Unlocking Tennis Insights from SportRadar!

This project isn't just about tennis; it's about conquering data. We've harnessed the power of the Sport Radar API to deliver a comprehensive tennis data analytics solution.

From raw API responses to interactive dashboards, we've covered the entire court in 3 steps.

1. Data Acquisition & Transformation (Python's Precision):

- API Web Scraping (Aggressive Baseline): We didn't just get the data; we *scraped* it!
 The "Tennis_data_web_scrapped.ipynb" notebook showcases how we pulled data directly from the Sport Radar API using Python's "requests" library.
- **JSON to Python Mastery (Parsing Power):** The API serves data in JSON format, and we're not afraid of nested structures. We used Python's **"json"** library to parse these responses, transforming them into Python objects for efficient manipulation.
- Pandas Powerhouse (Data Wrangling): Pandas is our secret weapon! We used it
 extensively to structure the data into Data Frames, enabling tasks like Creating tables
 for competitors, complexes, and doubles data.
- Merging DataFrames: With Pandas library we combined competitor rankings with competitor details similar to joins in SQL using attribute "merge()", creating a unified view of the data. Also, handled missing values ("fillna()") to ensure data integrity.
- Data Persistence (CSV Serve): We saved the processed DataFrames into CSV files ("competitor.csv", "complexes.csv", "doubles_data.csv") for easy access and use in our Streamlit app.

2. SQL (Query Control):

- Targeted Queries: Check out "SQL_queries.sql" We've crafted a set of SQL queries designed to extract specific insights from the tennis data. These queries demonstrate:
 - * Basic data retrieval ("select").
 - * Filtering data based on conditions ("where").
 - *Grouping data and using aggregate functions (group by, count(), sum() etc...).
 - *Ordering results (order by)
 - * Using window functions ("DENSE_RANK()") for advanced ranking.

* These queries are aligned with the analysis goals specified in the "Game Analytics" document.

3. Streamlit Showcase:

- **Dynamic Dashboard:** The "**Web_app_tennis_data.py**" script brings the data to life with a Streamlit web application.
- Data Loading (Ready to Play): The app reads the CSV files generated in the data processing step using "pd.read_csv()".
- Interactive Filtering: Streamlit's widgets ("st.selectbox()", "st.slider()") allow users to dynamically filter the data:
 - * Select different datasets (competitor, complexes, doubles).
 - * Filter competitors by category, type, and gender.
 - * Filter complexes by venue, country, timezone, and complex name.
 - * Filter doubles data by rank, competitor name, country, competitions played, and points.
- Data Display: The filtered data is presented in an interactive DataFrame using "st.dataframe()", along with row and column counts for quick analysis.

Your Analytical Advantage: 🚀

This project empowers you to

- Explore Competition Hierarchies: Dive deep into tournament structures.
- Analyze Event Trends: Visualize data distributions and patterns.
- Gain Player Insights: Analyze competitor data and performance metrics.
- Leverage Interactive Tools: Use the Streamlit app to filter and explore the data on your own terms.