

# Visit Limit Algorithm - Organized Document

## Presenting an Automated Solution for Comparative Analysis of Visit Limits

### ### Benefits of the Algorithm

1. Efficiency: Automates the analysis, reducing the need for manual review.
2. Scalability: Handles large datasets typical of insurance companies.
3. Cost-Effective: Minimizes labor costs associated with manual reviews.
4. Accuracy: Ensures consistent and reliable results by eliminating human error.

### Algorithm Overview

The algorithm consists of three main steps:

1. Database Preparation
2. Data Extraction
3. Parallel Processing

#### Step 1: Database Preparation

Indexing:

Purpose: Speed up data retrieval by creating indexes on key columns.

Relevant Columns:

- Claim Number: ClaimNumber
- Date of Service: DateOfService
- Procedure Code (CPT): CPT

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- Year: Year

Partitioning:

Purpose: Divide the database into smaller, more manageable pieces based on criteria like year.

Relevant Columns:

- Year: Year

### Step 2: Data Extraction

Optimized Queries:

Purpose: Efficiently extract relevant data segments.

Relevant Columns:

- Year: Year

- State: State

- Individual: Individual

- Date of Service: DateOfService

- Claim Number: ClaimNumber

- Billed Amount: BilledAmount

- Payment: Payment

- Deductible: Deductible

- Reason Code: ReasonCode

- Plan Category: PlanCategory

- Product: Product

- Plan Type: PlanType

- Mental Health/Substance Use Disorder ICD-10 Code: MH\_SUD\_ICD10

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- Medical/Surgical ICD-10 Code: Med\_Surg\_ICD10
- Procedure Code (CPT): CPT
- Number of Visits in Year 1: VisitsYear1
- Number of Visits in Year 2: VisitsYear2
- Number of Visits in Year 3: VisitsYear3

### Step 3: Parallel Processing with Apache Spark

Apache Spark:

Purpose: A powerful tool designed for processing large datasets in parallel.

Relevant Columns:

- Year: Year
- State: State
- Individual: Individual
- Date of Service: DateOfService
- Claim Number: ClaimNumber
- Billed Amount: BilledAmount
- Payment: Payment
- Deductible: Deductible
- Reason Code: ReasonCode
- Plan Category: PlanCategory
- Product: Product
- Plan Type: PlanType
- Mental Health/Substance Use Disorder ICD-10 Code: MH\_SUD\_ICD10
- Medical/Surgical ICD-10 Code: Med\_Surg\_ICD10

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- Procedure Code (CPT): CPT
- Number of Visits in Year 1: VisitsYear1
- Number of Visits in Year 2: VisitsYear2
- Number of Visits in Year 3: VisitsYear3

### Explanation of Flags with Column Descriptions

#### 1. Flag A: Reprocessed Claims

Purpose: Identifies claims that have been processed multiple times.

Relevant Columns:

- Claim Number: ClaimNumber
- Date of Service: DateOfService

Example: If a patient has multiple claim lines for the same date of service, it may indicate reprocessing due to visit limits.

#### 2. Flag B: Front-Loading Services

Purpose: Detects if services are predominantly used early in the year.

Relevant Columns:

- Procedure Code (CPT): CPT
- Date of Service: DateOfService

Example: If all therapy sessions for a patient occur between January and June, it suggests front-loading to use benefits before limits apply.

#### 3. Flag C: Diagnosis Switching

Purpose: Identifies if diagnoses switch from mental health to medical conditions.

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Relevant Columns:

- Individual: Individual
- Mental Health/Substance Use Disorder ICD-10 Code: MH\_SUD\_ICD10
- Medical/Surgical ICD-10 Code: Med\_Surg\_ICD10

Example: A patient initially diagnosed with an eating disorder switches to diabetes management, potentially to bypass visit limits on mental health services.

### 4. Flags D, E, F, G: Consistent Visit Counts

Purpose: Detects consistent numbers of approved visits over multiple years.

Relevant Columns:

- Individual: Individual
- Number of Visits in Year 1: VisitsYear1
- Number of Visits in Year 2: VisitsYear2
- Number of Visits in Year 3: VisitsYear3

Example: If a patient consistently receives 8 approved therapy sessions per year, it may indicate a visit limit.

## Example Implementation

Database Preparation

Indexing Example:

Creating an index on ClaimNumber for faster retrieval:

```
```sql
```

```
CREATE INDEX idx_claim_number ON Claims(ClaimNumber);
```

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...

Partitioning Example:

Partitioning the database by year:

```
```sql
```

```
CREATE TABLE ClaimsPartitioned PARTITION BY RANGE (Year) (  
PARTITION p2021 VALUES LESS THAN (2022),  
PARTITION p2022 VALUES LESS THAN (2023),  
PARTITION p2023 VALUES LESS THAN (2024)  
) AS SELECT * FROM Claims;
```

...

Data Extraction

Optimized Query Example:

Extracting data for the year 2021:

```
```python
```

```
import pandas as pd
```

```
import sqlite3
```

```
def extract_data(year):
```

```
    conn = sqlite3.connect('cigna_claims.db')
```

```
    query = f"""
```

```
    SELECT * FROM ClaimsPartitioned WHERE Year = {year};
```

```
    """
```

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```
data = pd.read_sql_query(query, conn)

conn.close()

return data
```

```
data_2021 = extract_data(2021)
```

```
...
```

### Parallel Processing with Apache Spark

#### Initializing Spark:

```
```python

from pyspark.sql import SparkSession

spark = SparkSession.builder \

    .appName("Claims Analysis Optimization") \

    .config("spark.sql.shuffle.partitions", "8") \

    .getOrCreate()

```
```

#### Loading and Analyzing Data:

##### Loading data into Spark and performing analysis:

```
```python

claims_df = spark.read.csv("claims_data.csv", header=True, inferSchema=True)
```

```
# Flag A: Identify reprocessed claims
```

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```
reprocessed_claims = claims_df.groupBy("ClaimNumber", "DateOfService").count().filter("count > 1")
```

```
# Save results
```

```
reprocessed_claims.write.csv("reprocessed_claims_output.csv")
```

```
spark.stop()
```

```
...
```