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Business Intelligence for Construction Company Acknowledgement Reporting System

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Abstract—As the one of Indonesian association of construction service companies, the association ensure to recommend and record every construction service company (their members) to works according to valid Indonesian law and constitution. The problem in this association, as the job is to give detailed reports to government, investor either domestic or international, and every institutions that need information such as market needs, resource developments, and improvement of construction technologies, is to create queries for every data from tables in MySOL to create reports quickly, efficiently, and automatic. Example of reporting needs are: In what month construction project done at the most within 1 year; Project duration, project cost, and experience at the most for each construction company; or amount of projects based on source of funds (APBN, APBD, LOAN PEMERINTAH, BUMN, or SWASTA) for each construction companies. Recommended solution is using Data Warehouse where it has structured storage and can analyze data with OLAP that has capability to process and visualize data at high speed. To construct the Data Warehouse, the design used Kimball Method so this association company will have capability to generate report according to every instances' needs.

Keywords—Data Warehouse, kimball, pentaho, report, Qlik Sense

I. INTRODUCTION

Construction service companies in Indonesia have lots of potentials, where the investments, either governmental or private (service clients) increases every years. This also connected to growth rate of construction service companies in Indonesia from small, medium, and big scale companies. By observe market needs, resource developments, and improvement of construction technologies, they are expected to expand the company's capability building. by expanding the company, it will open chance of employments, improvement of public infrastructure, and chance for local constructions to compete with domestic and international markets. Therefore, construction service companies are part

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of factors that affect Indonesia's development that equally important with education, economy, and culture developments.

These information are valuable assets to every companies, because knowledge has important role in guiding business analysts to take decisions[1]. And that is why, these assets needs to be secured and easy to be access in times they needed even in case of security breach[2]. But data available are so many that affect extraction of information to be difficult. Data Warehouse is a phenomenon of large collections of data processed into information from certain period of years and from that, to use data for reporting needs.

In this study case of association for construction service companies, the association work to recommend and register construction company in Indonesia, where this association already legitimated by government's ministry of public works and public housing since 2002, this association currently did not have reporting system that capable of quick, efficient, and automatic way to generate reports from database. Because of that limit, time needed for generate report is relatively too long. Currently, every reports have been done by export database from MySQL manually to Microsoft Excel file, then analyze the Excel file directly. Which is why Data Warehouse is needed to accommodate appropriate, quick, and easy to understand reporting.

By implement Data Warehouse and OLAP Analysis based on association needs, it is to be expected that analysis result capable to represent data from different reports where association might be need to take decision based from the data. The decision making need few historical data to be analyzed to support or object hypothesis and to take decision from the analysis [3].

Methodology used in this solution is using Kimball Method. Kimball Method is a system where transactional data collected from every sources using queries to analyze data. Few terms used for technologies for data transaction



and analysis are OLTP (Online Transactional Processing) and OLAP (Online Analytical Processing). OLTP and OLAP have important role in data transaction and analysis. This ideas from Ralph Kimball still used in current implementation of Data Warehouse [4].

II. RELATED WORKS

Business Intelligence (BI) is a capability or knowledge to process data gathered and stored in order to provide business with information that gives advantages in making decisions. BI requires data and tools to store historical data and process them for decision making. This will help business to expand itself by following decisions based from business company's data[5].

Data Warehouse is relation between databases constructed for query processes that aim for ease data reporting and analysis. Data Warehouse capable to help organization to analyze trend based on repository of organization's data in a period of time. Main functions of Data Warehouse are to facilitate organization in strategic planning for long term from data saved and to make quick decision that appropriate to the problem. Modeling dimension is used to implement Data Warehouse [6]. Modeling dimension is a data structure technique optimized and created to read, summary, and analyze numeric information such as value, sum, weight, etc. Few limitations to do Data Warehouse is required storage space and memory capacity to process these data into simplified views [7].

Star Schema is a model that still in use until now to represent multidimensional data that include 2 kinds of table, such as Fact table and Multidimensional table. these tables will undergo denormalization, to form a diagram of Entity Relationship Diagram (ERD) shaped like a star. this schema is drawn in relationship where fact table is placed in the middle of several Dimension tables [8]. Fact table is the table that store the facts, in this case, all historical data that will be used in analysis process. Then the dimension table is the table contains attributes that supports the facts[9].

Data Warehouse or usually called Online Analytical Processing (OLAP) is oriented to market compared to conventional OLTP. OLAP is a basic to create a Decision Support System (DSS) where this tool is used for create complex decision. OLAP can be integrated from many sources of data and capable to generate multidimensional reports that used for data analysis and decision making [10]. There are 3 variances from extension of OLAP, they are: Relational OLAP (ROLAP), Multidimensional OLAP (MOLAP), and Hybrid OLAP (HOLAP). ROLAP act as interface between Relational Database Management System with OLAP users. MOLAP store data from OLTP database in the cube. HOLAP combines both of them [11].

OLAP Cube is a feature from OLAP that connect every dimensions into fact table. This will help decision making and extraction of data in multidimensional way based on hierarchy of each dimensions for gaining information relevant based on selected level of details. Data Cube stored in cell with structure resemble a 3D spreadsheet [12]. Operations used in this method is Aggregation operations used to aggregate cell within the cube. Aggregate operations used are SUM, AVG, COUNT, MIN, and MAX. Example of syntax used for aggregate operation is : AggFunction (CubeName, Measure) [By Dimension*].

ETL (Extraction, Transformation, and Loading) is an order of steps in data processing from database that involve process of extraction data. by maintain data quality, preserve data standards, and reform data in certain shape (Transformation) later to be presented (Loading) to Data Warehouse [13].

Pentaho is open source application to execute business analysis tasks, ETL, dashboard, and reporting. Pentaho ease these tasks of business analysis with solutions in application to create and develop business analysis project [14]. ETL process in Pentaho is using application called Pentaho Data Integration to make data extraction and transformation process easy. For design of OLAP Cube is using Mondrian with Pentaho Workbench Schema. Then to design report from ETL process is using Pentaho Report Designer.

Qlik Sense is an interactive application for analyze and visualize large amount of data to create presentation, dashboard, and reporting. This application is easy to use because of drag and drop components feature. Qlik Sense is available in shape of desktop application (Qlik Sense Desktop) and cloud hosting (Qlik Sense Cloud) that free to use. For company usage, they provide Qlik Sense Enterprise that removes limitations in free version [15].

III. PROPOSED METHOD

In this part, there are steps that explained later in this paper to solve problems in association company. Starting from data collection and analysis, designing star schema, ETL process, create a OLAP cube, then finally, design reports and dashboards.

Start with data collection, every data that support in the case study is analyzed to produce information about calculations and measurements that will be shown in OLAP analysis, report design, and indicators to be shown inside dashboard. These data will become reference to develop the Data Warehouse. Data sources that used to develop Data Warehouse are from operational database taken from Relational Database Management System (RDBMS) MySQL and tables shown in ERD of Figure 1.

With these data sources, a step of extraction data will be done to extract data from OLTP database into OLAP database. these data will be reform into filtered data that used for analysis. Data will be divided into fact and dimensions. Dimensions are used for making categories that group the data from fact. The fact in other side will have role as main source of information to be analyzed. The design of data is represented by star schema in Figure 2. the star schema is composed with 1 fact table and 5 dimension tables.

Based on fact and dimensions in star schema data design, a step of ETL will be done using Pentaho Data Integration application. The ETL Process start from extracting data from OLTP database by selecting tables that possess required data. Then, select columns of data that needed for OLAP analysis. Data will be given with surrogate key that represent the identity for each data. Last, these data will be stored in table inside OLAP database. Figure 3,4,5,6,7 in order are ETL steps to create dimension tables. Start from construction company, category, source of funds, place, and time.



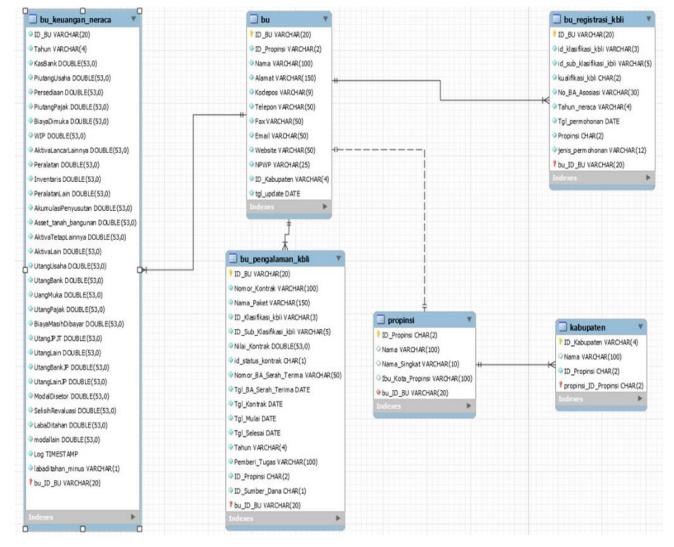


Fig. 1. ERD (Entity Relationship Diagram)

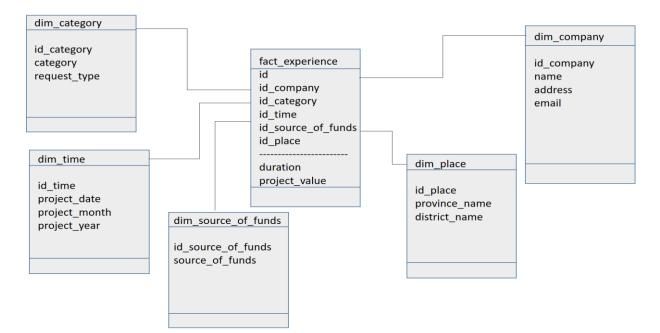


Fig. 2. Star Schema

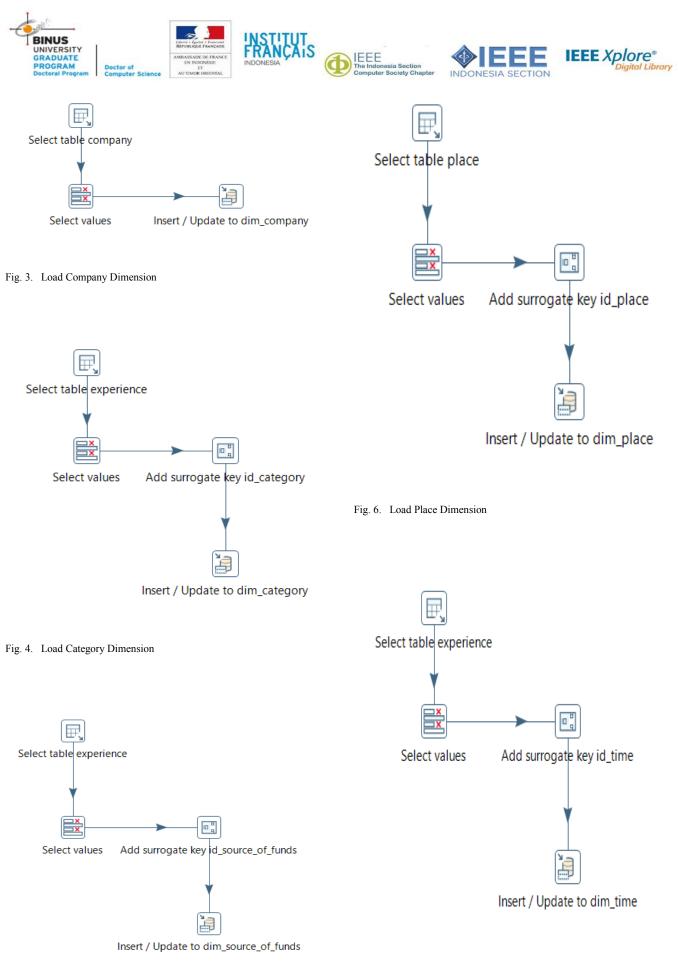


Fig. 7. Load Time Dimension

Fig. 5. Load Source of Funds Dimension



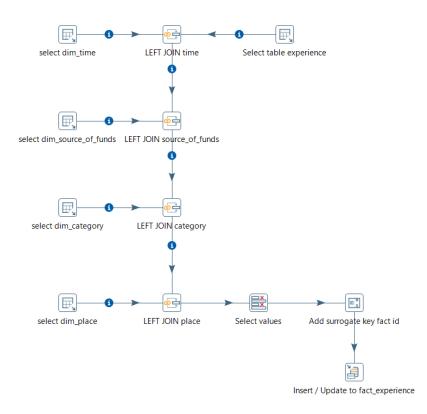


Fig. 8. Load Experience Fact Table

For fact table, each surrogate key from dimension tables will be taken for connecting the relevant data. These data relevant to surrogate key will be selected to become sample of fact table's data. Similar with dimension table, columns need to be chosen for taking core values needed in analysis. In this step, OLAP database is ready to use for dashboard and reporting. Figure 8 is an example of extraction process of experience table for construction companies' project (records of project done by each of company).

Next step is to form an OLAP Cube. the design for this case study is shown in Figure 9 that used Pentaho Workbench Schema application. Result from this OLAP Cube design is stored in form of xml file that also called as Mondrian file.

IV. DISCUSSION

We present an application of Data Warehouse to generate reports and dashboards to support decision making for construction company association management. The application intended to change manual and difficult way to generate report into more orderly way. This method used ETL process to manipulate data from operational database into star schema that easier to analyzed. The data will be represented with reports and dashboards to help management to review and take decision.

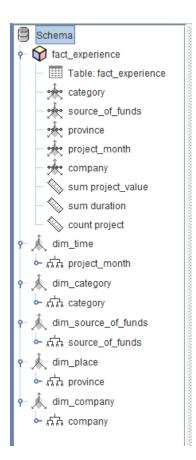


Fig. 9. OLAP Cube



V. RESULT

By using reporting and dashboard application, data from OLAP database can be processed into diagram or table that easy to understand. Reports are easy to understand by using diagrams that represent summary of data based on conditions given. If details of data are needed to be analyzed, dashboard will give more complete visualization.

In this case study, analysis will be done to find out summary of data with conditions:

- In what month construction projects have been done in a year.
- Accumulation of project execution duration for each construction companies.

- Accumulation of contract's value from contracts done by a construction company.
- Sum of projects based on source of funds for each construction companies.

Each of these conditions are represented by reports and dashboards. Figure 10 and 11 represent in what month construction projects have been done in a year. Figure 12 and 13 represent accumulation of project execution duration for each construction companies. Figure 14 and 15 represent accumulation of contract's value from contracts done by a construction company. Figure 16 and 17 represent sum of projects based on source of funds for each construction companies.

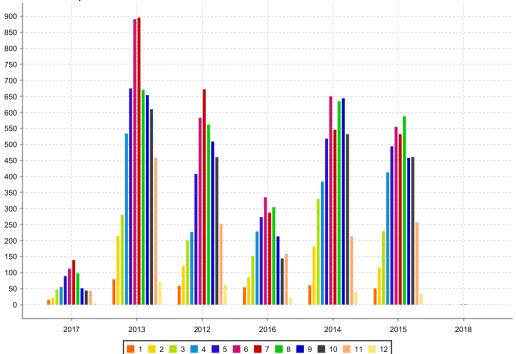


Fig. 10. Report of In What Month Construction Projects Have Been Done In a Year

1. In What Month Construction Projects Have Been Done In a Year

project_year Q project_month		Data						
2012	1	aid	Q	name	Q	project_year Q	project_month	C
				BARAKAT JUBATA		2013		
2013	2		2	BUNGA RAYA LESTARI		2014		
2014	3			BUNGA RAYA LESTARI		2014		
2015	4			BUNGA RAYA LESTARI		2014		
2015	4		5	BUNGA RAYA LESTARI		2013		
2016	5			BUNGA RAYA LESTARI		2012		
2017	6		7	BUNGA RAYA LESTARI		2012		
				LYCON ASIA MANDIRI		2015		
2018	7			LYCON ASIA MANDIRI		2015		
2021	8		10	BINTANG JAYA PRATAMA INDONESIA		2012		
				BINTANG JAYA PRATAMA INDONESIA		2014		
2022	9			BINTANG JAYA PRATAMA INDONESIA		2014		
			13	TRIMETRIC		2013		
n of projects				TRIMETRIC		2013		
4.5k			15	TRIMETRIC		2012		
			16	TRIMETRIC		2015		
			17	TRIMETRIC		2015		
3k			18	TRIMETRIC		2014		
3K			19	PERMATA JAYA		2016		
			20	PERMATA JAYA		2013		
			21	PERMATA JAYA		2016		
1.5k			22	PERMATA JAYA		2016		
			23	PERMATA JAYA		2015		
			24	UNGGUL SARANA KONSTRUKSI		2013		
			25	UNGGUL SARANA KONSTRUKSI		2012		
			26	UNGGUL SARANA KONSTRUKSI		2014		
1 2 3 4 5 6 7 8 9 10	11 12		27	UNGGUL SARANA KONSTRUKSI		2015		

Fig. 11. Dashboard of In What Month Construction Projects Have Been Done In a Year

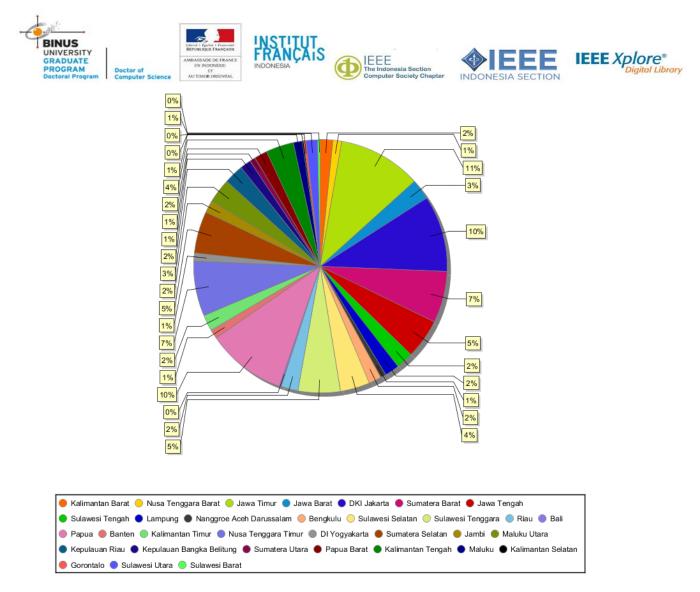


Fig. 12. Report of Accumulation of Project Execution Duration For Each Construction Companies

2. Accumulation of Project Execution Duration For Each Construction Companies

Jawa Barat

Jawa Tengah

Jambi

-		
υ	ata	

4 SAUDARA

5 PUTRI

7 BIKKU

id Q	name	Q	project_date Q	project_month Q	project_year Q	category Q	duration (days) Q	
1	BARAKAT JUBATA		26	7	2013	K2	59	
2	BUNGA RAYA LESTARI		16	5	2014	B2	178	
3	BUNGA RAYA LESTARI		17	9	2014	B2	212	
	BUNGA RAYA LESTARI		5	5	2014		207	
5	BUNGA RAYA LESTARI		16	4	2013	B2	216	
6	BUNGA RAYA LESTARI		6	2	2012		182	
7	BUNGA RAYA LESTARI		13	6	2012	B2	153	
8	LYCON ASIA MANDIRI		1	6	2015		107	
	LYCON ASIA MANDIRI		29	7	2015		113	
	BINTANG JAYA PRATAMA INDONES		30	8	2012		123	
11	BINTANG JAYA PRATAMA INDONES	IA	19	5	2014	M2	145	
	BINTANG JAYA PRATAMA INDONES	IA	19	5	2014	M2	153	
name		Q province_	name	♀ project_year	sum of	duration		
D" FAMILY		Bali			2012			50
.ITE PERKAS/	A	Banten			2013			
PUTRA ABADI	I	Bengkulu			2014			
GROUP		DI Yogyakarta			2015			
AS ENGINEE	RING	DKI Jakarta			2016			.53M
PUTRA		Gorontalo			2017		d	luration

2018

2021

2022

Fig. 13. Dashboard of Accumulation of Project Execution Duration For Each Construction Companies

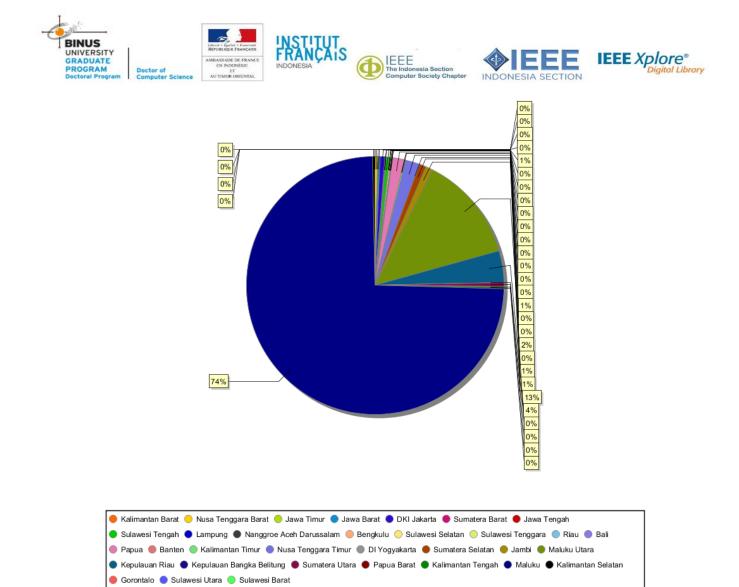


Fig. 14. Report of Accumulation of Contract's Value From Contracts Done By a Construction Company

						count project	
id (2	name	Q province_name	Q	project_value C		
	1	BARAKAT JUBATA	Kalimantan Barat		9962	50	
	2	BUNGA RAYA LESTARI	Nusa Tenggara Barat		4176410		
	3	BUNGA RAYA LESTARI	Nusa Tenggara Barat		512497		
	4	BUNGA RAYA LESTARI	Nusa Tenggara Barat		3949134		
	5	BUNGA RAYA LESTARI	Nusa Tenggara Barat		8554528		
	6	BUNGA RAYA LESTARI	Nusa Tenggara Barat		8688216	22051	
	7	BUNGA RAYA LESTARI	Nusa Tenggara Barat		1696509	projects	
	8	LYCON ASIA MANDIRI	Jawa Timur		990000		
	9	LYCON ASIA MANDIRI	Jawa Timur		467500		
1	0	BINTANG JAYA PRATAMA INDONESIA	Jawa Barat		680500	0 100	
1	1	BINTANG JAYA PRATAMA INDONESIA	Jawa Barat		4332145		
1	2	BINTANG JAYA PRATAMA INDONESIA	Jawa Barat		4332145		
1	3	TRIMETRIC	DKI Jakarta		800000		
1	4	TRIMETRIC	DKI Jakarta		450000	map	
1	.5	TRIMETRIC	DKI Jakarta		650000		province
1	6	TRIMETRIC	DKI Jakarta		859550		Area layer
1	7	TRIMETRIC	DKI Jakarta		409400		
1	.8	TRIMETRIC	DKI Jakarta		692600		
1	9	PERMATA JAYA	Sumatera Barat		22278		
2	0	PERMATA JAYA	Sumatera Barat		13440	and the second sec	
		PERMATA JAYA	Sumatera Barat		9977	A CALL AND A CALL AND A CALL	
		PERMATA JAYA	Sumatera Barat		9978		
2	13	PERMATA JAYA	Sumatera Barat		46040		
2	4	UNGGUL SARANA KONSTRUKSI	DKI Jakarta		2710		
2	25	UNGGUL SARANA KONSTRUKSI	DKI Jakarta		126250		
2	6	UNGGUL SARANA KONSTRUKSI	DKI Jakarta		80973		
2	7	UNGGUL SARANA KONSTRUKSI	DKI Jakarta		53120	1000 km) © Olik, <u>OpenStreet</u>	

3. Accumulation of Contract's Value From Contracts Done By a Construction Company

Fig. 15. Dashboard of Accumulation of Contract's Value From Contracts Done By a Construction Company

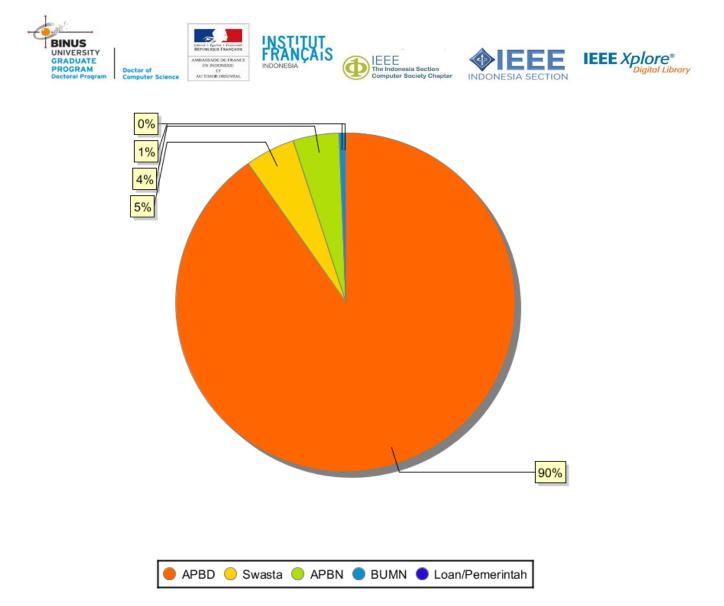


Fig. 16. Report of Sum of Projects Based On Source of Funds For Each Construction Companies

id Q	name	category Q	source_of_funds Q		
1	BARAKAT JUBATA	К2	APBD		
2	BUNGA RAYA LESTARI	B2	APBD		
3	BUNGA RAYA LESTARI	B2	Swasta		
4	BUNGA RAYA LESTARI	B2	APBD		
5	BUNGA RAYA LESTARI	B2	Swasta		
6	BUNGA RAYA LESTARI	B2	Swasta		
7	BUNGA RAYA LESTARI	B2	Swasta		
	LYCON ASIA MANDIRI	M1	Swasta		
	LYCON ASIA MANDIRI	M1	Swasta		
	BINTANG JAYA PRATAMA INDONESIA	M2	APBD		
	BINTANG JAYA PRATAMA INDONESIA	M2	APBD		
	BINTANG JAYA PRATAMA INDONESIA	M2	Swasta		
roject			Q category	\bigcirc source_of_funds	○ project_year
			B1	APBD	
			B2	APBN	
				BUMN	
			K1		
			K1 K2	Loan/Pemerintah	
				Loan/Pemerintah Swasta	
			К2		
APBD	state the second		K2 K3		
	HEAR BURNING STORES		К2 К3 М1		

4. Sum of Projects Based On Source of Funds For Each Construction Companies

Fig. 17. Dashboard of Sum of Projects Based On Source of Funds For Each Construction Companies



VI. CONCLUSIONS

By using dashboard and report, data analysis process become much easier to done. Data Warehouse is able to give capability to process fact data with multiple dimensions in OLAP database. Result from data processing can be presented with many kinds of diagrams, tables, and dashboards. Combination of reports and dashboards made a easy way for business analysts to take appropriate and quick decision to solve the problems.

We noted some limitations while doing the study case. Dashboard used in this case is not comply with real-time operational data. Qlik Sense Desktop needs to receive data from operational database source then create a data schema based on data at the time it received. This means every new data added in operational database needs to be loaded periodically in Qlik Sense Desktop, adding time costs and efforts for creating real-time reports and analysis.

Report designer used in this case can only use single query to generate a report. Query used in the report that contains join function will takes more time to process the query for generate report. There also a moment that extra queries are needed in order to create different diagrams or tables that relevant with the case in a single report.

For future works, we plan to create decision support system that helps the managers and high-level managements to take decisions from these reports and dashboards. Also, we plan to investigate different application that supports real-time operational data to create real-time business intelligence.

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