

The Potential Customer Duration/Visit Characteristic Based on Traffic Source in a Service Provider Web Site

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Abstract: *The purpose of this research is to find out if there are any significant duration/visit difference between potential customers who come from direct, search and referral traffic sources in a service provider web site. Sample data were collected from google analytics then analyzed using descriptive statistic and ANOVA. The result shows that duration/visit from referral is the lowest and direct is the highest compare to other traffic sources.*

Keywords: *service provider, duration/visit, traffic source, search, referral, direct traffic*

Abstrak: Tujuan dari penelitian ini adalah untuk menemukan apakah terdapat perbedaan durasi/kunjungan yang signifikan antara pelanggan potensial yang berasal dari pengunjung pencarian langsung (direct), pengunjung traffic, dan pengunjung referral (rujukan) pada situs web dari penyedia layanan. Data sampel dikumpulkan dari google analytic yang kemudian dianalisis dengan menggunakan statistik deskriptif dan ANOVA. Hasilnya menunjukkan bahwa durasi/kunjungan yang berasal dari sumber pencarian referral menunjukkan nilai yang terendah; sumber pencarian langsung menunjukkan yang tertinggi dibandingkan dengan sumber pencarian traffic.

Kata Kunci: penyedia layanan, pencarian langsung, pengunjung

E-commerce is business activity conducted through electronic means (Timmers, 2000) and in particular using the Internet network communication (Turban, King and Chung, 1999; Laudon, 2012). The most popular internet service to run e-commerce is web site.

Visitors in an e-commerce web site that provide service are very important. Those visitors are potential customer that can be converted to fulfill web site's goal.

Web site owner utilize many marketing promotion programs to attract visitors to their web site. These visitors are potential customers that should be monitored and optimized. Chaffey (2009:708) suggested that, monitoring visitors interaction with web site such as their duration is very important.

Moreover, Chaffey also mentioned that web site owner should monitor whether interaction leads to satisfactory marketing outcomes such as new leads or sales.

Visitor engagement is an important factor for a web site to maximize revenue especially for online publisher (Chaffey, 2009:83). Visitor engagement can be indicated by their stickiness with a web site.

Chaffey (2009:708) define stickiness as "*an indication of how long a visitor stays on-site*". Chaffey (2009:712) also mentioned that duration and page views can be used to measure stickiness.

Some study shows that traffic source has influence stickiness. Potential customers that come from various traffic sources have different stickiness characteristic. This gives an opportunity for web site owner to measure which traffic source is more effective to gain better stickiness.

Prasetyo, et al. (2013) in an operating content provider web site, found that there is no significant

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difference between direct and referral visitors average Page views/visit. Content provider web site object in this research use advertisement as it's revenue source.

Prasetio (2013) showed different result for different web site's goal, so it's still important to find out stickiness characteristic based on traffic source especially in an operating service provider web site whose goal is to convert potential customer into sales. Laudon (2009:506) explained that stickiness or duration is important to marketers because the longer the amount of time a visitor spends at a web site, the greater the probability of a purchase.

The object of this study is kursus-online.com, a service provider website. The service provided by this website is tutorial related to the learning how to create a website. This web site uses subscription revenue model as it's revenue source.

This research will try to find out if there are any differences between duration/visit based on the traffic sources. The traffic sources measured are direct, referral, and search.

MODEL, ANALYSIS, DESIGN AND IMPLEMENTATION

This is a quantitative comparative research to find if there are any differences between independent group. The period of observation is between January 2008 and December 2012.

The research used the one way analysis of variance (ANOVA) to compare the duration/visit from 3 group of traffic source data: Direct, referral and search. The hypothesis for this research is as follows:

- H₀ : There are no differences between duration/visit mean based on different traffic sources (*direct, referral, search visitors*) ($\mu_1 = \mu_2 = \mu_3$)
- H₁ : There are differences between duration/visit mean based on the traffic sources (*direct, referral, search visitors*) (one of the μ is not equal)

Duration/visit data retrieved from website statistics called Google Analytics. Google Analytics is the most widely used tool to measure the performance of the website. Statistics from BuiltWith (2013) showed that in May 2013, Google Analytics

used by 61.5% of top 100,000 web sites which use the web analytics.

Google Analytics has grouped data into 3 main category: direct, search and referral traffic source. This data group has met our research requirements.

The data used in this study is the monthly based data and using the saturated samples. This monthly duration/visit data is an average data calculated from all visitors that come in a particular month. We gather 60 datas for each group (traffic sources), that makes 180 datas in total.

THE RESULT

The visitor's profile for the kursus-online.com between January 2008 and December 2012 is presented in Table 3.1.

Table 3.1. Visitors of the www.kursus-online.com

Traffic Source	Total Visitors from Jan 2008 - Dec 2012
Search	68.068
Referral	541.481
Direct	86.971

The above data shows the dominance of the traffic come from referral visitors, followed consecutively by direct and search visitor. This is normal because kursus-online.com utilize affiliate systems to get visitors as main marketing online tool.

The descriptive data for this study presented in Table 3.2.

The descriptive analysis from that table can be described as follows:

- The average duration/visit for referral traffic source visitors has an average of 136.92 seconds/visit, the lowest compared to other traffic sources. With the 95% confidence level or 5% significance. Average duration/visit is at the range of 128.88 seconds/visit to 146.95 seconds/visit. Duration/visit minimum is 62 seconds/visit and the maximum is 248 seconds/visit.
- The direct visitors has the highest average of duration/visit with 367.35 seconds/visit. With the 95% confidence level or 5% significance, average duration/visit for this traffic source is at the range of 340.95 seconds/visit to 393.75 seconds/

visit. Duration/visit minimum is 166 seconds/visit and the maximum is 687 seconds/visit.

- The search visitors has the average of 273.07 seconds/visit. With the 95% confidence level or 5% significance, average duration/visit is at the range of 256.11 seconds/visit to 290.02 seconds/visit. Duration/visit minimum is 110 seconds/visit and the maximum is 417 seconds/visit.

The result shows that:

- the average duration/visit of Search traffic source is defferent from Direct anda Referral traffic source
- the average duration/visit of Direct traffic source is defferent from Search anda Referral traffic source

Table 3.2. Duration/visit Statistic

Descriptives								
Duration	N	Mean	Std. Deviation	Std. Error	95% Confidence Interval for			
					Mean		Minimum	Maximum
					Lower Bound	Upper Bound		
Search	60	273,07	65,624	8,472	256,11	290,02	110	417
Direct	60	367,35	102,196	13,193	340,95	393,75	166	687
Referral	60	136,92	38,845	5,015	126,88	146,95	62	248
Total	180	259,11	119,819	8,931	241,49	276,73	62	687

The samples didn't meet the homogeneity of variance assumptions of ANOVA, but since the data have equal data sizes in each group, we still can use ANOVA in this procedure (Levine:2006).

Table 4.3 Homogeneity Varians Test Result

Test of Homogeneity of Variances			
Duration			
Levene Statistic	df1	df2	Sig.
19,441	2	177	,000

Furthermore we present the result of the ANOVA calculation. The first result is concern about the partitioning of variations.

The above calculation shows the calculated F value is 148.576. Next, we tested the level of significance of the calculated F then tested compare with the F table . Using 5% alpha, numerator value of 2 and denominator value 177 we have the F tabel value of 3.05. Referring to the provision that if calculated $F > F$ table means that H_0 is rejected, that is the three population tested are different.

- the average duration/visit of Referral traffic source is defferent from Direct anda Search traffic source

The result of this research shows that potential customers or visitors that come from direct traffic source has the highest duration/visit average, followed by search and referral as the lowest. This result consistent with Omidvar's findings that direct traffic source has higher impact on stickiness compared to referral visitors (Omidvar, 2011).

Google define direct traffic as traffic that does not originate from search-engine results or a referring link in a domain for example visitors who use a bookmark or type a web site's URL into his or her browser.

Google's explanation implies that direct visitors are returning visitors who already visited the web site before. Wang (2010) showed that returning visitors has significantly more interactions toward a web site. More over, Wang explained that returning visitors are likely to have more "loyalty" to this site, since returning visitors are very likely purposeful searchers.

Multiple Comparisons							
Dependent Variable:Duration							
			Mean			95% Confidence Interval	
	(I) Traffic Source	(J) Traffic Source	Difference (I - J)	Std. Error	Sig.	Lower Bound	Upper Bound
Tukey HSD	Search	Direct	-94,283*	13,441	,000	-126,05	-62,51
		Referral	136,150*	13,441	,000	104,38	167,92
	Direct	Search	94,283*	13,441	,000	62,51	126,05
		Referral	230,433*	13,441	,000	198,66	262,20
	Referral	Search	-136,150*	13,441	,000	-167,92	-104,38
		Direct	-230,433*	13,441	,000	-262,20	-198,66
LSD	Search	Direct	-94,283*	13,441	,000	-120,81	-67,76
		Referral	136,150*	13,441	,000	109,62	162,68
	Direct	Search	94,283*	13,441	,000	67,76	120,81
		Referral	230,433*	13,441	,000	203,91	256,96
	Referral	Search	-136,150*	13,441	,000	-162,68	-109,62
		Direct	-230,433*	13,441	,000	-256,96	-203,91

*. The mean difference is significant at the 0.05 level.

Wang also mentioned that returning visitor’s high engagement with a web site are probably caused by their interests or information needs.

The nature of subscription revenue model customer also has impact to the length of visit. Customer who already subscript to the web site will also become a returning visitor. Due to their interest to learn on the web site, they are more likely to stay longer during their lerning session.

Referral and search visitors are usually new visitors. This type of visitor usually don’t stay long on the site and did not explore much of the contents on the site (Wang, 2010).

The result shows that referral visitors have the lowest duration/visit compared to other traffic sources. This result indicates that the referring web pages doesn’t have relevant topic with the product.

In order to understand search visitors behavior, Omidvar (2011) offers that their quarries meaning should be understood which is available through semantic web technologies. This analysis doesn’t included in this research.

CONCLUSION

Conclusion

The result shows that at kursus-online.com, a service provider web site, direct traffic source has the highest average duration/visit, followed by search and referral as the lowest.

This research support Omidvar’s findings that in a web site that sell product, direct visitors has better stickiness compared to other traffic sources. This result also indicates that direct visitors provide better potential customers for a service provider web site.

Recommendation

Further research can be study the other website using the same method or other factors to identify interactions between visitors with a web site to understand how to get better potential customer. The research can focus on other interaction characteristic such as connection speed and the types of visitors (Omidvar, 2011).

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