**ADVANDB MCO2: Data Warehousing and Online Analytical Processing (OLAP)**

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**1. INTRODUCTION**

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**2. DIMENSIONAL MODEL**

The schema that has been created for this project mainly deals with land and agricultural data as the primary facts. From the CBMS database, the hpq\_hh, hpq\_mem, hpq\_crop, hpq\_alp, and hpq\_arcdp\_mem tables were transformed into the group’s constellation schema, the former two holding most of the dimensions and the latter three holding most of the facts. The entire schema can be seen in Figure 1.

**2.1. Contents**

There are three fact tables in the schema: the crop table, the arcdp table, and the land\_parcel table.

The crop table contains facts regarding the crops produced by a certain household. The measure of this fact table is the crop volume crop\_vol, which will be aggregated using sum and average.

The arcdp table contains information about the Agrarian Reform Community Development Program beneficiaries. The measures of this table are the age, calculated grade, and daily work hours of the beneficiary. All values will be averaged.

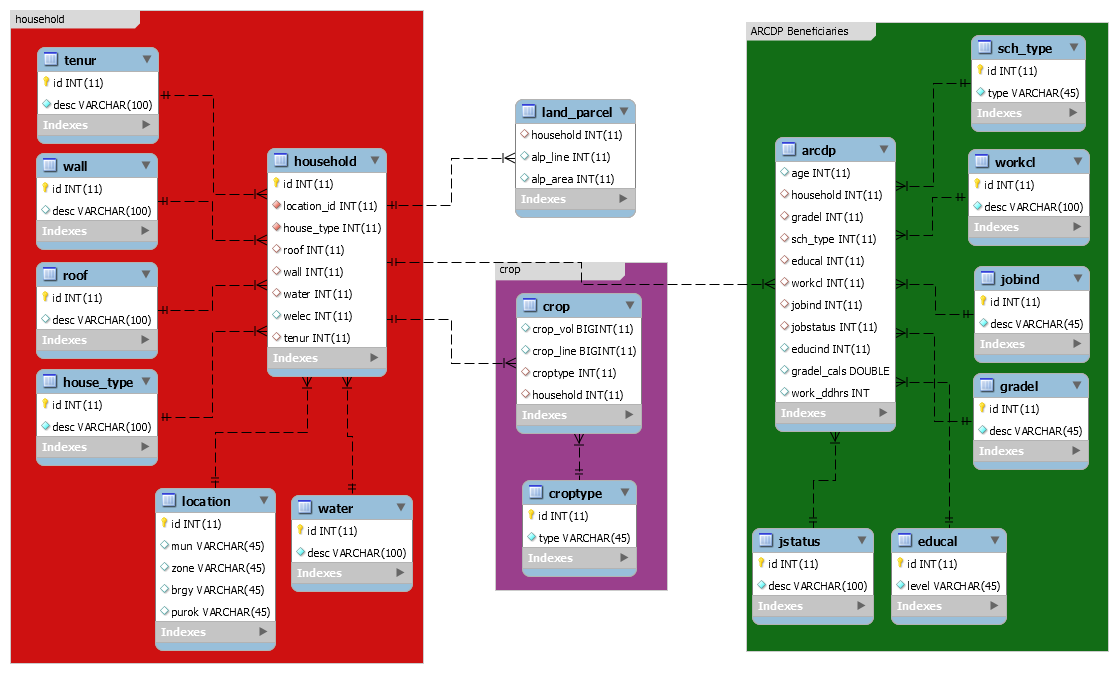
The land\_parcel table contains facts about the land owned by the households. The measure of this table is the land area, which will be aggregated using sum and average functions.

**2.2. Dimensions**

The dimensions that were used were categorical columns from the selected tables in the CBMS database. For the crop table, “croptype” is the type of crop farmed, which may be Sugarcane, palay, corn, coffee, or other.

For the households, ”house\_type” is the type of house, which may be single house, duplex, multi-unit residential, commercial/ industrial/agricultural/house, or others. “location” comprises the unique municipality, barangay, zone, and purok of the household. “wall” refers to the type of wall and “roof” is the type of roof of a household, both of which may be strong materials, light materials, salvaged materials, mixed but predominantly strong materials, mixed but predominantly light materials, mixed but predominantly salvaged materials, or not applicable. “tenur” is the tenure of the house owner, which can be owner of lot; renting lot; owner of house, renting lot; owner of house, not renting lot with consent of owner; owner of house, not renting lot without consent of owner; rent-free house and lot with consent; rent-free house and lot without consent; living in public space with rent; living in public space without rent; and other. “water” is the household’s source of water which may be own faucet, shared faucet, own deep well, shared deep well, shallow well, dug well, protected spring, unprotected spring, natural, peddler, bottled water, or others.

For the beneficiaries, “jobind” is whether the arcdp beneficiary is employed or unemployed., “jstatus” is the current status of the beneficiary at their job, which may be permanent, short-term, or working on different jobs week to week. “sch\_type” is whether the beneficiary is at a public or private school. “workcl” is what kind of employee the beneficiary is which may be for private household, for private business, for government, self-employed with no employee, employer in family business, working with pay on family-owned business, or working without pay on family-owned business. “educal” is the highest educational attainment and “gradel” is the current grade which may be none, day care, preschool, grades 1 to 12, 1st to 3rd year PSPS/N-T/TV, 1st to 4th



**Figure 1 – The Constellation Schema**

year college, post grad, ALS elementary or secondary, SPED elementary or secondary, Grade school, High School, Post secondary, college, or MD/PhD.

To distinguish which columns to use and under what context, a rudimentary understanding of statistics was used. For each table, having joined the hpq\_hh and hpq\_mem tables with the hpq\_crop, hpq\_alp, and hpq\_arcdp\_mem tables, the four classifications of variables: nominal, ordinal, interval, and ratio (Types of Variable, n.d.); were used to distinguish between dimension and fact column.

Nominal variables, which may be numeric or non-numeric, represent categories but have no intrinsic order. In the constellation schema, these were translated into the dimension tables croptype, house\_type, jobind, jstatus, location, roof, sch\_type, tenur, wall, water, and workcl. Ordinal variables also represent categories, but have an intrinsic order, such as the dimensions educal, and gradel. Since it would make no sense to perform aggregate functions on these types of variables since they hold no meaningful intrinsic value, it is better to classify the more meaningful data according to these types of variables, which means they are more suitable as dimensions.

Interval and Ratio variables do have a meaningful value, their only difference being that a zero in a ratio variables means that there is no quantity represented, a true zero. As such, there is meaning in adding and subtracting these values, which would make them ideal for aggregate functions such as sum and average. These variables ended up being the measures in the fact table such as crop\_vol, alp\_area, age, gradel\_calc, and workddhrs.

**2.3. Schema and Issues**

A constellation schema was implemented, with the primary fact tables being land\_parcel, crop, and arcdp. The crop table references a croptype table. All three fact tables reference household which itself references the tenur, wall, roof, house\_type, location, and water tables. The arcdp table also references the sch\_type, workcl, jobind, gradel, educal, and jstatus tables.

A constellation was selected to allow for more than one analytical view for data. Since the original CBMS database had three tables pertaining to agricultural information, a constellation was a prime opportunity to extract as much information as possible.

The initial issue with the design was that, ideally, a data warehouse would have fact tables referencing other dimension tables that have more data about that particular dimension. For example, a sales table bay have a branch dimension, but the branch table would have additional location and manager attributes. The original CBMS database did have a semblance of a star schema but , the most common dimension tables referenced were only two, although three were required for the MCO. After consultation, it was said that extracting individual data columns from the tables could account for more dimensions, so the aforementioned columns in section 2.2. were extracted into tables. The next problem is the lack of additional data in these tables. Unlike the branch table, the only other data in these tables is a descriptor, which would limit the possibilities of future slice and dice operations. Since the original CBMS database had no further data, this was simply accepted as a limitation.

**3. ETL PROCESS**

**3.1. Code Description**

The team decided to create their own ETL tool using MySQL. The ETL process was done using a Java extraction program and a MySQL transformation and loading script. The Java program read from the CSV files, loading them into the original CBMS schema. The resultant schema was then queried and manipulated to transform its data into the new schema. The results of the queries were loaded into the new schema using INSERT INTO… SELECT statements.

**3.2. Extraction**

The original data source were seventeen comma-separated values files, each file representing one relation. The Java extraction tool coded by one of the team members simply extracted the table name and all the column names to be used in the INSERT INTO <table name> (<col 1>[,<col2>...]). It then read each row to be added to the VALUES <tuple1>... part. After this, it established a connection with the database and executed the INSERT statements. This program was run for each csv file until the entire schema was loaded.

A data dictionary, which listed each table and the columns, along with a column description,  was also provided. This was used to generate the original empty schema.

The resultant database was queried in the next step, which constitutes the extraction step.

**3.3. Transformation**

For most of the dimension tables, the data dictionary was consulted in populating the tables with their actual meaning.

For the location, however, each location is a unique combination of municipality, barangay, zone, and purok. Each combination was given a unique id by giving the new table an autoincrement primary key and executing the following script.

INSERT INTO location(mun, zone, brgy, purok)

SELECT distinct mun, zone, brgy, purok FROM hpq\_hh;

For the household table, the location fields had to be transformed into the new ones generated by the previous query, so joining with the location table was required. The script is as follows.

INSERT INTO household

SELECT H.id, L.id AS location, house\_type,roof,wall,water,welec,tenur

FROM db\_hpq.hpq\_hh H INNER JOIN location L

ON H.mun = L.mun AND H.zone = L.zone AND H.brgy = L.brgy AND H.purok = L.purok

GROUP BY H.id;

For the crop and land\_parcel tables, they simply had to be loaded into the new database after projecting off some unnecessary columns.

The query for crop is

INSERT INTO crop(crop\_vol,crop\_line,croptype,household)

SELECT crop\_vol, crop\_line, croptype, hpq\_hh\_id AS household

FROM db\_hpq.hpq\_crop C INNER JOIN household H ON C.hpq\_hh\_id = H.id;

The query for land\_parcel is

INSERT INTO land\_parcel

SELECT id AS household, alp\_line, alp\_area

FROM db\_hpq.hpq\_alp A INNER JOIN household H

ON a.hpq\_hh\_id = H.id;

For the arcdp table, each beneficiary had to be associated with the community member, so the hpq\_arcdp\_mem table had to be joined with the hpq\_mem table. After the join, only a select few columns had to be retained. The query is as follows:

INSERT INTO arcdp

SELECT age, household, gradel, sch\_type, educal, workcl, jobind, jstatus, educind, gradel\_calc, work\_ddhrs

FROM (SELECT A.hpq\_hh\_id AS household, arcdp\_mem\_refno as memno

FROM db\_hpq.hpq\_arcdp\_mem A INNER JOIN household H

ON H.id = A.hpq\_hh\_id) X

INNER JOIN db\_hpq.hpq\_mem M

ON M.memno = X.memno AND M.id = X.household;

**3.4. Loading**

The new schema had primary keys and foreign key, which would enforce key integrity and referential integrity. During the loading of the three fact tables, MySQL complained about foreign key constraint violation. This would only be because the hpq\_crop, hpq\_arcdp\_mem, and hpq\_alp tables had rows that were not connected to any row in hpq\_hh. Since nothing could be done about the missing data in hpq\_hh, an INNER JOIN between the table in question and the hpq\_hh table was performed, eliminating the rows in the three peripheral tables that were not referencing an existent household.

**4. OLAP QUERIES**

To query the data warehouse, a QueryBuilder class was created, which queries the constellation schema based on the dimensions selected for drill down/rollup and slice and dice.

**4.1. Sample Queries**

For the purposes of discussion, assume the base OLAP query is viewing the average age, grade, and work hours of all ARCDP beneficiaries classified by highest educational attainment. The query for this as generated by QueryBuilder is

SELECT jobind.desc AS jobind ,educal.desc AS educal ,AVG(age) AS avgAge,AVG(gradel\_calc) AS avgGrade,AVG(work\_ddhrs) AS avgWorkHrs

FROM (SELECT \* FROM ARCDP A INNER JOIN household H ON A.household = H.id) X INNER JOIN jobind ON X.jobind = jobind.id INNER JOIN educal ON X.educal = educal.id

GROUP BY jobind,educal

ORDER BY avgAge DESC,avgGrade DESC,avgWorkHrs DESC

Which yields 41 rows, the first 10 of which are in Table 1.

**Table 1 – ARCDP Beneficiaries Classified by Job Index and Highest Educational Attainment**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Job Index** | **Highest Educational Attainment** | **Average Age** | **Average Grade** | **Average Work Hours** |
| Employed | Grade 5 | 69.0000 | 499 | 8.0000 |
| Employed | 4th year College or higher | 63.0000 | 499 | 16.0000 |
| Employed | 2nd year College | 62.0000 | 499 | 9.3333 |
| Employed | Post grad with units | 61.0000 | 499 | 8.0000 |
| Unemployed | 2nd year PS PS/N-T/TV | 60.0000 | 499 | 1.0000 |
| Employed | Grade 3 | 58.0000 | 499 | 8.1667 |
| Unemployed | College graduate | 55.0000 | 499 | 3.2000 |
| Employed | Grade 6 | 54.0000 | 499 | 8.2963 |
| Employed | College graduate | 50.0000 | 499 | 8.2222 |
| Employed | Grade 8 | 48.6000 | 499 | 8.7000 |

As can be seen here, the group with the oldest beneficiaries are the employed group whose highest educational attainment is fifth grade.

*4.1.1. Roll-up*

If this data is too comprhensive, we can roll up on either of the columns. Let both be demonstrated.

Rolling up on job index, we get the following query.

SELECT educal.desc AS educal ,AVG(age) AS avgAge,AVG(gradel\_calc) AS avgGrade,AVG(work\_ddhrs) AS avgWorkHrs

FROM (SELECT \* FROM ARCDP A INNER JOIN household H ON A.household = H.id) X INNER JOIN educal ON X.educal = educal.id

GROUP BY educal

ORDER BY avgAge DESC,avgGrade DESC,avgWorkHrs DESC

Which yields 24 rows, the top 10 of which are in Table 2.

**Table 2 - ARCDP Beneficiaries Classified by Highest Educational Attainment**

|  |  |  |  |
| --- | --- | --- | --- |
| **Highest Educational Attainment** | **Average Age** | **Average Grade** | **Average Work Hours** |
| 4th year College or higher | 63.0000 | 499 | 16.0000 |
| 2nd year PS PS/N-T/TV | 60.0000 | 499 | 5.7500 |
| Post grad with units | 54.5000 | 499 | 2.6667 |
| College graduate | 53.1250 | 499 | 5.5789 |
| Grade 6 | 48.1667 | 499 | 6.3784 |
| 2nd year College | 47.7500 | 499 | 5.1429 |
| Grade school graduate | 47.0000 | 499 | 4.0000 |
| Grade 9/(3rd Year HS | 43.8000 | 499 | 519.0000 |
| High school graduate | 41.5333 | 499 | 5.0526 |
| Grade 10/(4th Year HS | 40.6667 | 499 | 2.8750 |

From this data, we can see that the highest average age belongs to the group of people whose highest educational attainment is 4th year college. The group that works the most, however, is the group whose highes educational attainment is Grade 9 or 3rd year high school.

Rolling up on Highest Educatoinal Attainment, we get the following query:

SELECT jobind.desc AS jobind ,AVG(age) AS avgAge,AVG(gradel\_calc) AS avgGrade,AVG(work\_ddhrs) AS avgWorkHrs

FROM (SELECT \* FROM ARCDP A INNER JOIN household H ON A.household = H.id) X INNER JOIN jobind ON X.jobind = jobind.id

GROUP BY jobind

ORDER BY avgAge DESC,avgGrade DESC,avgWorkHrs DESC

Which results in 2 rows shown in Table 3.

**Table 3 - ARCDP Beneficiaries Classified by Job Index**

|  |  |  |  |
| --- | --- | --- | --- |
| **Job Index** | **Average Age** | **Average Grade** | **Average Work Hours** |
| Employed | 49.5610 | 499 | 56.7143 |
| Unemployed | 30.2656 | 499 | 0.7838 |

It can be seen here that employed beneficiaries are much older than the unemployed ones, which may indicate that younger people in their focus group may tend to be unemployed.

*4.1.2. Drill-down*

Any of the other ten dimensions of the ARCDP query can be utilized. This operations shall be demonstrated with the owner tenure and work class dimensions.

Drilling down from the original query on owner tenure, we get the following:

SELECT tenur.desc AS tenur ,jobind.desc AS jobind ,educal.desc AS educal ,AVG(age) AS avgAge,AVG(gradel\_calc) AS avgGrade,AVG(work\_ddhrs) AS avgWorkHrs

FROM (SELECT \* FROM ARCDP A INNER JOIN household H ON A.household = H.id) X INNER JOIN tenur ON X.tenur = tenur.id INNER JOIN jobind ON X.jobind = jobind.id INNER JOIN educal ON X.educal = educal.id

GROUP BY tenur,jobind,educal

ORDER BY avgAge DESC,avgGrade DESC,avgWorkHrs DESC

Which results in 86 rows, 10 of which have been included in Table 4.

**Table 4 - ARCDP Beneficiaries Classified by Owner Tenure, Job Index and Highest Educational Attainment**

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| **Owner Tenure** | **Job Index** | **Highest Educational Attainment** | **Average Age** | **Average Grade** | **Average Work Hours** |
| Living in a public space without rent | Employed | Grade 4 | 69.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Employed | Grade 5 | 69.0000 | 499 | 7.3333 |
| Owner, owner-like possession of house and lot | Employed | 4th year College or higher | 63.0000 | 499 | 16.0000 |
| Owner, owner-like possession of house and lot | Employed | 2nd year College | 62.0000 | 499 | 9.3333 |
| Rent-free house and lot with consent of owner | Unemployed | Grade 4 | 62.0000 | 499 | 0.0000 |
| Owner, owner-like possession of house and lot | Employed | Post grad with units | 61.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Unemployed | 2nd year PS PS/N-T/TV | 60.0000 | 499 | 1.0000 |
| Owner, owner-like possession of house and lot | Employed | Grade 3 | 58.0000 | 499 | 7.0000 |
| Owner, owner-like possession of house and lot | Employed | 3rd year College | 55.0000 | 499 | 9.0000 |
| Living in a public space without rent | Employed | Grade 6 | 55.0000 | 499 | 8.0000 |

It can be seen here that their oldest beneficiaries are living in a public space without rent, but employed and have achieved at least grade 4.

Drilling down further on work class, the resultant query is :

SELECT tenur.desc AS tenur ,workcl.desc AS workcl ,jobind.desc AS jobind ,educal.desc AS educal ,AVG(age) AS avgAge,AVG(gradel\_calc) AS avgGrade,AVG(work\_ddhrs) AS avgWorkHrs

FROM (SELECT \* FROM ARCDP A INNER JOIN household H ON A.household = H.id) X INNER JOIN tenur ON X.tenur = tenur.id INNER JOIN workcl ON X.workcl = workcl.id INNER JOIN jobind ON X.jobind = jobind.id INNER JOIN educal ON X.educal = educal.id

GROUP BY tenur,workcl,jobind,educal

ORDER BY avgAge DESC,avgGrade DESC,avgWorkHrs DESC

Which returns 98 rows, 10 of which are in Table 5.

**Table 5 - ARCDP Beneficiaries Classified by Owner Tenure, Work Class, Job Index and Highest Educational Attainment**

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Owner Tenure** | **Work Class** | **Job Index** | **Highest Educational Attainment** | **Average Age** | **Average Grade** | **Average Work Hours** |
| Owner, owner-like possession of house and lot | Working for government/corporation | Employed | High school graduate | 75.0000 | 499 | 8.0000 |
| Living in a public space without rent | Self-employed without any employee | Employed | Grade 4 | 69.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Working for private household | Employed | Grade 5 | 69.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Self-employed without any employee | Unemployed | College graduate | 67.5000 | 499 | 7.3333 |
| Owner, owner-like possession of house and lot | Working for private business/establishment/farm | Employed | 4th year College or higher | 63.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Self-employed without any employee | Employed | 2nd year College | 62.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Working for private household | Employed | Grade 3 | 61.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Working for private household | Employed | Post grad with units | 61.0000 | 499 | 8.0000 |
| Owner, owner-like possession of house and lot | Self-employed without any employee | Unemployed | 2nd year PS PS/N-T/TV | 60.0000 | 499 | 1.0000 |
| Owner, owner-like possession of house and lot | Employer in own family-operated or business | Employed | High school graduate | 57.0000 | 499 | 8.0000 |

From this data, it could be deduced that their oldest beneficiaries are government employees who own their house and lot, are employed, and are at least high school graduates.

*4.1.3. Dice and Slice*

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**4.2. Queries**

*4.2.1. Crop Analytics*

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*4.2.2. Land Parcel Analytics*

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*4.2.3. ARCDP Analytics*

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**5. RESULTS AND ANALYSIS**

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**6. CONCLUSION**

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