median().alias("median"),
    pl.col("token count").quantile(0.95).alias("p95"),
    pl.col("token count").max().alias("max"),
    pl.col("token count").min().alias("min")
1).to pandas()
print(" Token Length Statistics:")
print(f"
          Mean: {token stats['mean'][0]:.1f} tokens")
print(f"
           Median: {token stats['median'][0]:.0f} tokens")
           95th percentile: {token stats['p95'][0]:.0f} tokens")
print(f"
print(f"
          Max: {token stats['max'][0]:.0f} tokens")
# Histogram
fig, ax = plt.subplots(figsize=(12, 6))
token data = token counts.select("token count").to pandas()['token count']
ax.hist(token data, bins=100, color='#2E86AB', alpha=0.7, edgecolor='black')
ax.axvline(token stats['mean'][0], color='red', linestyle='--', linewidth=2, label=f"Mean: {token stats['mean'][0]:.1f}
ax.axvline(token stats['p95'][0], color='orange', linestyle='--', linewidth=2, label=f"P95: {token stats['p95'][0]:.0f}
ax.set xlabel('Tokens per Sentence', fontsize=12, fontweight='bold')
ax.set_ylabel('Frequency', fontsize=12, fontweight='bold')
ax.set title('Chart 4: Token Length Distribution (For Chunk Size Decision)', fontsize=14, fontweight='bold')
ax.legend()
ax.grid(axis='y', alpha=0.3)
plt.tight layout()
plt.show()
# Chunk size recommendation
print(f"\n ? Chunk Size Recommendation:")
print(f" 3-sentence chunks: ~{token stats['mean'][0] * 3:.0f} tokens (avg)")
print(f"
           5-sentence chunks: ~{token stats['mean'][0] * 5:.0f} tokens (avg)")
print(f" Embedding limit (512 tokens): Fits ~{512 / token stats['mean'][0]:.1f} sentences")
```

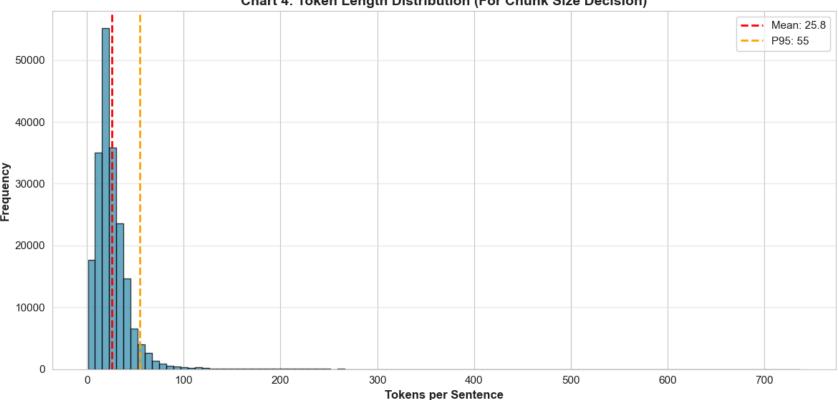
Token Length Statistics:

Mean: 25.8 tokens Median: 22 tokens

95th percentile: 55 tokens

Max: 737 tokens

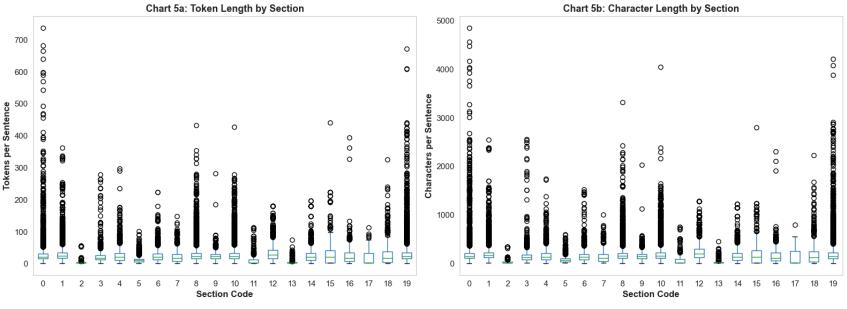
Chart 4: Token Length Distribution (For Chunk Size Decision)



P Chunk Size Recommendation:

```
3-sentence chunks: ~77 tokens (avg)
5-sentence chunks: ~129 tokens (avg)
Embedding limit (512 tokens): Fits ~19.9 sentences
```

```
# Boxplot for token count by section
fig, axes = plt.subplots(1, 2, figsize=(16, 6))
# Token count boxplot
section_text_pd.boxplot(column='token_count', by='section', ax=axes[0], grid=False)
axes[0].set_xlabel('Section Code', fontsize=12, fontweight='bold')
axes[0].set_ylabel('Tokens per Sentence', fontsize=12, fontweight='bold')
axes[0].set_title('Chart 5a: Token Length by Section', fontsize=13, fontweight='bold')
axes[0].get_figure().suptitle('') # Remove default title
# Char count boxplot
section_text_pd.boxplot(column='char_count', by='section', ax=axes[1], grid=False)
axes[1].set_xlabel('Section Code', fontsize=12, fontweight='bold')
axes[1].set_ylabel('Characters per Sentence', fontsize=12, fontweight='bold')
axes[1].set_title('Chart 5b: Character Length by Section', fontsize=13, fontweight='bold')
axes[1].get_figure().suptitle('')
plt.tight_layout()
plt.show()
```



```
In [33]: # Detailed stats by section
section_density = (
```

```
df.with columns([
        pl.col("sentence").str.len_chars().alias("char_count"),
        pl.col("sentence").str.split(" ").list.len().alias("token_count")
    .group_by("section")
    .agg([
        pl.col("char count").mean().alias("avg chars"),
        pl.col("token_count").mean().alias("avg_tokens"),
        pl.col("char_count").median().alias("median_chars"),
        pl.col("token_count").median().alias("median_tokens"),
        pl.col("char_count").max().alias("max_chars"),
        pl.col("token count").max().alias("max tokens"),
        pl.count("sentence").alias("sentence_count")
    .sort("section")
section_density_pd = section_density.to_pandas()
section_density_pd = section_density_pd.round(1)
display_table_with_html(
    section_density_pd,
   title="Table 6: Text Density by Section (Item 7 vs Item 1A Comparison)"
# Highlight extremes
densest_section = section_density_pd.loc[section_density_pd['avg_tokens'].idxmax(), 'section']
sparsest_section = section_density_pd.loc[section_density_pd['avg_tokens'].idxmin(), 'section']
print(f"\n Q Density Analysis:")
print(f" Densest section: {densest_section} ({section_density_pd.loc[section_density_pd['avg_tokens'].idxmax(), 'avg_
          Sparsest section: {sparsest_section} ({section_density_pd.loc[section_density_pd['avg_tokens'].idxmin(), 'av
print(f"
```

Table 6: Text Density by Section (Item 7 vs Item 1A Comparison)

	section	avg_chars	avg_tokens	median_chars	median_tokens	max_chars	max_tokens	sentence_count
0	0	174.2	25.4	145.0	21.0	4850	737	21311
1	1	182.5	27.7	160.0	24.0	2544	362	24627
2	2	27.6	4.5	8.0	2.0	342	55	374
3	3	140.4	20.7	116.0	18.0	2557	279	2317
4	4	152.0	23.6	131.0	21.0	1743	296	4534
5	5	74.6	11.6	68.0	10.0	596	101	3893
6	6	149.2	23.5	127.0	20.0	1515	223	2836
7	7	129.7	20.5	112.0	18.0	1002	147	1355
8	8	168.8	26.0	148.0	23.0	3317	433	47677
9	9	146.2	22.7	140.0	22.0	2031	281	3993
10	10	172.1	26.2	150.0	23.0	4045	428	60256
11	11	67.2	10.3	8.0	2.0	751	113	608
12	12	229.8	33.4	199.0	28.0	1273	179	2906
13	13	27.0	4.7	17.0	3.0	451	74	479
14	14	162.6	25.1	127.0	20.0	1223	197	3166
15	15	179.3	27.6	129.0	21.0	2800	440	1125
16	16	158.9	24.5	113.0	17.0	2298	394	1889
17	17	117.4	17.4	8.0	2.0	795	112	661
18	18	165.8	25.7	114.0	18.0	2222	325	1681
19	19	183.7	28.4	140.0	22.0	4214	672	14312

Density Analysis:

Densest section: 12 (33.4 avg tokens) Sparsest section: 2 (4.5 avg tokens)

```
In [34]: # Identify extreme outliers (likely tables)
         outliers = df.with columns([
             pl.col("sentence").str.len_chars().alias("char_count")
         1).filter(
             pl.col("char_count") > 1000 # Threshold for "table-like" text
         ).select(["section", "char_count", "sentence"]).head(5)
         outliers pd = outliers.to pandas()
         print("  Outlier Examples (Likely Tables/Lists):")
         for idx, row in outliers_pd.iterrows():
             print(f"\n Section {row['section']} | {row['char_count']} chars")
             print(f" Preview: {row['sentence'][:150]}...")
        Outlier Examples (Likely Tables/Lists):
           Section 1 | 1174 chars
           Preview: Consequently, we are subject to a variety of risks that are specific to international operations, including
        the following: ●military conflicts, civil ...
           Section 10 | 1022 chars
           Preview: Sales by segment for these customers are as follows: AAR CORP. AND SUBSIDIARIES NOTES TO CONSOLIDATED FINANC
        IAL STATEMENTS (Continued) (Dollars in mil...
           Section 12 | 1040 chars
           Preview: The Company's internal control over financial reporting is a process designed by, or under the supervision o
        f, our Chief Executive Officer and Chief F...
           Section 19 | 1330 chars
           Preview: 4.3 Description of Capital Stock (filed herewith) 4.4 Rights Agreement, dated as of March 30, 2020, by and b
        etween AAR CORP. and Computershare Trust C...
           Section 19 | 1850 chars
           Preview: Material Contracts 10.1* Amended and Restated AAR CORP. Stock Benefit Plan effective October 1, 2001 (incorp
        orated by reference to Exhibit 10.1 to the...
In [ ]:
```

Next EDA acts:

- 1. Top n-grams by section
- 2. Approx. duplication by section (simple, fast SimHash-style approach; no extra deps)
- 3. KPI signal scan (where numbers/units live, by section)
- 4. Section label suggestions (human labels from n-gram signatures)

```
In [39]: # --- imports & config ---
         import os, re, hashlib, random
         import numpy as np
         import pandas as pd
         from collections import Counter, defaultdict
         from sklearn.feature_extraction.text import TfidfVectorizer
         # path to your parquet sample (adjust if needed)
         PARQUET_PATH = os.getenv(
             "FINRAG_PARQUET",
             r"D:\JoelDesktop folds_24\NEU FALL2025\MLops IE7374 Project\finrag-insights-mlops\data\exports\sec_filings_small_fu
         SAVE_CSV = True # set True to save outputs in data/eda/
         os.makedirs("data/eda", exist_ok=True)
         # --- Load df ---
         df = pd.read_parquet(PARQUET_PATH)
         df["sentence"] = df["sentence"].astype(str)
         df["section"] = df["section"].astype(int)
         # --- tiny helpers ---
         tok_re = re.compile(r"[A-Za-z0-9%$\.]+")
         def toks(s: str): return tok_re.findall((s or "").lower())
         def shingles(tokens, k=5):
             if len(tokens) < k: return [" ".join(tokens)] if tokens else []</pre>
             return [" ".join(tokens[i:i+k]) for i in range(len(tokens)-k+1)]
         def simhash64(tokens, bits=64):
             # simple simhash on tokens
             v = [0]*bits
             for t in tokens:
                 h = int(hashlib.md5(t.encode()).hexdigest(), 16)
                 for b in range(bits):
```

```
v[b] += 1 if (h>>b)&1 else -1
             x = 0
             for b in range(bits):
                 if v[b] >= 0: x = (1 << b)
             return x
         def hamming(a, b): return (a ^ b).bit_count()
In [40]: # params
         NGRAM_RANGE = (1, 2)
         TOP_K = 25
         MIN DF = 5
         MAX DF = 0.6
         SAMPLE PER SECTION = 12000 # keep fast; adjust as needed
         rows = []
         for sec, grp in df.groupby("section"):
             sub = grp["sentence"]
             if len(sub) == 0: continue
             if len(sub) > SAMPLE PER SECTION:
                 sub = sub.sample(SAMPLE PER SECTION, random state=13)
             vect = TfidfVectorizer(tokenizer=toks, ngram_range=NGRAM_RANGE, min_df=MIN_DF, max_df=MAX_DF, lowercase=True)
             try:
                 X = vect.fit transform(sub.values)
             except ValueError:
                 continue
             vocab = np.array(sorted(vect.vocabulary_, key=vect.vocabulary_.get))
             scores = np.asarray(X.mean(axis=0)).ravel()
             top_idx = np.argsort(scores)[::-1][:TOP_K]
             for rank, j in enumerate(top_idx, 1):
                 rows.append({"section": sec, "ngram": vocab[j], "score": float(scores[j]), "rank": rank})
         ngrams_df = pd.DataFrame(rows).sort_values(["section","rank"]).reset_index(drop=True)
         display(ngrams_df.head(20))
         if SAVE CSV:
             path = "data/eda/top_ngrams_by_section.csv"
             ngrams_df.to_csv(path, index=False)
             print("saved:", path)
```

saved: data/eda/top_ngrams_by_section.csv

```
In [41]: # params
         NGRAM RANGE = (1, 2)
         TOP K = 25
         MIN DF = 5
         MAX DF = 0.6
         SAMPLE PER SECTION = 12000 # keep fast; adjust as needed
         rows = []
         for sec, grp in df.groupby("section"):
             sub = grp["sentence"]
             if len(sub) == 0: continue
             if len(sub) > SAMPLE PER SECTION:
                 sub = sub.sample(SAMPLE PER SECTION, random state=13)
             vect = TfidfVectorizer(tokenizer=toks, ngram range=NGRAM RANGE, min df=MIN DF, max df=MAX DF, lowercase=True)
             try:
                 X = vect.fit transform(sub.values)
             except ValueError:
                 continue
             vocab = np.array(sorted(vect.vocabulary , key=vect.vocabulary .get))
             scores = np.asarray(X.mean(axis=0)).ravel()
             top idx = np.argsort(scores)[::-1][:TOP K]
             for rank, j in enumerate(top idx, 1):
                 rows.append({"section": sec, "ngram": vocab[j], "score": float(scores[j]), "rank": rank})
         ngrams df = pd.DataFrame(rows).sort values(["section","rank"]).reset index(drop=True)
         display(ngrams df.head(20))
         if SAVE CSV:
             path = "data/eda/top ngrams by section.csv"
             ngrams df.to csv(path, index=False)
             print("saved:", path)
```

saved: data/eda/top_ngrams_by_section.csv

```
In [42]: # params
         SAMPLE PER SECTION DUP = 6000
         SHINGLE K = 5
         PREFIX_BITS = 16 # bucket by first 16 bits
         HAMMING THRESHOLD = 3 # pairs within a bucket considered near-dup if <= 3 bits
         PAIR CHECK LIMIT = 8000 # cap pair checks per bucket to keep fast
         dup rows = []
         for sec, grp in df.groupby("section"):
             texts = grp["sentence"]
             if len(texts) == 0:
                 dup_rows.append({"section": sec, "n_sampled": 0, "n_near_dupes": 0, "dup_rate": 0.0})
                 continue
             if len(texts) > SAMPLE PER SECTION DUP:
                 texts = texts.sample(SAMPLE_PER_SECTION_DUP, random_state=17)
             # build simhash buckets
             buckets = defaultdict(list)
             for s in texts:
                 tokens = toks(s)
                 sh = shingles(tokens, k=SHINGLE_K) or tokens
                 hv = simhash64(sh or [""])
                 prefix = hv >> (64 - PREFIX BITS)
                 buckets[prefix].append(hv)
             n = len(texts)
             near dupes = 0
             for _, vals in buckets.items():
                 m = len(vals)
                 if m <= 1: continue</pre>
                 # sample limited pairs inside bucket
                 idxs = list(range(m))
                 checked = 0
                 # simple random sampling of pairs
                 for i in range(m):
```

```
for j in range(i+1, m):
                          if checked >= PAIR_CHECK_LIMIT: break
                          if hamming(vals[i], vals[j]) <= HAMMING_THRESHOLD:</pre>
                              near dupes += 1
                          checked += 1
                      if checked >= PAIR_CHECK_LIMIT: break
             dup_rows.append({"section": sec, "n_sampled": int(n), "n_near_dupes": int(near_dupes), "dup_rate": round(near_dupes
         dup_df = pd.DataFrame(dup_rows).sort_values("section").reset_index(drop=True)
         display(dup df.head(20))
         if SAVE CSV:
             path = "data/eda/duplication_by_section.csv"
             dup_df.to_csv(path, index=False)
             print("saved:", path)
              Loading ITables v2.5.2 from the init_notebook_mode cell... (need help?)
        saved: data/eda/duplication_by_section.csv
In [43]: # KPI signal scan (what % of sentences carry numeric/scale cues)
         PATTERNS = {
             "currency": re.compile(r"(\$|usd|us\$)\s?\d[\d,]*\.?\d*", re.I),
             "percent": re.compile(r"\b\d{1,3}\.\d+%|\b\d{1,3}\"),
             "eps": re.compile(r"\beps\b|\bearnings per share\b", re.I),
             "units": re.compile(r"in (thousands|millions|billions)", re.I),
             "yoy": re.compile(r"\b(year[- ]over[- ]year|yoy)\b", re.I),
             "growth": re.compile(r"\b(increase[d]?|decrease[d]?|grew|declined)\b", re.I),
         SAMPLE_PER_SECTION_KPI = 15000
         kpi rows = []
         for sec, grp in df.groupby("section"):
             texts = grp["sentence"]
             if len(texts) == 0: continue
             if len(texts) > SAMPLE_PER_SECTION_KPI:
                 texts = texts.sample(SAMPLE_PER_SECTION_KPI, random_state=23)
             n = len(texts)
```

```
counts = Counter()
   for s in texts:
        s low = s.lower()
       for k, pat in PATTERNS.items():
            if pat.search(s_low): counts[k] += 1
    row = {"section": sec, "n": n}
   for k in PATTERNS.keys(): row[f"pct_{k}"] = round(100.0 * counts[k] / n, 2)
    kpi rows.append(row)
kpi_df = pd.DataFrame(kpi_rows).sort_values("section").reset_index(drop=True)
display(kpi_df.head(20))
if SAVE CSV:
    path = "data/eda/kpi signal scan by section.csv"
   kpi df.to csv(path, index=False); print("saved:", path)
# Section label suggestions from top n-grams
LABEL HINTS = [
    ("Business / Overview", ("business", "segment", "customers", "products", "overview")),
   ("Risk Factors", ("risk", "risks", "adverse", "uncertain")),
   ("MD&A (Discussion & Analysis)", ("management", "discussion", "analysis", "operations", "md&a")),
   ("Financial Statements / Notes", ("consolidated", "statement", "net", "sales", "revenue", "income", "note")),
   ("Controls & Procedures", ("controls", "disclosure", "procedures", "internal control")),
   ("Legal / Exhibits", ("exhibit", "agreement", "incorporated", "reference", "schedule")),
label rows = []
for sec, grp in ngrams_df.groupby("section"):
    top_terms = grp.sort_values("rank").head(15)["ngram"].tolist()
   best_label, best_hits = "Unclear", []
   best count = -1
   for label, keys in LABEL HINTS:
        hits = [k for k in keys if any(k in t for t in top_terms)]
        if len(hits) > best count:
            best count, best label, best hits = len(hits), label, hits
   label_rows.append({"section": sec, "guess_label": best_label if best_count>0 else "Unclear", "support_terms": ", ".
labels_df = pd.DataFrame(label_rows).sort_values("section").reset_index(drop=True)
display(labels df)
if SAVE_CSV:
```

```
path = "data/eda/section_label_suggestions.csv"
             labels_df.to_csv(path, index=False); print("saved:", path)
                Loading ITables v2.5.2 from the init_notebook_mode cell... (need help?)
       saved: data/eda/kpi_signal_scan_by_section.csv
                Loading ITables v2.5.2 from the init_notebook_mode cell... (need help?)
       saved: data/eda/section_label_suggestions.csv
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