

| SQL Operation | Pandas Equivalent | PySpark Equivalent | Example |
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| SELECT | df['column_name'] | df.select('column_name') | Select the 'Age' column: SQL: SELECT Age FROM table; Pandas: df['Age'] PySpark: df.select('Age') |
| SELECT Multiple Columns | df[['column1', 'column2']] | df.select('column1', 'column2') | Select 'Age' and 'Name' columns: SQL: SELECT Age, Name FROM table; Pandas: df[['Age', 'Name']] PySpark: df.select('Age', 'Name') |
| SELECT DISTINCT | df['column'].drop_duplicates() | df.select('column').distinct() | Select unique country names: SQL: SELECT DISTINCT Country FROM table; Pandas: df['Country'].drop_duplicates() PySpark: df.select('Country').distinct() |
| WHERE | df[df['column'] condition] | df.filter(df['column'] condition) | Select rows where age is greater than 30: SQL: SELECT * FROM table WHERE Age > 30; Pandas: df[df['Age'] > 30] PySpark: df.filter(df['Age'] > 30) |
| ORDER BY | df.sort_values('column') | df.orderBy('column') | Sort data by 'Age' column: SQL: SELECT * FROM table ORDER BY Age; Pandas: df.sort_values('Age') PySpark: df.orderBy('Age') |
| GROUP BY | df.groupby('column').mean() | df.groupBy('column').agg({'column': 'mean'}) | Calculate mean for each 'Country': SQL: SELECT Country, AVG(Age) FROM table GROUP BY Country; Pandas: df.groupby('Country')['Age'].mean() PySpark: df.groupBy('Country').agg({'Age': 'avg'}) |

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| JOIN | pd.merge(df1, df2, on='column', how='type') | df1.join(df2, on='column', how='type') | Inner join on 'ID': SQL: SELECT * FROM table1 INNER JOIN table2 ON table1.ID = table2.ID; Pandas: pd.merge(df1, df2, on='ID', how='inner') PySpark: df1.join(df2, on='ID', how='inner') |
| INSERT INTO | df.append(new_row, ignore_index=True) | df.union(spark.createDataFrame([new_row])) | Insert a new row: SQL: INSERT INTO table (Name, Age) VALUES ('John', 30); Pandas: df.append({'Name': 'John', 'Age': 30}, ignore_index=True) PySpark: df.union(spark.createDataFrame([{'Name': 'John', 'Age': 30}])) |
| UPDATE | df.loc[df['column'] condition, 'column'] = new_value | df.withColumn('column', F.when(df['column'] condition, new_value).otherwise(df['column'])) | Update 'Category' based on condition: SQL: UPDATE table SET Category = 'Senior' WHERE Age > 30; Pandas: df.loc[df['Age'] > 30, 'Category'] = 'Senior' PySpark: df.withColumn('Category', F.when(df['Age'] > 30, 'Senior').otherwise(df['Category'])) |
| DELETE FROM | df = df[~(df['column'] condition)] | df.filter(df['column'] condition) | Delete rows where age is less than 18: SQL: DELETE FROM table WHERE Age < 18; Pandas: df = df[~(df['Age'] < 18)] PySpark: df.filter(df['Age'] >= 18) |
| LIMIT | df.head(n) | df.limit(n) | Select the first 5 rows: SQL: SELECT * FROM table LIMIT 5; Pandas: df.head(5) PySpark: df.limit(5) |

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| COUNT | df['column'].count() () | df.select(F.count('column')) | Count non-null entries: SQL: SELECT COUNT(Age) FROM table; Pandas: df['Age'].count() PySpark: df.select(F.count('Age')) |
| SUM | df['column'].sum() | df.select(F.sum('column')) | Sum of 'Sales' column: SQL: SELECT SUM(Sales) FROM table; Pandas: df['Sales'].sum() PySpark: df.select(F.sum('Sales')) |
| AVG (Average) | df['column'].mean() () | df.select(F.avg('column')) | Average of 'Price' column: SQL: SELECT AVG(Price) FROM table; Pandas: df['Price'].mean() PySpark: df.select(F.avg('Price')) |
| MIN (Minimum) | df['column'].min() | df.select(F.min('column')) | Minimum value of 'Age': SQL: SELECT MIN(Age) FROM table; Pandas: df['Age'].min() PySpark: df.select(F.min('Age')) |
| MAX (Maximum) | df['column'].max() | df.select(F.max('column')) | Maximum value of 'Age': SQL: SELECT MAX(Age) FROM table; Pandas: df['Age'].max() PySpark: df.select(F.max('Age')) |
| HAVING | df.groupby('column').filter(lambda x: condition) | df.groupBy('column').agg(F.sum('other_column').alias('total')).filter('total > value') | Filter groups with total sales greater than 1000: SQL: SELECT Department, SUM(Sales) FROM table GROUP BY Department HAVING SUM(Sales) > 1000; Pandas: df.groupby('Department').filter(lambda x: x['Sales'].sum() > 1000) PySpark: df.groupBy('Department').agg(F.sum('Sales').alias('total_sales')).filter('total_sales > 1000') |

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| CONCATENATE | df['new_column'] = df['column1'] + df['column2'] | df.withColumn('new_column', F.concat_ws(' ', df['column1'], df['column2'])) | Concatenate 'FirstName' and 'LastName': SQL: `SELECT FirstName |
| LIKE (Pattern Match) | df[df['column'].str.contains('pattern')] | df.filter(df['column'].like('%pattern%')) | Select rows where 'Name' contains 'John': SQL: SELECT * FROM table WHERE Name LIKE '%John%'; Pandas: df[df['Name'].str.contains('John')] PySpark: df.filter(df['Name'].like('%John%')) |
| IN (List) | df[df['column'].isin(['value1', 'value2'])] | df.filter(df['column'].isin(['value1', 'value2'])) | Select rows where 'Country' is 'USA' or 'Canada': SQL: SELECT * FROM table WHERE Country IN ('USA', 'Canada'); Pandas: df[df['Country'].isin(['USA', 'Canada'])] PySpark: df.filter(df['Country'].isin(['USA', 'Canada'])) |
| BETWEEN | df[df['column'].between(value1, value2)] | df.filter(df['column'].between(value1, value2)) | Select rows where 'Age' is between 18 and 30: SQL: SELECT * FROM table WHERE Age BETWEEN 18 AND 30; Pandas: df[df['Age'].between(18, 30)] PySpark: df.filter(df['Age'].between(18, 30)) |
| CASE WHEN THEN ELSE | df['column'].apply(lambda x: 'value_if_true' if condition else 'value_if_false') | df.withColumn('new_column', F.when(df['column'] condition, 'value_if_true').otherwise('value_if_false')) | Assign 'Adult' or 'Minor' based on 'Age': SQL: SELECT CASE WHEN Age >= 18 THEN 'Adult' ELSE 'Minor' END AS Category FROM table; Pandas: df['Category'] = df['Age'].apply(lambda x: 'Adult' if x >= 18 else 'Minor') PySpark: |

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| | | | df.withColumn('Category', F.when(df['Age'] >= 18, 'Adult').otherwise('Minor')) |
| SUBSTRING | df['column'].str.slice(start, end) | df.withColumn('new_column', F.substring('column', start, end)) | Extract first letter from 'Name': SQL: SELECT SUBSTRING(Name, 1, 1) AS Initials FROM table; Pandas: df['Initials'] = df['Name'].str.slice(0, 1) PySpark: df.withColumn('Initials', F.substring('Name', 1, 1)) |
| LENGTH | df['column'].str.len() | df.withColumn('new_column', F.length('column')) | Length of each string in 'Name': SQL: SELECT LENGTH(Name) AS NameLength FROM table; Pandas: df['NameLength'] = df['Name'].str.len() PySpark: df.withColumn('NameLength', F.length('Name')) |
| REPLACE | df['column'].str.replace('old', 'new') | df.withColumn('column', F.regexp_replace('column', 'old', 'new')) | Remove dashes from 'Phone': SQL: SELECT REPLACE(Phone, '-', '') AS Phone FROM table; Pandas: df['Phone'] = df['Phone'].str.replace('-', '') PySpark: df.withColumn('Phone', F.regexp_replace('Phone', '-', '')) |
| DISTINCT COUNT | df['column'].nunique() | df.select(F.countDistinct('column')) | Count unique countries: SQL: SELECT COUNT(DISTINCT Country) FROM table; Pandas: df['Country'].nunique() PySpark: df.select(F.countDistinct('Country')) |