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# **Trends and Patterns of The Internet Use During School Holidays**

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Article history:	Abstract
Received 25 March 2020 Revised 4 August 2020 Accepted 18 August 2020 Available online 28 October 2020	<b>Background:</b> The Internet use according to Indonesian Internet Services Provider Association (APJII) can be an indicator for parents and educators to monitor students' mental development and learning behaviors. <b>Objective:</b> This study aims to analyze trends and patterns of the Internet use among students
Keywords:	during the school holidays. <b>Methods:</b> This study uses data from XYZ operator, one of the most affordable mobile
Data Traffic Education	service providers in Indonesia in 2019. The data was analyzed by using Online Analytical Processing (OLAP).
Holiday OLAP School The Internet	<b>Result:</b> The results shows that the use of 3G and 4G data increased significantly during the school holidays, compared to school days. The highest increase of the Internet traffic is during the semester break, occurred at the rate of 22 to 24 hours a day, with the peak reaching 20.87% at 10:00. <b>Conclusion:</b> The research findings can inform relevant parties, both parents and school
	teachers in guiding their children to use the Internet.

# I. INTRODUCTION

In the era of information technology or the Internet of Thing (IoT), business, commerce, trading, communication, and networking is now shifting from conventional to digital mode. Despite the convenience, there is also a threat, such as the impact on the younger generation. Research has shown the negative sides of the Internet use on children if not monitored and controlled [1-4]. Research shows that most teenagers spend 1 to 8 hours per day on the Internet. Some may even experience the fear of missing out (FOMO) when they are off-line. They assume that the 'addiction' is normal [3] when in fact, the Internet addiction highly influences the time management and may cause withdrawal and other behavioral problems [4]. Screen time also greatly affects eye fatigue [5] and obesity [6] at the age of children and adolescents. Aside from this, there is also risk to mental health because the Internet users are prone to cyberbullying [7] and pornography [8].

Based on survey conducted by APJII (Indonesian Association for the Internet Service Providers), in 2018 the Internet users in Indonesia reached 64.80% of the total population, raising 54.68% in 2017, and was dominated by the millennial generation (those who were born in or after 2000). Further in the APJII survey, 91% of the Internet users are aged 15-19 years; and 88.5% are aged 20-24 years. Meanwhile, the education level shows that: 41.4% elementary schools, 80.4% junior high school, 90.2% senior high school, and then 92.6% are university students [9].

Referring to BPS (Indonesian Central Bureau of Statistics) data [10], regarding the gross rate in education participant based on its level, elementary student is the biggest participant number compared to junior and senior high school student, as presented in Table 1.

Excessive use of the Internet not only affects teenagers but also adolescents (productive working age), such as by decreasing productivity [11-13]. As an illustration, the data below in Table 2 shows the comparison between the competitiveness index scores issued by the World Economic Forum [14] and INSEAD Business School [15], reading scores issued by Program for International Students Assessment (PISA) [16], as well as data on usage and speed of the Internet from several countries [17] [18].

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COMPARISON FOR EDUCATION PARTICIPANT VS THE INTERNET USER IN EACH EDUCATION LEVEL									
Level	Gross rate participant BPS 2018 (%) [10]	The Internet user APJII 2018 (%) [9]							
Elementary	38.78	41.4							
Junior High School	32.63	80.4							
Senior High School	28.58	90.2							

TABLE 1

In Indonesia, there have not been many research studies on the analysis of the Internet use behavior. If there are, they mostly use descriptive qualitative data [19] [20], which analyze the behavior without further examining the Internet data use trend. The current research aims to fill the gap in the literature. Data was collected from one of telecommunication cellular operator in Indonesia throughout 2019. Trends and patterns of the Internet use during school active days and school holidays was analyzed by using OLAP (Online Analytical Processing). OLAP method was selected because it allows easy and interactive explorative data analysis at various levels by following and applying a multidimensional approach [21-23]. The result of this research could give a clear description of the Internet use by school students so parents and teachers could provide guidance to them.

TABLE 2
DATA GLOBAL COMPETITIVENESS INDEX, READING SCORE, AND DIGITAL YEARBOOK

Country Rank	Global Competitiv	The Global Talent	Mean reading score in	DIGITAL 2019, Global Digital Yearbook Simon Kemp, We are Social Ltd										
	eness Index 4.0 2019 edition Klaus Schwab, World Economic Forum	Competitive- ness Index INSEAD, the Adecco Group, and Tata Communi- cations	PISA 2018, Organization for Economic Co-operation, and Development	Percentag e of the total popula- tion that uses the Internet	Time Per Day Spent Using the Internet via any device	Time Per Day Spent Using Mobile the Internet	Average Fixed the Internet Connec- tion Speeds	Average Mobile the Internet Connection Speeds	Time per day spent using social media	Indivi- dual Use of Social Media for Work				
	Totali		(world average: 487)	(world average: 57)	(world average: 06.42)	(world average: 03.14)	(world average: 54.3)	(world average: 25.1)	(world average: 02.16)	(world average: 24)				
				%	hour	hour	MBPS	MBPS	Hour	%				
Singapore	1	2	549	54	07.02	02.58	190.9	61.0	02.08	21				
Switzerland	5	1	484	95	04.58	01.54	104.2	47.5	01.16	20				
Japan	6	22	504	94	03.45	01.25	91.8	30.9	00.36	6				
China	12	45	555	57	05.52	03.19	89.2	29.4	01.57	24				
Malaysia	27	27	415	80	08.05	04.02	63.5	19.9	02.58	32				
Indonesia	50	67	371	54	08.36	04.35	15.5	10.5	03.26	37				
Philippines	64	58	340	71	10.02	04.58	19.0	15.1	04.12	34				

: highest in the world : lowest in the world

II. METHODS

This research is descriptive quantitative with five main steps: establish the context, collecting data, pre-processing, data processing using OLAP, and result analysis, summarise in the chart below (Fig. 1).



Fig. 1 Research Methodology

# 1. Establish the context

The data of the Internet use was collected from cellullar telecommunication operator (XYZ operator) analyzed against the academic calendar, with the scope of study covering East Java Province only. This was to avoid too many variables. Besides, the permission to use the data had been granted by the local authorities. In the 2018 APJII survey, the number of users in Java was the highest compared to other islands, which was 55.7%, then followed by Sumatra 21.6%. East Java alone ranked third with 13.5%, after West Java (16.7%) and Central Java (14.3%). With such number of Internet users, data from East Java is considered reasonable to represent the trends and patterns of the Internet use in Indonesia. The educational school calendar year are 2018/2019 and 2019/2020 starting from 1 January to 31 December 2019 (see Fig. 2). With this context, the process continues to data collection phase.

No	BULAN		TANGGAL																													
NO	DULAN	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31
	Mei'18						LU							LU							LU				PPDB	PPDB	PPDB	LU	PPDB	PPOB	PPDB	PPDB
	JUNI'18				PPDB	PPDB	PPDB	PP/DU	DU	ĸ	LU							W							LU							
1	JULI'18	LU							LU							LU	1	2	3	4	5	6	LU	7	8	9	10	11	12	LU	13	14
2	AGUSTUS,18	15	16	17	18	LU	19	20	21	22	23	24	LU	25	26	27	28	LHB	29	LU	30	31	LHB	32	33	34	LU	35	36	37	38	39
3	SEPTEMBER'18	40	LU	41	42	43	44	45	46	LU	47	LHB	48	49	50	51	LU	52	53	54	55	56	57	LU	58	59	60	KTS	KTS	KTS	LU	
4	OKTOBER'18	61	62	63	64	65	66	LU	67	68	69	70	71	72	LU	73	74	75	76	77	78	LU	79	80	81	82	83	84	LU	85	86	87
5	NOPEMBER'18	88	89	90	LU	91	92	93	94	95	96	LU	97	98	99	100	101	102	LU	103	LHB	104	105	106	107	LU	108	109	110	111	112	
6	DESEMBER'18	113	LU	114	115	116	117	118	119	LU	120	121	122	123	124	125	LU	LS1	LS1	LS1	LS1	LS1	LS1	LU	LHB	LHB	LS1	LS1	LS1	L51	LU	LS1
7	JANUARI'19	LHB	1	2	3	4	LU	5	6	7	8	9	10	LU	11	12	13	14	15	16	LU	17	18	19	20	21	22	LU	23	24	25	26
8	PEBRUARI'19	27	28	LU	29	LHB	30	31	32	33	LU	34	35	36	37	38	39	LU	40	41	42	43	44	45	LU	46	47	48	49			
9	MARET'19	50	51	LU	52	53	54	LHB	55	56	LU	57	58	59	60	61	62	ω	63	64	65	66	67	68	LU	69	70	71	72	73	74	LU
10	APRIL'19	75	76	LHB	77	78	79	LU	80	81	82	83	84	85	LU	86	87	88	89	LHB	90	LU	91	92	93	94	95	96	LU	97	98	
11	MEI'19	LHB	99	100	LPP	LU	LPP	LPP	101	102	103	104	LU	105	106	107	108	109	110	LU	111	112	113	114	115	116	LU	EF	EF	EF	LHB	LHR
12	JUNI'19	LHB	LU	LHR	LHR	LHB	LHB	LHR	LHR	LU	LHR	LHR	LHR	LHR	117	118	LU	119	120	121	122	123	124	LU	LS2	LS2	LS2	LS2	LS2	LS2	LU	
	JULI'19	LS2	LS2	LS2	LS2	LS2	LS2	LU	LS2	LS2	LS2	LS2	LS2	LS2	LU							LU							LU			

Fig. 2 East Java province, educational school calendar year 2018/2019

# 2. Data Collecting

The selected operator had a market share of 12%, and considered to be one of cheapest operator in Indonesia. The data being collected is the measurements of hourly the Internet payload traffic data for all sites' Base Transceiver Stations (BTS) in East Java.

# 3. Pre-processing

To be processed by using OLAP, data was prepared in advance by date labeling process, which is to give label status to all dates for a year by adding activity list from educational school calendar published by East Java educational authority institution. Data labeling aims to observe the trends and patterns of the Internet use between the school active days and the school holidays in accordance with the educational school calendar (see Fig. 2).

- a. Active school days
- b. Effective Facultative days (EF)
- c. Public Holidays (LHB)
- d. Religious Ceremonial Holiday (LHR)
- e. Fasting Initial Holiday (LPP)
- f. Semester Break (LS1 & LS2)
- g. Common Holiday-Sunday (LU)

#### 4. Online Analytical Processing (OLAP)

Data was processed by using OLAP, a database technology that has been optimized for querying and reporting. OLAP uses data sources from transactional database (Online Transactional Processing (OLTP) that are extracted, transformed and loaded (ETL) and stored in a data warehouse [24]. OLAP data is derived from historical data and aggregated into structures or schemes which allows sophisticated analysis. OLAP data is also organized hierarchically and stored in a cube form, and not in tables [25], [26]. It is such an advanced sophisticated technology that uses multidimensional structures to provide a quick data access for data analysis. In this research, the OLAP process design is presented in Fig. 3.

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Fig. 4 Multidimentional scheme



Fig. 5 Example OLAP operation

The database measurement is used to store the multidimensional data of all network's key performance index (KPI) measurement. The multidimensional scheme of the database measurement is presented in Fig 4. Transaction data is stored to the fine-grained level per hour. When traffic payload data is collected, it is processed by using OLAP, with the oprations: Roll up-drill down, slice-dice, agregation and pivoting as illustrated in Fig. 5.

Roll up-drill down aims to increase or decrease the level/hierarchy of summary and data aggregation. In this study, aggregation level is determined at hourly level (hourly average), then to be analyzed and compared the Internet use trends and patterns during the whole day, between school active days and holidays. Slice and dice aim to determine one and/or two dimensions of data chosen to be sub-cube being analyzed. In this case, a province or a branch is determined for an area dimension, while date, day, and hours are determined for time dimension. Aggregate operation is the process of determining the desired summary type. The aggregate chosen is *average*, since the number of days between school active days and holidays are quite imbalance. Pivot operation is applied to rotate the cube axis to

obtain other analysis viewpoints. Visualization and reporting are used to present and compare data of the Internet use patterns in accordance with the output results of OLAP process in the previous stage.

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#### 5. Result Analysis

After all data processing steps are completed, the results in the form of graphs and tables will be analyzed based on 4 (four) aspects, as follow:

a. The Internet traffic payload trends and patterns between active school days (asd) and holidays (h)

The percentage of difference/delta in this section is referred to as *delta* A as shown in (1), which compares the difference in usage between as and h based on the time period.

$$delta A = \left(\frac{h}{asd} x \ 100\right) - 100 \tag{1}$$

b. The Internet traffic payload trends and patterns between active school days (*as*) and semester break (*sb*) The percentage of difference/delta in this section is referred to as *delta B* as shown in (2), which compares the difference in usage between *as* and *h* based on the time period.

$$delta B = \left(\frac{sb}{asd} \times 100\right) - 100 \tag{2}$$

c. The Internet traffic payload trends and patterns between active school days (*as*) and initial fasting holiday (*ifh*) The percentage of difference/delta in this section is referred to as *delta* C as shown in (3), which compares the difference in usage between *as* and *ifh* based on the time period.

$$delta C = \left(\frac{ifh}{asd} \times 100\right) - 100 \tag{3}$$

d. The Internet traffic payload trends and patterns between semester break (*sb*) and non-semester break (*nsb*) The percentage of difference/delta in this section is referred to as *delta D* as shown in (4), which compares the difference in usage between *sb* and *nsb* based on the time period.

$$delta D = \left(\frac{nsb}{sb} x \ 100\right) - 100 \tag{4}$$

It is expected that the results of this study could clearly illustrate the data on the Internet use during school holidays.

# III. RESULTS

After performing roll-up or drill-down as ilustrated in Fig. 6, slice-dice as ilustrated in Fig. 8 and pivoting as ilustrated in Fig. 7, an analysis was then carried out. The following graphs and tables show the difference of trends and patterns in the Internet use presented on an hourly base. This is the fundamental difference between this study and previous studies with similar topic themes [19][20].



Fig. 6 Drill Down (a to b) - Roll UP (b to a)

iii Columns	HOUR(Time)				iii Columns	■ MC	NTH(Time)	School	Activity 1	
E Rows	MONTH(Time)	School Ac	tivity 1		⊞ Rows	HO	UR(Time)			
Payload	d by school Activity				Payload	by schoo	l Activity	/		
Month of Ti	School Activity 1	0	1	2	Hour of Time	Active scho	January Common H.,	Public Holi	Active scho	Time February Common H
January	Active school days	1,060	972	787	0	1,060	1,253	2,687	1,070	1,24
	Common Holiday-Sunday	1,253	1,170	970	1	972	1,170	2,563	964	1,1
	Public Holidays	2,687	2,563	2,054	2	787	970	2,054	776	94
February	Active school days	1,070	964	776	3	643	784	1,480		7
	Common Holiday-Sunday	1,249	1,153	949	4	733	804	1,276		71
	Public Holidays	1,087	1,018	865	5	1,231	1,263	1,712		1,2
March	Active school days	1,062	935	717	6	1,250	1,478	1,966	1,298	1,5
	Common Holiday-Sunday	1,276	1,122	912	7	1,354	1,665	2,350	1,393	1,72
	Public Holidays	1,112	964	775	8	1,454	1,786	2,651	1,477	1,79
	(a)							(b)		

Fig. 7 Pivot ((a to b) or (b to a))

#### a. The Internet traffic payload trends and patterns between active school days and holidays

School holidays are defined as all types of holidays, including public holidays, religious holidays, the beginning of fasting month holiday, semester break, and other holidays and Sundays. The results of data processing are in Table 3. The percentage figure of *traffic delta* is calculated by delta traffic of *School Holidays* and *School Active Days*, then divided by *School Active Days* traffic. Such *traffic delta* is presented in the form of a percentage to give clear description on differences in trends & patterns of data the Internet use.

Table 3 shows that the peak time of the Internet use per day in East Java both in active days and holidays is at 20:00. Meanwhile the lowest point of the Internet use is at 3:00. As illustrated in Fig. 9 (A), in general, the Internet use in school active days share similar trends and patterns to the Internet use during school holidays, only there is evidently significant increase of the Internet payload traffic during school holidays. The traffic increment per day starting at 00.00 is 14.95% and ending at 23.00 is 3.26%. While the highest increment is at 3:00 to 4:00 which is 24.24% and the lowest increment are at 5:00 to 6:00 and 19:00 to 20:00 which is -0.33%.

As shown in Fig. 3, the increment point of the Internet use is 00:00 - 04:00 and 06:00 - 16:00. We may conclude that during school holidays, during the bedtime and the daytime when parents are working, children/teenagers spend their time browsing the Internet. By observing the chart diagram, we assume that at midnight, they are still awake surfing on the Internet since there is no school rush the next morning.

(c)

ii Columns	HOUR(Time)												
Rows	■ MONTH(Time)	S	chool Activ	ity 1									
Payload	by school Activi	+											
Payload	by school Activi	LY											
							Tim	e					
Month of Ti	School Activity 1	0	1	2	3	4	5	6	7	8	9	10	11
January	Active school days	1,060	972	787	643	733	1,231	1,250	1,354	1,454	1,395	1,461	1,551
	Common Holiday-Sund	1,253	1,170	970	784	804	1,263	1,478	1,665	1,786	1,738	1,757	1,770
	Public Holidays	2,687	2,563	2,054	1,480	1,276	1,712	1,966	2,350	2,651	2,646	2,700	2,716
February	Active school days	1,070	964	776	646	717	1,237	1,298	1,393	1,477	1,415	1,477	1,569
	Common Holiday-Sund	1,249	1,153	949	768	780	1,288	1,550	1,724	1,799	1,801	1,833	1,853
	Public Holidays	1,087	1,018	865	730	802	1,377	1,634	1,820	1,896	1,806	1,812	1,806
March	Active school days	1,062	935	717	612	678	1,203	1,362	1,467	1,549	1,488	1,567	1,678
	Common Holiday-Sund	1,276	1,122	912	752	760	1,253	1,596	1,783	1,908	1,860	1,886	1,908
	Public Holidays	1,112	964	775	665	734	1,254	1,499	1,634	1,730	1,663	1,695	1,750
April	Active school days	1,160	974	776	641	683	1,206	1,435	1,565	1,659	1,599	1,677	1,724
	Common Holiday-Sund	1,314	1,141	924	738	748	1,233	1,582	1,784	1,914	1,893	1,934	1,949
	Public Holidays	1,266	1,133	922	748	762	1,300	1,682	1,877	1,981	1,954	1,972	1,983

(b)

ii Columns	HOUR(Time)									iii Columns							
Rows	MONTH(Tim	e) So	hool Activ	ity 1						⊞ Rows	MONTH(Time)	So	hool Activ	ity 1			
Payload	by school Activ	vity								Payload	by school Activi	tv					
							Ti	me			-,	-)					
Month of Ti	School Activity 1	0	1	2	3	4	5	6	7		<u></u>	-			10	Time	
January	Active school days	1,060	972	787	643	733	1,231	1,250	1,354			/	8	9	10	11	1
February	Active school days	1,070	964	776	646	717	1,237	1,298	1,393	January	Active school days	1,354	1,454	1,395	1,461	1,551	1,60
March	Active school days	1,062	935	717	612	678	1,203	1,362	1,467		Common Holiday-Sund	1,665	1,786	1,738	1,757	1,770	1,75
April	Active school days	1,160	974	776	641	683	1,206	1,435	1,565		Public Holidays	2,350	2,651	2,646	2,700	2,716	2,67
May	Active school days	1,264	1,121	983	1,370	1,504	1,237	1,268	1,420	February	Active school days	1,393	1,477	1,415	1,477	1,569	1,61
June	Active school days	1,366	1,173	916	715	722	1,222	1,520	1,723		Common Holiday-Sund	1,724	1,799	1,801	1.833	1,853	1,8
	Semester Break	1,352	1,200	945	747	732	1,235	1,601	1,798		Public Holidays	1,820	1,896	1,806	1,812	1,806	1,77
July	Active school days	1,265	1,085	849	682	724	1,300	1,451	1,547	March	Active school days	1,467	1,549	1,488	1,567	1,678	1,68
	Semester Break	1,416	1,241	984	780	762	1,238	1,601	1,798		Common Holiday-Sund.	1,783	1,908	1,860	1,886	1,908	1,88
August	Active school days	1,306	1,102	858	698	766	1,374	1,539	1,638		Public Holidays						
September	Active school days	1,342	1,117	864	700	803	1,445	1,580	1,667	April		1,634	1,730	1,663	1,695	1,750	1,7
October	Active school days	1,467	1,219	935	751	937	1,618	1,727	1,808	April	Active school days	1,565	1,659	1,599	1,677	1,724	1,75
November	Active school days	1,609	1,305	996	805	1,048	1,731	1,870	1,963		Common Holiday-Sund	1,784	1,914	1,893	1,934	1,949	1,9
December	Active school days	1,689	1,386	1,044	840	1,063	1,763	2,032	2,158		Public Holidays	1,877	1,981	1,954	1,972	1,983	1,91

# Fig. 8 Slice (a to b) – Dice (a to c)

			TABLE 3	
AVERAGE OF THE INTERNET	TRAFFIC PAYL	OAD AT SCHOOL		BREAK AND INITIAL FASTING HOLIDAY

Time		Average of The Inter	met traffic Payload (	Traffic Delta (%)							
Time	School Active Days	School Holidays	Semester Break	Initial Fasting holiday	Delta A	Delta B	Delta C				
1	2	3	4	5	6 = ((3/2) *100)-100	7 = ((4/2) *100)-100	8 = ((5/2) *100)-100				
0:00:00	1.297,94	1.491,99	1.501,58	1.133,26	14,95	15,69	-12,69				
1:00:00	1.105,47	1.308,51	1.311,87	982,52	18,37	18,67	-11,12				
2:00:00	868,24	1.062,51	1.035,90	883,08	22,38	19,31	1,71				
3:00:00	747,38	928,52	818,42	1.247,21	24,24	9,51	66,88				
4:00:00	860,09	970,00	832,58	1.283,44	12,78	-3,20	49,22				
5:00:00	1.392,23	1.387,65	1.357,98	1.200,86	-0,33	-2,46	-13,75				
6:00:00	1.528,92	1.688,54	1.749,44	1.332,67	10,44	14,42	-12,84				
7:00:00	1.636,72	1.913,82	1.967,95	1.571,69	16,93	20,24	-3,97				
8:00:00	1.738,87	2.061,04	2.096,57	1.756,19	18,53	20,57	1,00				
9:00:00	1.706,57	2.048,37	2.062,75	1.795,78	20,03	20,87	5,23				
10:00:00	1.792,28	2.097,93	2.120,78	1.884,83	17,05	18,33	5,16				
11:00:00	1.888,28	2.122,17	2.162,44	1.895,42	12,39	14,52	0,38				
12:00:00	1.936,25	2.085,32	2.155,17	1.907,30	7,70	11,31	-1,50				
13:00:00	1.882,27	2.020,00	2.082,60	1.818,91	7,32	10,64	-3,37				
14:00:00	1.871,04	1.966,35	1.987,70	1.792,02	5,09	6,24	-4,22				
15:00:00	1.843,67	1.912,96	1.912,06	1.724,30	3,76	3,71	-6,47				
16:00:00	1.893,79	1.924,93	1.938,69	1.734,35	1,64	2,37	-8,42				
17:00:00	2.029,63	2.025,34	2.087,84	1.716,94	-0,21	2,87	-15,41				
18:00:00	2.138,91	2.152,87	2.255,44	1.973,30	0,65	5,45	-7,74				
19:00:00	2.184,54	2.177,44	2.296,14	1.880,88	-0,33	5,11	-13,90				
20:00:00	2.223,62	2.220,97	2.326,50	2.028,16	-0,12	4,63	-8,79				
21:00:00	2.194,42	2.194,72	2.308,81	2.017,89	0,01	5,21	-8,04				
22:00:00	2.006,20	2.031,06	2.164,44	1.876,43	1,24	7,89	-6,47				
23:00:00	1.673,06	1.727,61	1.867,14	1.569,13	3,26	11,60	-6,21				



Fig. 9 Trends and patterns of The Internet use (A) School Active Days and all type of School Holidays, (B) School active days and semester break, (C) School active days and Initial fasting holiday, (D) all types of school holiday

#### b. The Internet traffic payload trends and patterns between active school days and semester break

In general, semester break is the longest school holiday period in Indonesia. Table 3 below describes the comparison of the Internet use between active school days and the semester break. Table 3 has similar highest and lowest peaks of the Internet use which is at 20.00 and at 03.00. A significant increase of the Internet use (above 20%) between the school active days and semester break is at 08.00 - 10.00. As shown in the data, the increase rate in the average of the Internet use is still quite high (above 10%) until 13:00, and it begin to decrease at 14.00.

Table 3 also showed that the average use of the Internet continues to increase significantly up to afternoon and evening. During the semester break, the average use of the Internet is quite high above than in school active days even until late midnight. Data visualization of Table 3 is presented in diagram (Fig. 9 (B)). It shows that for almost 22 hours during semester break, the Internet traffic payload keeps increasing significantly above the average use at school active days.

#### c. The Internet traffic payload trends and patterns between active school days and initial fasting holiday

During Ramadhan, the daily routines are changing. The data analysis is presented accordingly in Table 3, where at 00.00 there is still fewer the Internet user and yet start to increase significantly at 3:00 to 4:00 which is commonly acknowledged as salur time (early breakfast to mark the beginning of fasting). The Internet use then decreased significantly at 17.00, as it is time for breaking the fast, and at 19:00 during the Tarawih prayer time. The data visualization of Table 3 is presented in diagram (Fig. 9 (C)). It shows in general that the average the Internet use during initial fasting holiday is below of the average use at school active days.

#### d. The Internet traffic payload trends and patterns between semester break and non-semester break

Trends and patterns between all types of holidays differ from one another. As in previous section C, it is shown that during initial fasting holiday there is an anomaly, where the Internet use indeed decrease. As shown in Fig. 9 (D), there are indeed differences in trends and patterns resulting from the types of holiday. Yet, they are somewhat similat except for the initial fasting holiday. Further analysis aims to find out whether semester breaks still occupy the highest rate of the Internet use compared to other types of holiday. The results of data processing are presented in Table 4. Table 4 describe that for almost 20 hours during semester break, the average use of the Internet is above than the average use at non-semester break. The data visualization of Table 4 is presented in diagram at Fig. 10.

<b>T</b>	Average of The Intern	et traffic Payload (GB)	$T_{\rm eff} = \frac{1}{2} \left( \frac{1}{2} \right)$
Time	Non-Semester Break	Semester Break	Traffic Delta (%)
1	2	3	4 = ((3/2) * 100) - 100
00:00:00	1.44	9,54 1.501,58	3,59
01:00:00	1.27	1,74 1.311,87	2,91
02:00:00	1.05	3,78 1.035,90	-2,16
03:00:00	1.01	3,26 818,42	-19,63
04:00:00	1.06	3,95 832,58	-22,11
05:00:00	1.36	),65 1.357,98	-0,20
06:00:00	1.58	3,88 1.749,44	10,45
07:00:00	1.80	3,69 1.967,95	8,81
08:00:00	1.97	2.096,57	6,34
09:00:00	1.97	4,19 2.062,75	4,49
10:00:00	2.02	3,27 2.120,78	4,82
11:00:00	2.04	7,96 2.162,44	5,59
12:00:00	2.03	3,33 2.155,17	5,99
13:00:00	1.97	9,35 2.082,60	5,22
14:00:00	1.94	1,16 1.987,70	2,40
15:00:00	1.87	1,53 1.912,06	2,17
16:00:00	1.87	5,02 1.938,69	3,34
17:00:00	1.94	2.087,84	7,11
18:00:00	2.08	1,09 2.255,44	8,38
19:00:00	2.094	4,02 2.296,14	9,65
20:00:00	2.11	3,68 2.326,50	9,81
21:00:00	2.08	2.308,81	10,55
22:00:00	1.96	1,47 2.164,44	10,18
23:00:00	1.73	5,74 1.867,14	7,51

 TABLE 4

 Average of The Internet Traffic Payload at Semester Break and Non-Semester Break



Fig. 10 Trends and patterns of the Internet use between semester break and non-semester break

# IV. DISCUSSION

Previous research [19] has shown the duration Internet use by students was around 2-3 hours per day. Most respondent sample majorities use the Internet during the office hours on campus by utilizing the free Wi-Fi facility. Previous research [20] also shows that only 37.1% of urban teenagers (respondents of junior and senior high school students in Surabaya) used the Internet to find reading sources and to complete school work, whilst the remaining students use the Internet for fun activities (chatting, playing online game, creating a social networking account, or even visiting pornographic sites). This figure is slightly better compared to research in the city of Surakarta in 2014 that shows only 17.5% of teenagers use it for school work [27]. The current study extends the findings from previous studies by analyzing the hour spend during school holidays.

The results show that the Internet use during school holidays tend to increase significantly at certain hours compared to school active days. These findings need to be considered by parents who provide the Internet facilities to their children, as whether full supervision and accompaniment to children has already been carried out, since the Internet have both positive and negative effects. Moreover, Indonesia has the second world highest case of cyberbullying [28], so parents are expected to give more attention to their children.

#### V. CONCLUSIONS

In conclusion, nowadays the Internet is becoming a means of communication and information access that is widely used by students to spend their free time during school holidays. The research findings could be a point of reference for parents and teachers to limit and educate students in using the Internet in an orderly manner. Parents could provide extra supervision to their children while they are online. Teachers need to apply proper policy regarding the Internet use during the school hours.

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