

Data Warehousing Assignment

This problem set consists of two data modeling scenarios. You will be asked to analyze the strengths and weaknesses of some design alternatives for each scenario. Short answers are fine – one or two paragraphs per question would be an appropriate length.

Scenario I In this scenario, we are interested in modeling student enrollment in Stanford courses. We would like to answer questions such as:

- Which courses are most popular? Which instructors are most popular?
- Which courses are most popular among graduate students? Undergraduates?
- Are there courses for which the assigned classrooms is too large or too small?

We are planning to have a course enrollment fact table with the grain of one row per student per course enrollment. In other words, if a student enrolls in 5 courses there will be 5 rows for that student in the fact table. We will use the following dimensions: Course, Department, Student, Term, Classroom, and Instructor. There will be a single fact measurement column, EnrollmentCount. Its value will always be equal to 1.

We are considering several options for dealing with the Instructor dimension. Interesting attributes of instructors include FirstName, LastName, Title (e.g. Assistant Professor), Department, and TenuredFlag. The difficulty is that a few courses (less than 5%) have multiple instructors. Thus it appears we cannot include the Instructor dimension in the fact table because it doesn't match the intended grain. Here are the options under consideration:

Option A

Option B

Option C

Modify the Instructor dimension by adding special rows representing instructor teams. For example, CS276a is taught by Manning and Raghavan, so there will be an Instructor row representing “Manning/Raghavan” (as well as separate rows for Manning and Raghavan, assuming that they sometimes teach courses as sole instructors). In this way, the Instructor dimension becomes true to the grain and we can include it in the fact table.

Change the grain of the fact table to be one row per student enrollment per course per instructor. For example, there will be two fact rows for each student enrolled in CS 276a, one that points to Manning as an instructor and one that points to Raghavan. However, each of the two rows will have a value of 0.5 in the EnrollmentCount field instead of a value of 1, in order to allow the fact to aggregate properly. (Enrollments are “allocated” equally among the multiple instructors.)

Create two fact tables. The first has the grain of one row per student enrollment per course and doesn't include the Instructor dimension. The second has the grain of one row per student enrollment per course per instructor and includes the Instructor dimension (as well as all the other dimensions). Unlike Option B, the value of

EnrollmentCount will be 1 for all rows in the second fact. Tell warehouse users to use the second fact table for queries involving attributes of the instructor dimension and the first fact table for all other queries.

Please answer the following questions.

Question 1. What are the strengths and weaknesses of each option?

Answer.

Strengths of each option:

- A. at granular level perspective It would be better If we have multiple instructors in one row for particular course.
- B. As in this option we are populating the data in two rows individually there would not be any confusion. As we know that there is not much records which has multiple instructors.
- C. In this case the best point is two different entities (student enrollment related data as well instructor) are separating into two different fact tables, because of this the drawbacks of both the options A and B will overcome, as the data is segregated individually independent of the relation dependency on each other.

Weaknesses of each option:

- A. There is no need to add multiple rows mentioning , Instructor row representing “Manning/Raghavan” (as well as separate rows for Manning and Raghavan, assuming that they sometimes teach courses as sole instructors.
- B. in this case we have to add two separate rows mentioning multiple instructors.
- C. In this case as we need to create two separate fact tables mentioning The first has the grain of one row per student enrollment per course and doesn’t include the Instructor dimension. The second has the grain of one row per student enrollment per course per instructor and includes the Instructor dimension (as well as all the other dimensions) due to this to fetch information we need to use join operations on both tables to get info. Which is costly operation.

Question 2. Which option would you choose and why?

Answer. Option C , I would like to go with the reason is as we know the records which has multiple instructor is less than 5% so there will not be any memory issue. So its better to add multiple rows instead of going for option c which includes join operation which will be costly.

Question 3. Would your answer to Question 2 be different if the majority of classes had multiple instructors? How about if only one or two classes had multiple instructors? (Explain your answer.)

Answer. No. because even majority of classes has multiple instructors it would be better option, and if only one or two classes has multiple instructors then I would like two go with option B because adding multiple rows is more efficient.

Question 4. [OPTIONAL] Can you think of another reasonable alternative design besides Options A, B, and C? If so, what are the advantages and disadvantages of your alternative design?

Scenario II In this scenario, we are building a data warehouse for an online brokerage company. The company makes money by charging commissions when customers buy and sell stocks. We are planning to have a Trades fact table with the grain of one row per stock trade. We will use the following dimensions: Date, Customer, Account, Security (i.e. which stock was traded), and TradeType.

The company’s data analysts have told us that they have developed two customer scoring techniques that are used extensively in their analyses.

- Each customer is placed into one of nine Customer Activity Segments based on their frequency of transactions, average transaction size, and recency of transactions.

- Each customer is assigned a Customer Profitability Score based on the profits earned as a result of that customer's trades. The score can be either 1,2,3,4, or 5, with 5 being the most profitable.

These two scores are frequently used as filters or grouping attributes in queries. For example:

- How many trades were placed in July by customers in each customer activity segment?
- What was the total commission earned in each quarter of 2003 on trades of IBM stock by customers with a profitability score of 4 or 5?

There are a total of 100,000 customers, and scores are recalculated every three months. The activity level or profitability level of some customers changes over time, and users are very interested in understanding how and why this occurs.

We are considering several options for dealing with the customer scores:

Option A

The scores are attributes of the Customer dimension. When scores change, the old score is over written with the new score (Type 1 Slowly Changing Dimension).

Option B

The scores are attributes of the Customer dimension. When scores change, new Customer dimension rows are created using the updated scores (Type 2 Slowly Changing Dimension).

Option C

The scores are stored in a separate CustomerScores dimension which contains 45 rows, one for each combination of activity and profitability scores. The Trades fact table includes a foreign key to the CustomerScores dimension.

Option D

The scores are stored in a CustomerScores outrigger table which contains 45 rows. The Customer dimension includes a foreign key to the outrigger table (but the fact table does not). When scores change, the foreign key column in the Customer table is updated to point to the correct outrigger row.

Please answer the following questions.

Question 5. What are the strengths and weaknesses of each option?

Answer.

Strengths of each option:

- If at the end we are not required to maintain the previous history scores of customer then the strength of this option is the complexity is reduced no need to create separate dimension which increases complexity with join operation.
- As mention above the strength is same. Even we can have incremental data so we have store the past history and we can use it later for analysis.
- In this case if we have to talk about strength is that, as we knows the combination of activity (1-9) and scores (1-5) both is 45 so we separate as chart in separate dimension and its foreign key is in fact table so we can map it easily.
- In this case in above case just we have attribute in customer dimension and the outrigger table has no direct connection with fact table.

Weaknesses of each option:

- As the new score is overwritten with old one we cannot trace the history of scores.
- Only for the updated score complete rows with except the score column values all values has to append every after 3 months which may cause memory issue.
- As fact table directly as foreign key for an separate CustomerScores dimension we can not maintain consistency
- Weekness is just as we have to add one more CustomerScores outrigger table.

Question 6. Which option would you choose and why?

Answer. I would like to go with option D as we are maintain separate CustomerScores outrigger table and customer dimension has the foreign key for this table so we can maintain consistency.

Question 7. Would your answer to Question 6 be different if the number of customers and/or the time interval between score recalculations was much larger or much smaller? (Explain your answer.)

Answer. No. because irrespective of customer count this solution is ideal. As we have separate CustomerScores outrigger table.

Question 8. [OPTIONAL] Can you think of another reasonable alternative design besides Options A, B, C, and D? If so, what are the advantages and disadvantages of your alternative design?