#### **BI MODULE ASSIGNMENT 1 ANSWERS**

#### **Question 1**

a. Query 1: Profit Margins by Event Promotion Type and Marketeer.

## Figure 1

## Query 1

```
SELECT CONCATURER(Cd.Fornthamp, '-', ef.Fornthamp') AS (Event Name),

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```

The query above highlights the profitability of promotions from multiple dimensions to help key decision makers optimize marketing spend, identify successful campaigns, and focus on high-performing promotions, promotion types, marketers, and combinations of both when planning upcoming events.

Promotion profit is calculated by subtracting PromotionCost from PromotionRevenue in the EventFact table. The COALESCE function ensures that the query effectively handles null values in both columns. The profit is then aggregated and ranked by various grouping dimensions. The use of GROUPING SETS provides detailed and

flexible insight into the total and average profit by Event Name, Promotion Type, Marketeer Name, and all possible combinations between the three groups. Unlike simple RANK() functions, the PARTITION BY clause had to be defined using the CASE WHEN statement to ensure that each record is ranked in the appropriate group, in descending order of both Total Promotion Profit (TPP) and Average Promotion Profit (APP). However, the row orders are displayed using the TPP. Figure 1 above shows the query, while Figure 2 and Figure 3 below show snippets of the results.

Figure 2

Query1 Result Snippet a

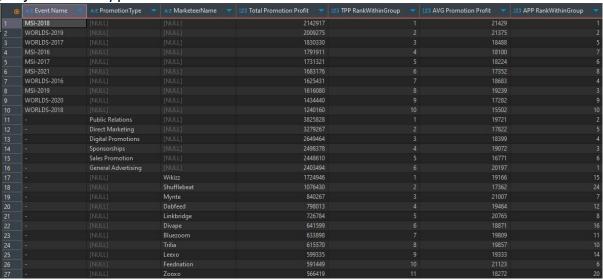


Figure 2 shows the different groups ranked in order of total profit. We can also see the difference in the ranks in terms of TPP and APP, for example, in row 16 we see that General Advertising has the lowest total promotion profit, but performs best on average.

Figure 3

Query1 Result Snippet b

0	** Event Name	AZ PromotionType	▼ AZ MarketeerName	▼ 123 Total Promotion Profit ▼	123 TPP RankWithinGroup	123 AVG Promotion Profit	123 APP RankWithinGroup
103	WORLDS-2016	General Advertising	[NULL]	184965	52	26424	2
104	WORLDS-2020	Public Relations					53
105	MSI-2019	Sponsorships					59
106	MSI-2017	General Advertising		139879			7
107	WORLDS-2018	General Advertising		138960		19851	23
108	MSI-2021	Digital Promotions		134987			58
109	MSI-2019	General Advertising					31
110	WORLDS-2018	Digital Promotions		88961			49
111	WORLDS-2020	Sales Promotion		73710			60
112	MSI-2021		Wikizz	256496			88
113	MSI-2018		Mynte	238025		26447	50
114	MSI-2018						93
115	WORLDS-2018		Wikizz			20467	117
116	MSI-2016			215545		15396	196
117	WORLDS-2019		Wikizz	213990		21399	108
118	MSI-2017		Shufflebeat				140
119	MSI-2019		Mynte			26035	52
120	WORLDS-2017						193
121	MSI-2019		Wikizz	152399		19050	139
122	MSI-2017		Wikizz				176
123	MSI-2018	[NULL]	Linkbridge	150144	12	25024	62

Figure 3 highlights the least performing promotion types and the best the Marketeers across the events.

b. Query 2: Club and Player Attendance by Location.

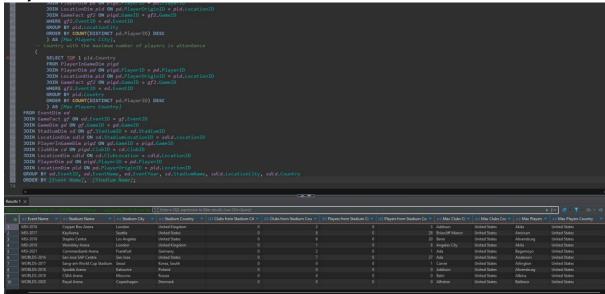
## Figure 4

# Query 2

This query provides insights into the geographic distribution of clubs and players attending World Championships, assisting Tior Games in optimising event planning and marketing strategies. It calculates the number of clubs and players from the same city and country as the stadium, offering insights into local representation. To identify the top contributing regions, the query uses subqueries to accurately determine the city and country with the highest club and player attendance. The results are then grouped by event, stadium, and location, thus offering a clear view of attendance patterns. The idea behind this analysis is to assist Tior Games in targeting future events in high-performing regions, focusing recruitment efforts on talent-rich areas, and optimising marketing campaigns. The query's flexible aggregation and ranking logic ensures accurate insights, empowering Tior Games to make data-driven decisions that enhance the championship's popularity, player engagement, and overall success. Figures 4 and 5 show the query and the results, respectively.

Figure 5

Query 2 continued with Result

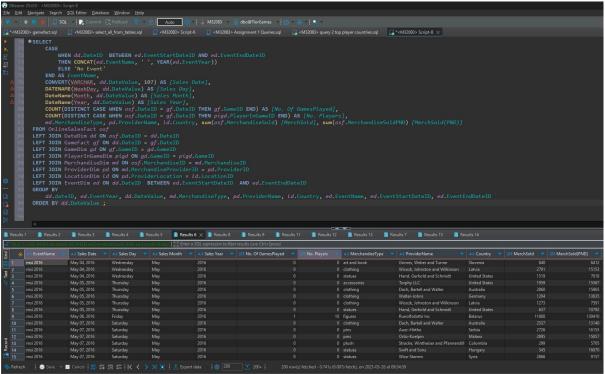


However, the results show that irrespective of the event country, the US consistently exhibits the highest level of participation. In some rows, we have no attendance from the club Country and City.

c. Query 3: Merchandise Sales Performance Analysis.

Figure 6

#### Query 3 and Result Snippet



The query above is designed to provide insights into merchandise sales performance based on the date of sales. It allows us to analyse how merchandise sales vary across different events, months, years, merchandise types, and merchandise providers. One of the key objectives of the query is to determine whether sales occur on dates outside of events. To achieve this, LEFT JOINs are used, ensuring that the query returns sales data even if there are no matching game facts or event dates. However, after reviewing the 952 records in the result, it appears that all online sales occurred during events, although not always on game days. The query also counts the number of players and games to explore potential correlations with merchandise performance — for instance, whether more games or players lead to higher sales. Additionally, it enables us to compare merchandise types and their providers, helping identify which products and suppliers are performing well.

# d. Query 4: Comprehensive Event Sales and Refund Analysis.

Query 4 provides insight into the number of refunds for tickets and merchandise for each event. This offers Tior Games key insights into refund trends. The query calculates the refund rate in quantity and pounds. In addition, the query incorporates 'netsales' columns, showing the calculation of the total revenue after sales and refunds. This provides a foundation for event profitability analysis, facilitating the identification of tickets, merchandise, or events that are particularly susceptible to refunds by examining the refund rates.

# Query 4 and Result

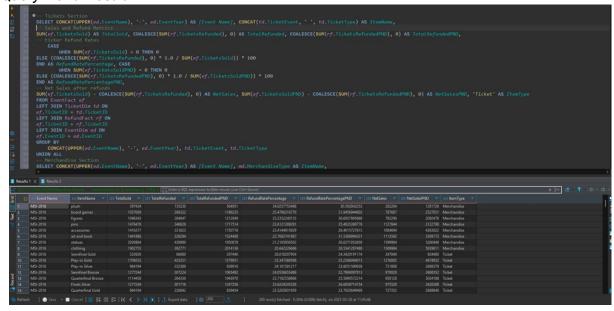
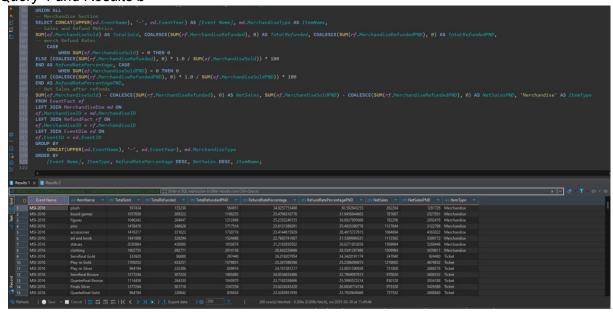


Figure 8

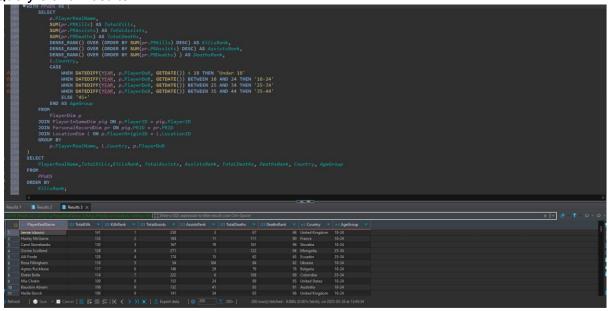
# Query 4 and Results b



e. Query 5: Player Performance And Demographic Summary

Figure 9

### Query 5 and Results



The query aggregates player performance data alongside demographic information such as country and age group, providing essential insights into player behavior. The key insights include metrics like total kills, assists, and deaths, which help identify top performers. The TotalKills and Total Assists are ranked in descending order while the TotalDeaths are ranked in ascending order. Demographic analysis reveals regional trends and preferences based on age, aiding in targeted marketing and game design decisions. Linking performance metrics to demographics allows businesses to tailor features and campaigns to specific audiences. In the results, we can observe that the dominant age groups are 18-24 and 25-34.

The query uses a comprehensive approach by combining performance and demographic data for deeper engagement insights. This information enables Tior Games to craft effective marketing campaigns, enhance gameplay based on player preferences, and improve community engagement. Additionally, it aids in identifying high-performing players for retention strategies and optimising resource allocation.

#### Question 2

Below are the two dimension tables I would suggest for more insights.

1. ProductDim: A dimension table that gives all the information needed about the products Tior Games has ever sold. Below is its data dictionary.

#### **ProductDim**

Column Name	Data Type	Description
ProductID	INT (PK)	Unique ID for each product
ProductName	VARCHAR(50)	Name of the product
		Specific details about Merch (White Tior Games
ProductDescription	VARCHAR(255)	T-Shirt with game character printed on the
		back)
Category	VARCHAR(30)	Product category (loot box, skin, jersey, etc.)
Price	FLOAT	Base price of the product
Currency	VARCHAR(10)	Currency used for the product sale (USD, NGN)
ReleaseDate	DATE	Date when the product was introduced
IsLimitedEdition	BOOLEAN	Flag indicating if the product is limited edition
MerchandiseID	INT (FK)	Foreign key linking to DimMerchandise

Using ProductDim as a bridge dimension table between OnlineSalesFact/RefundFact and MerchandiseDim improves Tior's data warehouse architecture. By separating detailed product information from broader merchandise categories, the schema becomes more normalized and redundancies are eliminated. This additional database allows Tior Games to store product-specific attributes-such as price, currency, and release date, in ProductDim while using MerchandiseDim to represent higher-level categories, such as apparel or figurines, improving data integrity and making the schema more scalable, allowing new products or categories to be added seamlessly without changing the fact table.

From a business perspective, we can also gain deeper insights and more flexible reporting. Tior Games can easily analyse product performance by merchandise category and/or vendor, while tracking sales of limited-edition items and identifying pricing trends across different currencies. Such multi-level aggregation capabilities support more granular trend analysis, such as evaluating the popularity of specific product types (e.g., shirts vs. facecaps) or identifying high revenue merchandise categories and/or vendors. Adding the ProductDim to the db will help Tior Games executives and decision makers gain greater visibility into product sales trends and improve data-driven decisions that can enhance marketing strategies, optimise pricing models, and ultimately increase profitability. Another helpful dimensional table related to sales will be a GameItemDim, which will provide details on perks, skins, gems, etc., that players can use to improve their chances of winning, but this will require a fact table.

2. GameCharacterDim: Another Important dimension table is the GameCharacterDim. As a computer-based Multiplayer Online Battle Arena competition, League of Fun needs to keep track of all its game characters from inception to date. The addition of this dimension table improves the depth of game-related insights. By linking this dimension to PlayerInGameDim, Tior Games can track player performance by character, revealing which characters lead to a higher percentage of kills, assists, or deaths. It also enables detailed player profiling, such as identifying preferred characters, win rates by champion, and playstyle tendencies. The CharacterType and

Special attributes also support meta-analysis, allowing Tior Games to monitor the effectiveness of specific roles (e.g., Tanks vs. Assassins) and assess the impact of character abilities on match outcomes.

Table 2

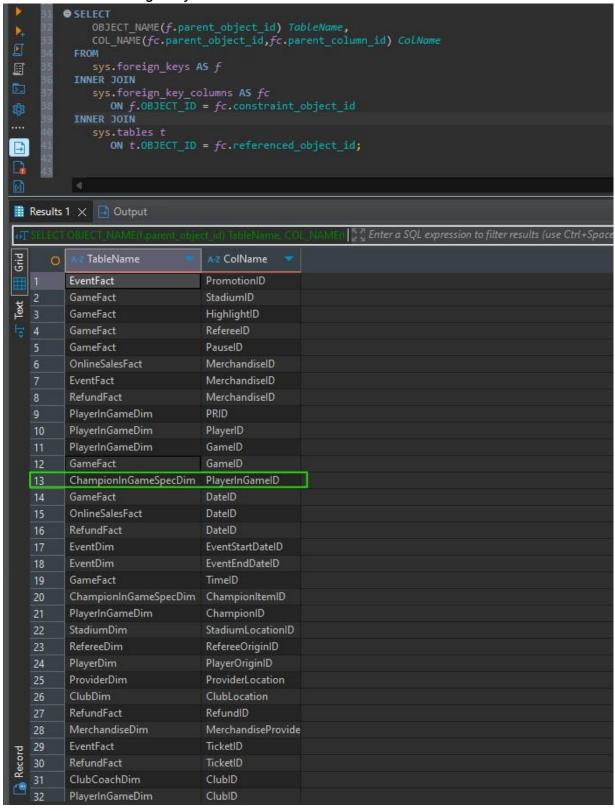
GameCharacterDim

Column Name	Data Type	Description
CharacterID	INT (PK)	Unique ID for each in-game character
CharacterName	VARCHAR(55)	Name of the character
CharacterType	VARCHAR(75)	Character class or role (e.g., Tank, Support, Assassin)
ReleaseDate	DATE	Date the character was introduced in the game
Special Abilities	VARCHAR(255)	Description of unique abilities or powers of the character
BaseHealth	FLOAT	Initial health points of the character
BaseDamage	FLOAT	Initial attack damage of the character

This also aids in gaining insights for game balancing and marketing. For instance, if a particular character is used by a larger percentage of players, it becomes easier to investigate the reasons why using the GameCharacterDim table. This facilitates indepth analysis and character popularity trends by tracking pick, win, and ban rates, which helps developers fine-tune overpowered or underutilised characters. If character attributes are changed frequently, an additional attribute such as drop-date can be added to monitor modifications over time, allowing for a slowly changing dimensions Type 2 approach, and ensuring that we understand character trends among players.

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Table Names and Foreign Keys within the Tables



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ERD Snippet Highlighting PlayerInGameDim and ChampinInGameSpecDim

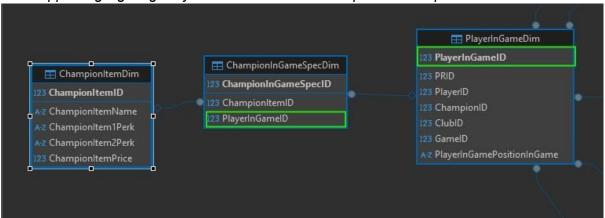
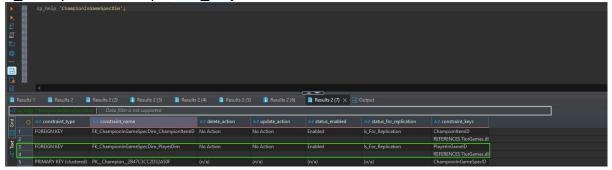


Figure 12

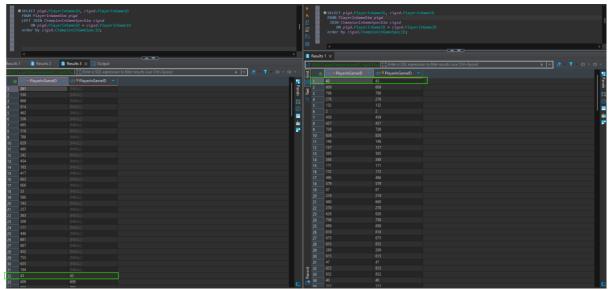
FK\_ChampionInGameSpecDim\_PlayerDim Constraint



c. To resolve this issue, one option is to modify the code to use an inner join instead of a left join. This would exclude any records that are not present in both tables. Figure 13 illustrates the difference between using a left join and an inner join.

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### Difference between Left Join and Inner Join



However, a more permanent solution would be to enforce referential integrity, which prevents null values from appearing in a child table like ChampionInGameSpecDim. Additionally, it would be important to define the actions to be taken when insert, update, or delete operations are performed on the parent table, PlayerInGameDim. This could be achieved using triggers or specifying CASCADE, SET DEFAULT, or NO ACTION when creating the foreign key constraint.

#### References

Sherman, R. (2014). Business intelligence guidebook: From data integration to analytics. Newnes.