Uebung3

Benötigte Pakete:

```
Oracle-Pandas: 'pip install pandas-oracle', https://github.com/cwade/pandas_oraclecx_Oracle: siehe untenSeaborn, Pandas, Numpy, etc
```

Anleitung:

- Installiert cx Oracle
- Ladet den Oracle Instant Client herunter: https://www.oracle.com/database/technologies/instant-client/downloads.html
- Führt die entsprechenden Installationsanweisungen durch

```
In [1]: # Bei Bedarf auskommentieren
!python -m pip install cx_Oracle --upgrade
!python -m pip install pandas-oracle
```

Requirement already satisfied: cx_Oracle in /opt/anaconda3/envs/data/lib/python3.1 1/site-packages (8.3.0)

Requirement already satisfied: pandas-oracle in /opt/anaconda3/envs/data/lib/python 3.11/site-packages (2.1.4)

Requirement already satisfied: pandas in /opt/anaconda3/envs/data/lib/python3.11/si te-packages (from pandas-oracle) (2.2.3)

Requirement already satisfied: cx-Oracle in /opt/anaconda3/envs/data/lib/python3.1 1/site-packages (from pandas-oracle) (8.3.0)

Requirement already satisfied: pyaml in /opt/anaconda3/envs/data/lib/python3.11/sit e-packages (from pandas-oracle) (24.12.1)

Requirement already satisfied: numpy>=1.23.2 in /opt/anaconda3/envs/data/lib/python 3.11/site-packages (from pandas->pandas-oracle) (1.26.4)

Requirement already satisfied: python-dateutil>=2.8.2 in /opt/anaconda3/envs/data/lib/python3.11/site-packages (from pandas->pandas-oracle) (2.9.0)

Requirement already satisfied: pytz>=2020.1 in /opt/anaconda3/envs/data/lib/python 3.11/site-packages (from pandas->pandas-oracle) (2024.1)

Requirement already satisfied: tzdata>=2022.7 in /opt/anaconda3/envs/data/lib/pytho n3.11/site-packages (from pandas->pandas-oracle) (2023.3)

Requirement already satisfied: PyYAML in /opt/anaconda3/envs/data/lib/python3.11/site-packages (from pyaml->pandas-oracle) (6.0.2)

Requirement already satisfied: six>=1.5 in /opt/anaconda3/envs/data/lib/python3.11/site-packages (from python-dateutil>=2.8.2->pandas->pandas-oracle) (1.16.0)

Pfad setzen

HU-VPN aktivieren oder in HU-Netzwerk (WLAN) einloggen.

Anleitungen: https://www.cms.hu-berlin.de/de/dl/netze/vpn

Verbindungsdaten:

Die Verbindungsdaten werden in config.yml gespeichert. Die Datein *muss* im gleichen Ordner liegen.

```
import cx_Oracle
import pandas_oracle.tools as pt
import numpy as np
import pandas as pd

# visualization
import seaborn as sns
import matplotlib.pyplot as plt
import folium
from folium import plugins
```

Verbindungsaufbau:

```
In [5]: def run_query(query) :
    ## opening conn
    conn = pt.open_connection("config.yml")

try:
    ## passing the conn object to the query_to_df
    df1 = pt.query_to_df(query, conn, 10000)
    return df1
except Exception as e:
    print("An exception occurred")
    print(str(e))
finally:
    ## close connection
    pt.close_connection(conn)
```

Wichtig: die Verbindungs zur DB muss geschlossen werden, wenn ein Fehler auftritt.

EXPLAIN PLAN

```
In [6]: def explain_plan_query(query) :
    ## opening conn
    conn = pt.open_connection("config.yml")
    try:
        # There is small change
        # ("+query+")" => " + query
        # query_explain = "EXPLAIN PLAN FOR ("+query+")"
        query_explain = "EXPLAIN PLAN FOR " + query
        pt.execute(query_explain, conn)
        query2 = "SELECT PLAN_TABLE_OUTPUT FROM TABLE(DBMS_XPLAN.DISPLAY())"

        df = pt.query_to_df(query2, conn, 10000)
        return df
        except Exception as e:
            print("An exception occurred:", str(e))
```

```
finally:
    ## close connection
    pt.close_connection(conn)
```

```
In [7]: query = "SELECT * FROM delays"
df = explain_plan_query(query)
print(df.to_string())
```

```
PLAN_TABLE_OUTPUT
0
                                                   Plan hash value: 3893133417
1
2
  | Id | Operation
3
                            | Name | Rows | Bytes | Cost (%CPU)| Time
4
5
       0 | SELECT STATEMENT |
                                        5332K|
                                                 472M| 22662
                                                               (2) | 00:00:01 |
6
      1 | TABLE ACCESS FULL| DELAYS | 5332K|
                                                 472M| 22662
                                                             (2) | 00:00:01 |
```

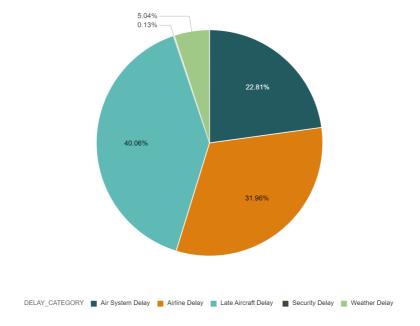
1. Ursache der Verspätung

SQL

```
query1_1 = """
In [8]:
        SELECT
            CASE
                WHEN delay_type = 'air_system_delay' THEN 'Air System Delay'
                WHEN delay_type = 'security_delay' THEN 'Security Delay'
                WHEN delay_type = 'airline_delay' THEN 'Airline Delay'
                WHEN delay_type = 'late_aircraft_delay' THEN 'Late Aircraft Delay'
                WHEN delay_type = 'weather_delay' THEN 'Weather Delay'
            END as delay_category,
            SUM(delay_minutes) as total_minutes,
            ROUND(SUM(delay_minutes) * 100.0 / SUM(SUM(delay_minutes)) OVER (), 2) as perce
        FROM (
            SELECT 'air_system_delay' as delay_type, COALESCE(air_system_delay, 0) as delay
            SELECT 'security_delay', COALESCE(security_delay, 0) FROM delays
            SELECT 'airline_delay', COALESCE(airline_delay, 0) FROM delays
            UNION ALL
            SELECT 'late_aircraft_delay', COALESCE(late_aircraft_delay, 0) FROM delays
            UNION ALL
            SELECT 'weather_delay', COALESCE(weather_delay, 0) FROM delays
        ) unpivoted
        GROUP BY delay_type
        HAVING SUM(delay_minutes) > 0
        ORDER BY total minutes DESC
        df1_1 = run_query(query1_1)
        df1 1
```

Out[8]: DELAY_CATEGORY TOTAL_MINUTES PERCENTAGE

0	Late Aircraft Delay	23767674	40.06
1	Airline Delay	18966945	31.96
2	Air System Delay	13533065	22.81
3	Weather Delay	2991008	5.04
4	Security Delay	77945	0.13



PLAN_TABLE_OUTP

```
UT
                                                         Plan hash value: 38516895
0
23
1
2
    | Id | Operation
                                  | Name | Rows | Bytes | Cost (%CPU) | Time
3
4
5
    | 0 | SELECT STATEMENT
                                  26M|
                                                       610M|
                                                              115K (4) | 00:00:05
6
       1 | WINDOW SORT
                                  I
                                                26M|
                                                       610M|
                                                              115K (4) | 00:00:05
7
    |* 2 |
             FILTER
                                           3 |
              HASH GROUP BY
                                                26M|
                                                       610M|
                                                              115K (4) | 00:00:05
8
                                  I
       4 |
               VIEW
                                                       610M|
                                                              113K (2) | 00:00:05
9
                                           26M|
                UNION-ALL
       5 |
10
                                           11
       6 |
                 TABLE ACCESS FULL | DELAYS | 5332K
                                                       10M| 22630
                                                                    (2) | 00:00:01
                 TABLE ACCESS FULL | DELAYS | 5332K|
12
       7 |
                                                       10M| 22630
                                                                    (2) | 00:00:01
13
       8 |
                 TABLE ACCESS FULL | DELAYS | 5332K
                                                       10M| 22630
                                                                    (2) | 00:00:01
14
       9 |
                 TABLE ACCESS FULL | DELAYS | 5332K
                                                       10M| 22630
                                                                    (2) | 00:00:01
                 TABLE ACCESS FULL | DELAYS | 5332K
15
      10 |
                                                       10M| 22630 (2)| 00:00:01
16
17
18
                                 Predicate Information (identified by operation i
d):
19
20
21
                                                  2 - filter(SUM("DELAY_MINUTES")>
0)
```

3.2 Analyse: Region

3.2.1 SQL

```
In [12]: query2 1 ="""
         SELECT
             f.origin_state as STATE,
             f.origin_airport as AIRPORT,
             f.origin_city as City,
             f.origin_latitude/100000.0 as LATITUDE,
             f.origin_longitude/100000.0 as LONGITUDE,
             COUNT(*) as TOTAL_FLIGHTS,
             SUM(CASE WHEN d.departure_delay > 0 THEN 1 ELSE 0 END) as DELAYED_FLIGHTS,
             ROUND(SUM(CASE WHEN d.departure_delay > 0 THEN 1 ELSE 0 END) * 100.0 / COUNT(*)
             ROUND(AVG(CASE WHEN d.departure_delay > 0 THEN d.departure_delay ELSE NULL END)
         FROM Flight f
         JOIN Delays d ON f.flight_id = d.flight_id
         GROUP BY
             f.origin_state,
             f.origin_airport,
             f.origin_city,
```

```
f.origin_latitude,
  f.origin_longitude
HAVING COUNT(*) > 100
"""

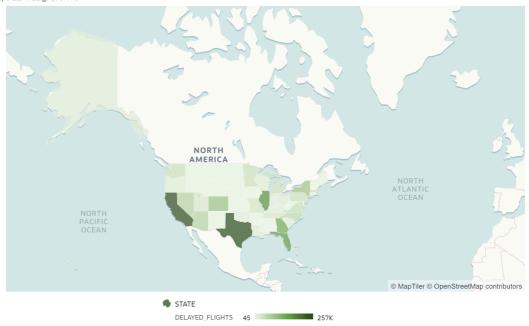
df2_1 = run_query(query2_1)
df2_1
```

Out[12]:		STATE	AIRPORT	CITY	LATITUDE	LONGITUDE	TOTAL_FLIGHTS	DELAYED
	0	TN	McGhee Tyson Airport	Knoxville	35.81249	-83.99286	6963	
	1	GA	Savannah/Hilton Head International Airport	Savannah	32.12758	-81.20214	7542	
	2	NC	Albert J. Ellis Airport	Jacksonville	34.82916	-77.61214	1136	
	3	TX	Easterwood Airport	College Station	30.58859	-96.36382	2311	
	4	TX	William P. Hobby Airport	Houston	29.64542	-95.27889	52042	
	•••							
	309	AK	Ketchikan International Airport	Ketchikan	55.35557	-131.71374	2319	
	310	МІ	Bishop International Airport	Flint	42.96550	-83.74346	4746	
	311	UT	Canyonlands Field	Moab	38.75496	-109.75484	206	
	312	MI	Kalamazoo/Battle Creek International Airport	Kalamazoo	42.23488	-85.55206	1845	
	313	AK	Petersburg James A. Johnson Airport	Petersburg	56.80165	-132.94528	664	

Visualisierung

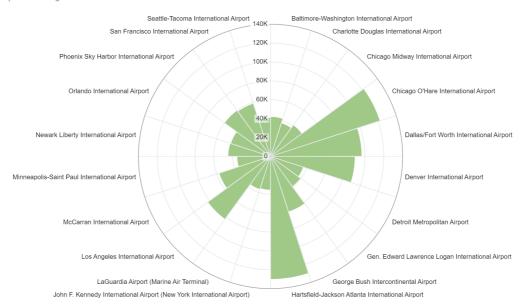
314 rows × 9 columns

STATE, DELAYED_FLIGHTS



DELAYED_FLIGHTS by AIRPORT

Top 20 DELAYED_FLIGHTS



```
df2_3 = explain_plan_query(query2_1)
print(df2_3.to_string())
                                                     PLAN TABL
E OUTPUT
                                              Plan hash value: 5
0
78984892
1
2
 3
4
5 | 0 | SELECT STATEMENT |
                             | 3507 | 267K| | 33622 (3)| 0
0:00:02 |
 |* 1 | FILTER
                      | 2 | HASH GROUP BY
                                      267K|
                      | 3507 |
                                              | 33622 (3)| 0
0:00:02 |
                             | 5264K|
8 |* 3 |
                      HASH JOIN
                                      391M| 38M| 33225 (2)| 0
0:00:02 |
9 | 4 |
          TABLE ACCESS FULL| FLIGHT | 493K|
                                              | 3322 (1)| 0
                                       32M|
0:00:01 |
10 | 5 |
          TABLE ACCESS FULL | DELAYS | 5332K | 45M |
                                              | 22630 (2)| 0
0:00:01 |
12
13
                            Predicate Information (identified by operat
ion id):
```

1 - filter(COUNT

3 - access("F"."FLIGHT_ID"="D"."FLI

3. Analyse: Weather

3.1 10 Tage meisten Stornierungen wegen Wetter

SQL

14

15

16 (*)>100) 17

GHT_ID")

0.00

```
In [14]: # Query 1: Top 10 days with most weather-related cancellations
         query3_1 = """
         SELECT
              f.year,
              f.month,
             f.day,
              COUNT(*) as WEATHER_CANCELLATIONS,
              ROUND(COUNT(*) * 100.0 / (SELECT COUNT(*)
                                         FROM Delays
                                         WHERE cancelled = 1
                                         AND cancellation_reason = 'B'), 2) as PERCENTAGE_OF_T
         FROM Delays f
         WHERE f.cancelled = 1
         AND f.cancellation_reason = 'B'
         GROUP BY f.year, f.month, f.day
         \mathbf{n} \mathbf{n}
```

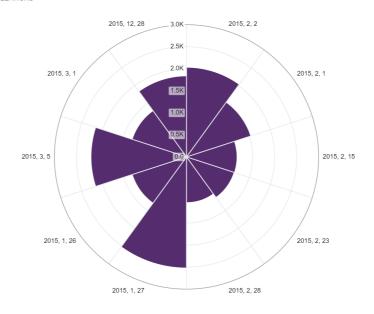
df3_1 = run_query(query3_1)
df3_1

Out[14]:		YEAR	MONTH	DAY	WEATHER_CANCELLATIONS	PERCENTAGE_OF_TOTAL
	0	2015	1	9	136	0.28
	1	2015	1	14	81	0.17
	2	2015	1	15	27	0.06
	3	2015	1	16	19	0.04
	4	2015	2	6	135	0.28
	•••					
	326	2015	11	30	30	0.06
	327	2015	12	8	34	0.07
	328	2015	12	11	14	0.03
	329	2015	12	10	22	0.05
	330	2015	12	18	72	0.15

331 rows × 5 columns

Visualisierung

WEATHER_CANCELLATIONS by YEAR, MONTH, DAY
Top 10 WEATHER_CANCELLATIONS



```
In [15]: query3_2 = """
y-- Index for filtering conditions
CREATE INDEX idx_delays_cancel ON Delays(cancelled, cancellation_reason);

-- Composite index for grouping columns
CREATE INDEX idx_delays_date ON Delays(year, month, day);

-- Combined index covering all conditions and grouping columns
CREATE INDEX idx_delays_cancel_date ON Delays(cancelled, cancellation_reason, year,
"""
```

```
df3_3 = explain_plan_query(query3_1)
print(df3_3.to_string())
```

```
PLAN TABLE OUTPUT
0
                                                       Plan hash value: 1046980958
1
2
3
    | Id | Operation
                               | Name
                                        | Rows | Bytes | Cost (%CPU)| Time
4
                                                                   (2) | 00:00:02 |
5
        0 | SELECT STATEMENT
                                            331 | 4965 | 45240
                                                      5 |
6
        1 | SORT AGGREGATE
                                              1 |
7
       2 |
            TABLE ACCESS FULL | DELAYS | 47874 |
                                                                   (2) | 00:00:01 |
    |*
                                                    233K| 22618
8
        3 | HASH GROUP BY
                                            331 | 4965 | 45240
                                                                   (2) | 00:00:02 |
9
    |* 4 | TABLE ACCESS FULL | DELAYS | 47874 | 701K | 22618
                                                                   (2) | 00:00:01 |
10
11
12
                              Predicate Information (identified by operation id):
13
14
                          2 - filter("CANCELLED"=1 AND "CANCELLATION_REASON"='B')
15
                  4 - filter("F"."CANCELLED"=1 AND "F"."CANCELLATION_REASON"='B')
16
17
18
                                                                              Note
19
20
                            - dynamic statistics used: dynamic sampling (level=2)
21
                                   - 1 Sql Plan Directive used for this statement
```

3.2 An welchen Flughafen

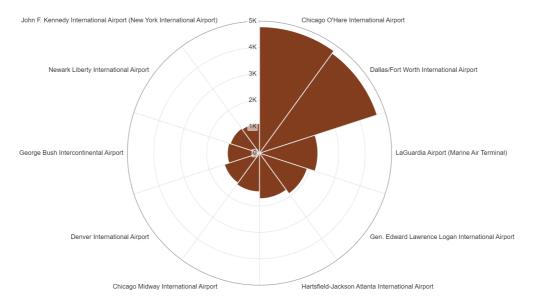
```
In [16]: query3_4 = """
         SELECT
             f.origin_state as STATE,
             f.origin_airport as AIRPORT,
             COUNT(*) as TOTAL CANCELLATIONS,
             SUM(CASE WHEN d.cancellation_reason = 'B' THEN 1 ELSE 0 END) as WEATHER_CANCELL
             ROUND (
                 (SUM(CASE WHEN d.cancellation_reason = 'B' THEN 1 ELSE 0 END) * 100.0) /
                 NULLIF(COUNT(*), 0),
             ) as CANCELLATION_PERCENTAGE,
             COUNT(DISTINCT TRUNC(f.scheduled_departure)) as AFFECTED_DAYS
         FROM Flight f
         JOIN Delays d ON f.flight_id = d.flight_id
         WHERE d.cancelled = 1
         GROUP BY
             f.origin_state,
             f.origin_airport
         HAVING SUM(CASE WHEN d.cancellation_reason = 'B' THEN 1 ELSE 0 END) > 0
         ORDER BY WEATHER_CANCELLATIONS DESC
         df3_4 = run_query(query3_4)
         df3_4
```

	STATE	AIRPORT	TOTAL_CANCELLATIONS	WEATHER_CANCELLATIONS	CANCELLATIO
0	IL	Chicago O'Hare International Airport	8547	4769	
1	TX	Dallas/Fort Worth International Airport	6254	4664	
2	NY	LaGuardia Airport (Marine Air Terminal)	4531	2191	
3	MA	Gen. Edward Lawrence Logan International Airport	2654	1882	
4	GA	Hartsfield- Jackson Atlanta International Airport	2557	1707	
•••	•••				
306	UT	Canyonlands Field	1	1	
307	NY	Plattsburgh International Airport	2	1	
308	HI	Lihue Airport	44	1	
309	NY	Ithaca Tompkins Regional Airport	4	1	
310	ME	Bangor International Airport	3	1	
311 ro	ws × 6 co	olumns			

Visualisierung

Out[16]:

Top 10 WEATHER_CANCELLATIONS by AIRPORT



```
In [17]: query3_5 = """
         -- 1. Index for JOIN operations
         CREATE INDEX idx_delays_flight_id ON Delays(flight_id);
         CREATE INDEX idx_flight_id ON Flight(flight_id);
         -- 2. Composite index for filtering and analysis on Delays table
         CREATE INDEX idx_delays_cancel ON Delays(
             cancelled,
             cancellation_reason,
             flight_id
         );
         -- 3. Composite index for grouping columns on Flight table
         CREATE INDEX idx_flight_origin ON Flight(
             origin_state,
             origin_airport,
             scheduled_departure
         );
         nim
         df3_5 = explain_plan_query(query3_4)
         print(df3_5.to_string())
```

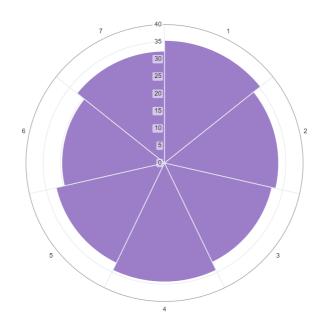
```
TABLE OUTPUT
                                                        Plan hash valu
e: 2411417394
1
3 | Id | Operation
                            | Name | Rows | Bytes |TempSpc| Cost (%CP
U)| Time |
5 | 0 | SELECT STATEMENT |
                                     | 615 | 57195 | | 26620
(2) | 00:00:02 |
6 | 1 | SORT ORDER BY
                            | 615 | 57195 |
                                                         | 26620
(2) | 00:00:02 |
7 |* 2 | FILTER
                                     | 615 | 57195 |
                            | 26620
(2) | 00:00:02 |
            VIEW
                            | VW_DAG_0 | 23655 | 2148K|
9 | 4 |
                                                         | 26616
(2) | 00:00:02 |
             HASH GROUP BY
                                     | 23655 | 1409K| 6552K| 26616
10 | 5 |
(2) | 00:00:02 |
                                      | 87429 | 5208K|
11 |* 6 |
             HASH JOIN |
(2) | 00:00:02 |
              TABLE ACCESS FULL| DELAYS | 87430 | 853K| | 22602
12 |* 7 |
(2) | 00:00:01 |
13 | 8 |
              TABLE ACCESS FULL | FLIGHT | 493K | 24M | 3322
(1) | 00:00:01 |
14 -----
15
                                     Predicate Information (identified by o
16
peration id):
17
18
                                                        2 - filter(SUM
19
("ITEM_4")>0)
                                           6 - access("F"."FLIGHT_I
20
D"="D"."FLIGHT ID")
                                                       7 - filter
("D"."CANCELLED"=1)
22
23
Note
24
                                   - dynamic statistics used: dynamic sampl
25
ing (level=2)
                                         - 1 Sql Plan Directive used for t
26
his statement
4. Beste Zeit
4.1 Wochentage
```

Out[18]:

	DAY_OF_WEEK	AVERAGE_DELAY_MINUTES	TOTAL_FLIGHTS
0	6	29.53	629444
1	3	31.69	780519
2	5	31.97	772027
3	7	32.21	741681
4	2	32.82	765893
5	4	34.24	777598
6	1	35.31	778322

Visualisierung

AVERAGE_DELAY_MINUTES by DAY_OF_WEEK



```
day,
    cancelled
);
-- 2. Composite index for the JOIN conditions on Time table
CREATE INDEX idx_time_date ON Time(
    year,
    month,
    day,
    day_of_week
);
-- 3. Index for delay columns (optional, for covering index)
CREATE INDEX idx_delays_all_delays ON Delays(
    departure_delay,
    taxi_out,
    security_delay,
    airline_delay,
    weather_delay,
   air_system_delay
);
df4_3 = explain_plan_query(query4_1)
print(df4_3.to_string())
                                                                  PLAN TABLE OUTPUT
```

		PLAN_TABLE_OUTPUT
0		Plan hash value: 2274342594
1		
2		
3	Id Operation Nam	ne Rows Bytes Cost (%CPU) Time
5	0 SELECT STATEMENT	7 287 23483 (5) 00:00:01
6	1 SORT ORDER BY	7 287 23483 (5) 00:00:01
7	2 HASH GROUP BY	7 287 23483 (5) 00:00:01
8	* 3 HASH JOIN	5245K 205M 22691 (2) 00:00:01
9	4 TABLE ACCESS FULL TIM	1E 365 4745 3 (0) 00:00:01
10	·	AYS 5245K 140M 22649 (2) 00:00:01
11		
12		
13	Predic	cate Information (identified by operation id):
14		
15		
16	3 - access("F","DA	AY"="T"."DAY" AND "F"."MONTH"="T"."MONTH" AND
17	J decess(I I DA	"F"."YEAR"="T"."YEAR")
18		5 - filter("F"."CANCELLED"=0)
TO		J - IIILLEI (F . CANCELLED -0)

4.2 Sationale Unterschiede

```
JOIN TIME t ON
    f.DAY = t.DAY AND
    f.MONTH = t.MONTH AND
    f.YEAR = t.YEAR
WHERE f.CANCELLED = 0
GROUP BY t.MONTH
ORDER BY t.MONTH ASC
df4_4 = run_query(query4_4)
df4_4
```

494777

515912

Out[33]:		MONTH	AVERAGE_DELAY_MINUTES	TOTAL_FLIGHTS
	0	1	33.30	457986
	1	2	37.24	408674
	2	3	32.17	493310
	3	4	29.29	480631
	4	5	31.78	491299

7 32.67 505484 8 9 25.19 462871 9 11 28.34 463373 10 12 35.42 471167

38.95

34.57

Visualisierung

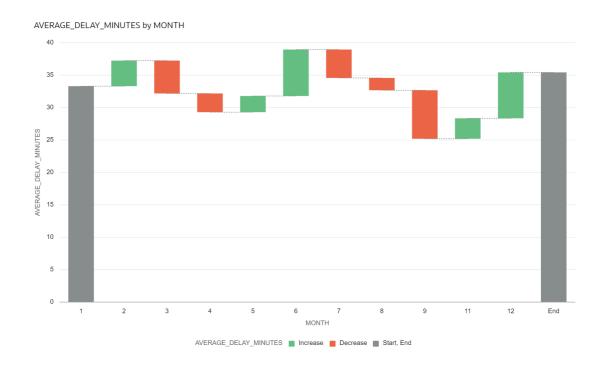
6

7

8

5

6



```
In [21]: query4_5 = """
         -- 1. Composite index for the JOIN conditions on Delays table
         CREATE INDEX idx_delays_date_cancel ON Delays(
             year,
```

```
month,
    day,
    cancelled
);
\operatorname{--} 2. Composite index for the JOIN conditions on Time table
CREATE INDEX idx_time_date_month ON Time(
    year,
    month,
    day
);
-- 3. Index for delay columns
CREATE INDEX idx_delays_all_delays ON Delays(
   departure_delay,
   taxi_out,
   security_delay,
   airline_delay,
   weather_delay,
   air_system_delay
);
df4_5 = explain_plan_query(query4_4)
print(df4_5.to_string())
```

PLAN_TABLE_OUT

```
PUT
                                                    Plan hash value: 1954953
516
1
2
  | Id | Operation
                             | Name | Rows | Bytes | Cost (%CPU)| Time
4
5
  | 0 | SELECT STATEMENT
                             11 | 649 | 23046 (4) | 00:00:0
1 |
     1 | SORT GROUP BY
                                           11 | 649 | 23046
6
                             (4) | 00:00:0
1 |
7
   | 2 | NESTED LOOPS
                                      | 334 | 19706 | 23045
                                                             (4) | 00:00:0
                             1 |
   | 3 | VIEW
                             | VW_GBC_6 | 334 | 16366 | 23045
                                                              (4) | 00:00:0
8
1 |
  | 4 | HASH GROUP BY
                             | 334 | 9352 | 23045
                                                              (4) | 00:00:0
9
1 |
10 |* 5 | TABLE ACCESS FULL| DELAYS | 5245K| 140M| 22649
                                                              (2) | 00:00:0
1 |
11 |* 6 | INDEX UNIQUE SCAN | PK_TIME | 1 | 10 | 0 (0) | 00:00:0
1 |
12
13
14
                              Predicate Information (identified by operation i
d):
15
___
16
                                                  5 - filter("F"."CANCELLE
17
D"=0)
                       6 - access("ITEM_3"="T"."DAY" AND "ITEM_2"="T"."MONTH" A
18
ND
                                                          "ITEM 1"="T"."YEA
19
R")
20
                                                                        Ν
21
ote
22
23
                             - dynamic statistics used: dynamic sampling (level
=2)
```

5. Analyse: Fluglinien

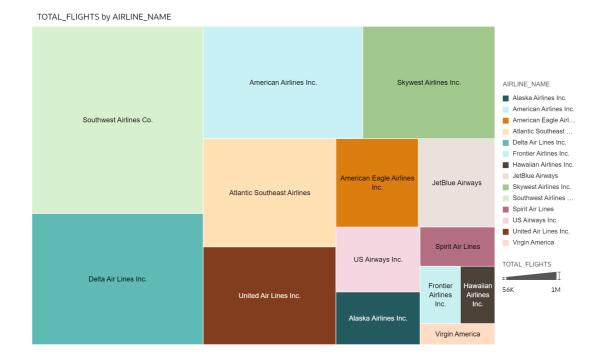
5.1 Fluglinien mit meisten Flügen

```
df5_1 = run_query(query5_1)
df5_1
```

\sim			г	-	-	п	
11		т.		- /	~	-	
U	u	L	ı.	\angle	J	J.	

	AIRLINE_NAME	TOTAL_FLIGHTS
0	Southwest Airlines Co.	1141618
1	Delta Air Lines Inc.	796522
2	American Airlines Inc.	638500
3	Skywest Airlines Inc.	529826
4	Atlantic Southeast Airlines	511495
5	United Air Lines Inc.	463384
6	American Eagle Airlines Inc.	257898
7	JetBlue Airways	240989
8	US Airways Inc.	194648
9	Alaska Airlines Inc.	157418
10	Spirit Air Lines	105228
11	Frontier Airlines Inc.	82159
12	Hawaiian Airlines Inc.	69871
13	Virgin America	55928

Visualisierung



```
-- 2. Index for airline lookup
CREATE INDEX idx_airline_iata ON Airline(
    iata_code,
    airline
);
"""

df5_3 = explain_plan_query(query5_1)
print(df5_3.to_string())
```

```
PLAN_TABLE_OUTPU
Т
0
                                                    Plan hash value: 79856489
2
1
2
3
  | Id | Operation
                            | Name | Rows | Bytes | Cost (%CPU)| Time
4
5
   | 0 | SELECT STATEMENT
                                     | 14 | 434 | 23483 (5)| 00:00:01
                            | 1 | SORT ORDER BY
                            | 14 | 434 | 23483
                                                            (5) | 00:00:01
6
   | 2 | HASH GROUP BY
                                     | 14 | 434 | 23483
7
                             (5) | 00:00:01
                                     | 5245K| 155M| 22691
8
   |* 3 | HASH JOIN
                            (2) | 00:00:01
9
   | 4 |
            TABLE ACCESS FULL | AIRLINE | 14 | 350 | 3
                                                            (0) | 00:00:01
10
   |* 5 |
            TABLE ACCESS FULL | DELAYS | 5245K | 30M | 22649 (2) | 00:00:01
11
12
                             Predicate Information (identified by operation i
13
d):
14
15
                                 3 - access("F"."AIRLINE IATA"="A"."IATA COD
16
E")
                                                 5 - filter("F"."CANCELLED"=
17
0)
```

5.2 Fluglinien mit höchste Verspätungen

Ο.			\neg	Е	1
UI	IJι	. 1	7	Э.	

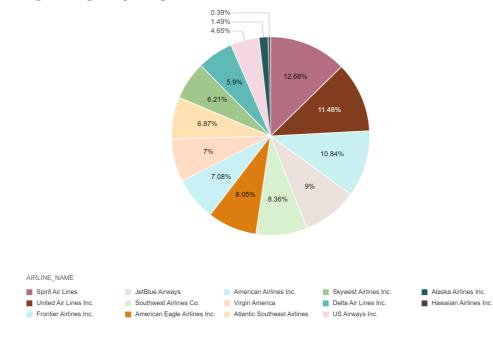
	AIRLINE_NAME	AVERAGE_DEPARTURE_DELAY
0	Spirit Air Lines	16.65
1	United Air Lines Inc.	15.07
2	Frontier Airlines Inc.	14.23
3	JetBlue Airways	11.81
4	Southwest Airlines Co.	10.98
5	American Eagle Airlines Inc.	10.57
6	American Airlines Inc.	9.29
7	Virgin America	9.19
8	Atlantic Southeast Airlines	9.02
9	Skywest Airlines Inc.	8.15
10	Delta Air Lines Inc.	7.75
11	US Airways Inc.	6.11
12	Alaska Airlines Inc.	1.95

Visualisierung

13



Hawaiian Airlines Inc.



0.51

```
iata_code,
airline
);
"""

df5_5 = explain_plan_query(query5_4)
print(df5_5.to_string())

PLAN TABLE OUTPU
```

```
Т
0
                                                     Plan hash value: 79856489
2
1
2
3
   | Id | Operation
                             | Name | Rows | Bytes | Cost (%CPU)| Time
5
   | 0 | SELECT STATEMENT
                                      | 14 | 490 | 23476
                                                              (5) | 00:00:01
   | 1 | SORT ORDER BY
                              | 14 |
                                                  490 | 23476
                                                              (5) | 00:00:01
6
       2 | HASH GROUP BY
                                      | 14 |
                                                  490 | 23476
                                                              (5) | 00:00:01
7
                             |* 3 |
                                      | 5163K|
                                                  172M| 22697
8
           HASH JOIN
                             (2) | 00:00:01
             TABLE ACCESS FULL | AIRLINE | 14 |
9
   | 4 |
                                                  350 | 3
                                                               (0) | 00:00:01
   |* 5 |
10
             TABLE ACCESS FULL | DELAYS | 5163K | 49M | 22656
                                                              (2) | 00:00:01
11
12
                              Predicate Information (identified by operation i
13
d):
14
15
                                  3 - access("F"."AIRLINE_IATA"="A"."IATA_COD
16
E")
17
               5 - filter("F"."DEPARTURE_DELAY" IS NOT NULL AND "F"."CANCELLED"=
0)
```

6.Beliebte Flüge

6.1 Beliebteste Flüge

```
df6_1 = run_query(query6_1)
df6_1
```

Out[27]:		CITY_NAME	STATE	INCOMING_FLIGHTS
	0	Chicago	IL	355664
	1	Atlanta	GA	344189
	2	Dallas-Fort Worth	TX	232833
	3	Houston	TX	195516
	4	Denver	СО	193700
	•••			
	313	St Cloud	MN	77
	314	Dillingham	AK	77
	315	Gustavus	AK	76
	316	King Salmon	AK	63
	317	Ithaca	NY	31

318 rows × 3 columns

Visulaisierung

INCOMING_FLIGHTS by CITY_NAME, INCOMING_FLIGHTS

Top 30 INCOMING_FLIGHTS Detroit Orlando Boston Phoenix Los Angeles Dallas-Fort Worth Chicago Salt Lake City Las Vegas Houston New York San Diego Philadelphia Minneapolis Ft. Lauderdale Atlanta Tampa Portland Nashville San Francisco Arlington Dallas St Louis Honolulu INCOMING_FLIGHTS 43K 356K

INCOMING_FLIGHTS 43K I 356K

```
);
0.00
df6_3 = explain_plan_query(query6_1)
print(df6_3.to_string())
                                                              PLAN_TABL
E_OUTPUT
a
                                                       Plan hash value: 6
73129677
1
2
3 | Id | Operation | Name | Rows | Bytes | TempSpc | Cost (%CPU) | Ti
me
    5 | 0 | SELECT STATEMENT | 318 | 8268 | | 32498 (4) | 0
0:00:02 |
6 | 1 | SORT ORDER BY
                                  | 318 | 8268 |
                          - 1
                                                      | 32498 (4)| 0
0:00:02 |
7 | 2 |
          HASH GROUP BY
                          | 318 | 8268 |
                                                      | 32498 (4)| 0
0:00:02 |
                          | 5177K| 128M| 14M| 31717 (2)| 0
8 |* 3 | HASH JOIN
0:00:02 |
            TABLE ACCESS FULL| FLIGHT | 493K| 8678K| | 3322 (1)| 0
9 | 4 |
0:00:01 |
10 |* 5 | TABLE ACCESS FULL| DELAYS | 5245K| 40M|
                                                      | 22649 (2)| 0
0:00:01 |
11 -----
12
                                 Predicate Information (identified by operat
13
ion id):
14
15
                                        3 - access("F"."FLIGHT_ID"="D"."FLI
16
GHT ID")
                                                    5 - filter("D"."CANCE
17
LLED"=0)
```

CREATE INDEX idx_flight_destination ON Flight(

destination_city,
destination_state

6.2 Beliebsteste Abflughafen

```
df6_4 = run_query(query6_4)
df6_4
```

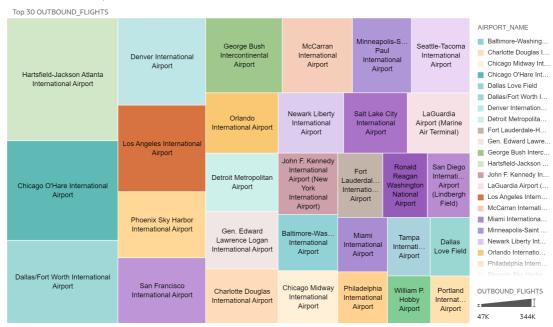
	AIRPORT_NAME	CITY	STATE	OUTBOUND_FLIGHTS
0	Hartsfield-Jackson Atlanta International Airport	Atlanta	GA	344279
1	Chicago O'Hare International Airport	Chicago	IL	277336
2	Dallas/Fort Worth International Airport	Dallas-Fort Worth	TX	233297
3	Denver International Airport	Denver	СО	193932
4	Los Angeles International Airport	Los Angeles	CA	192509
•••				
317	St. Cloud Regional Airport	St Cloud	MN	78
318	Dillingham Airport	Dillingham	AK	77
319	Gustavus Airport	Gustavus	AK	76
320	King Salmon Airport	King Salmon	AK	63
321	Ithaca Tompkins Regional Airport	Ithaca	NY	30

322 rows × 4 columns

Visualisierung

Out[30]:

OUTBOUND_FLIGHTS by AIRPORT_NAME



```
origin_airport,
   origin_city,
   origin_state
 );
0.00
df6_5 = explain_plan_query(query6_4)
print(df6_5.to_string())
                                                              PLAN_TABL
E_OUTPUT
                                                       Plan hash value: 6
0
73129677
3
  | Id | Operation
                     | Name | Rows | Bytes |TempSpc| Cost (%CPU)| Ti
me
    5 | 0 | SELECT STATEMENT
                          | | 12296 | 768K| | 33386 (4)|0
0:00:02 |
6 | 1 | SORT ORDER BY
                          | 12296 |
                                            768K|
                                                      | 33386 (4)| 0
0:00:02 |
                                  | 12296 |
                                            768K|
7 | 2 | HASH GROUP BY
                          | 33386 (4)| 0
0:00:02 |
          HASH JOIN
8 |* 3 |
                     | | 5177K|
                                            316M| 32M| 32605 (2)| 0
0:00:02 |
            TABLE ACCESS FULL| FLIGHT | 493K|
9 | 4 |
                                             26M|
                                                     | 3322 (1)| 0
0:00:01 |
            TABLE ACCESS FULL| DELAYS | 5245K| 40M|
                                                      | 22649 (2)| 0
10 |* 5 |
0:00:01 |
11 -----
12
13
                                 Predicate Information (identified by operat
ion id):
14
15
                                        3 - access("D"."FLIGHT ID"="F"."FLI
16
GHT ID")
                                                    5 - filter("D"."CANCE
17
LLED"=0)
```

-- 2. Composite index for origin grouping
CREATE INDEX idx_flight_origin ON Flight(