Business Case: Swiggy

Introduction

Swiggy is one of the largest food ecommerce platforms in the country. Every day more than 1 million users are transacting on the platform. Let's say you are growth and strategy analyst of Swiggy, and you need to generate insight on company's performance in 2019. For this, you are going to use 'Funnel Case Study Data' workbook which has 3 work sheets. You can find the details of the sheets below.

Q.1 - Identify date of highs and lows in the orders with respect to same day last week.

Ans – In below table I shown fluctuation which is more than 20% and less than -20%. As this tolerance rang given in question.

Refer Excel - Ans Sheet 1

Date	Listing	Overall conversion	Order Change with respect to same day last week
10-01-2019	10641496	6%	
17-01-2019	22368860	6%	106%
21-01-2019	22151687	7%	23%
22-01-2019	37570998	6%	85%
29-01-2019	22368860	3%	-72%
31-01-2019	20848646	6%	
05-02-2019	22368860	6%	115%
19-02-2019	21934513	3%	-56%
26-02-2019	22368860	6%	120%
28-02-2019	22586034	6%	
02-03-2019	46685340	2%	-38%
09-03-2019	46685340	4%	
19-03-2019	21934513	3%	
24-03-2019	45338648	4%	
26-03-2019	20848646	6%	
04-04-2019	22151687	3%	-52%
11-04-2019	20631473	6%	92%
12-04-2019	20631473	6%	-27%
14-04-2019	46685340	4%	28%
18-04-2019	22803207	9%	
19-04-2019	22151687	6%	25%
25-04-2019	22803207	6%	-39%
20-06-2019	10207150	6%	-54%
27-06-2019	22368860	6%	
16-07-2019	20631473	2%	-63%
23-07-2019	21282993	6%	135%
11-08-2019	43991955	2%	-54%
18-08-2019	45338648	3%	107%
14-09-2019	44440853	2%	-54%
21-09-2019	43991955	3%	112%
09-10-2019	20631473	7%	22%
21-10-2019	22803207	6%	32%
09-11-2019	45787545	4%	
17-11-2019	43991955	2%	
24-11-2019	46236443	4%	
01-12-2019	46685340	4%	
22-12-2019	43094160	4%	21%

Highest order change date -

Highest order	
change	Highest order Date
135%	24-11-2019
135%	23-07-2019

Lowest order change date -

Lowest order change	Lowest order Date
-72%	29-01-2019

Q.2 - Check if there is change in traffic as compared to same day last week, if there is change in traffic, identify the source of traffic change using Channel wise traffic sheet.

Ans –

In below table it shows change in traffic as compared to same day last week.

Date	Day	Listing	Change in traffic	Reason
				Facebook by -95%,
10-01-2019	Thursday	10641496	-49%	youtube & twitter by -50%
				Recoverd by last week
				same day traffic loss in
				facebook, youtube,
17-01-2019	Thursday	22368860	110%	twitter
				Twitter by 628% &
22-01-2019	Tuesday	37570998	77%	facebook by 70%
				Fluctuation come to
				average rang from last
29-01-2019	Tuesday	22368860	-40%	week same day
				From overall source traffic
20-06-2019	Thursday	10207150	-53%	decrease by -50%
				Recovered all traffic
				which we loss on last
27-06-2019	Thursday	22368860	119%	week same day

- By analysing the above table, it can be observed that if there is a change in traffic percentage (positive or negative) on a specific date, a opposite change is also reflected in the traffic percentage on the same day of the following week.
- Why next week same day is inversely proportional to current day?

 So, the reason could be if on current date traffic is decrease by 50% compared to last week same day, it means due to some issue half traffic didn't came to platform.
- Till next week same day, Issue got solved Platform got there traffic again which they loss in last week.
- So, in next week same day traffic change will be 100%. And that's why current day is inversely proportional to next week same day.

Change in traffic from each source (Facebook, YouTube, Twitter, Other)

Refer excel sheet - Ans Sheet 1



10-01-2019	Thursday	10641496	-49%
17-01-2019	Thursday	22368860	110%

Date - 10-01-2019.

	Avg. faceboo	Avg. youtube	Avg. twitter	Avg. Other
Average on Thu	7546035	5709196	2325969	5563971
Same day	387156	2873204	1170564	6210572
Change in				
percentage	-95%	-50%	-50%	12%

As per above table -

On an average traffic on platform, there is good decrease in Facebook, YouTube, Twitter compared to average traffic.

- Facebook = Decrease by 95%
- YouTube = Decrease by 50%
- Twitter = Decrease by 50%
- Other = Increase by 12%

Date - 17-01-2019.

	Avg. facebool	Avg. youtube	Avg. twitter	Avg. Other
Average on Thu	7546035	5709196	2325969	5563971
Same day	8052789	6039592	2460574	5815903
Change in				
percentage	7%	6%	6%	5%

As per above table -

- There is no such fluctuation in any of the source but in traffic change it showing 110% increase.
- It is because whatever traffic Facebook, YouTube, Twitter loss on 10-01-2019. They gain all that traffic on 17-01-2019.
- Therefore, its showing 110% increase.

B.

22-01-2019	Tuesday	37570998	77%
29-01-2019	Tuesday	22368860	-40%

Date - 22-01-2019.

	Avg. faceboo	Avg. youtube	Avg. twitter	Avg. Other
Average on Tue	7946579	5806815	2724245	5596192
Same day	13525559	2028833	19827367	2189238
Change in				
percentage	70%	-65%	628%	-61%

As per above table -

- Twitter showing hug increase in traffic 628% as compared to average.
- Facebook increase by 70%.
- YouTube & Other decrease by -65%, -61%.

Date - 29-01-2019.

	Avg. facebool	Avg. youtube	Avg. twitter	Avg. Other
Average on Tue	7946579	5806815	2724245	5596192
Same day	8052789	6039592	2460574	5815903
Change in				
percentage	1%	4%	-10%	4%

As per above table -

- There is no such fluctuation in any of the source but in traffic change it showing -40% Decrease.
- Because on 22-01-2019 due to some external factor Sources gain 77% compared to average traffic
- Now current day overall traffic come towards average traffic.
- Therefore, its showing -40% Decrease.

C.

20-06-2019	Thursday	10207150	-53%
27-06-2019	Thursday	22368860	119%

Date - 20-06-2019.

	Avg. faceboo	Avg. youtube	Avg. twitter	Avg. Other
Average on Thu	7546035	5709196	2325969	5563971
Same day	3674574	2755930	1122786	2653859
Change in				
percentage	-51%	-52%	-52%	-52%

As per above table -

• All sources are decrease by -52%

Date - 27-06-2019.

	Avg. faceboo	Avg. youtube	Avg. twitter	Avg. Other
Average on Thu	7546035	5709196	2325969	5563971
Same day	8052789	6039592	2460574	5815903
Change in				
percentage	7%	6%	6%	5%

As per above table -

- There is no such fluctuation in any of the source but in traffic change it showing 119% Increase.
- Because on 20-06-2019 due to some reason it decreases by -53% as compared to average.
- On current date (27-06-2019) its recovered and come to average area.

Q.3 - Identify which one of the conversions is fluctuating, create hypotheses on what could be the possibility for fluctuation in conversions, Validate the hypotheses using Supporting data.

Ans –

Refer excel sheet – Ans Sheet 2 (for better understanding)

Below table content conversion which has high fluctuation													
Date	Days	Conversion change %	L2M	M2C	C2P	P20	Reason						
							Decrease in count of						
29-01-2019	Tuesday	-52%	12%	42%	72%	80%	restaurant by -30%						
							Recovered by last week						
05-02-2019	Tuesday	115%	26%	40%	71%	80%	same day						
							Minor change in packagin						
							and image and also same						
19-02-2019	Tuesday	-54%	26%	17%	77%	85%	external reason						
							Recovered by last week						
26-02-2019	Tuesday	116%	24%	41%	74%	81%	same day						
							Increase in Delivery charg						
02-03-2019	Saturday	-42%	21%	34%	33%	81%	by 100%						
							Recovered by last week						
09-03-2019	Saturday	102%	21%	34%	71%	79%	same day						
							Success rate of payment is						
19-03-2019	Tuesday	-47%	26%	42%	76%	39%	decrease by 30%						
							Recovered by last week						
26-03-2019	Tuesday	87%	24%	40%	72%	85%	same day						
04-04-2019	Thursday	-53%	26%	20%	69%	78%	Discount decrease by -45						
							Recovered by last week						
11-04-2019	Thursday	107%	25%	39%	76%	80%	same day						
18-04-2019	Thursday	57%	24%	67%	73%	79%	Discount increse by 60%						
							Discount came to normal						
25-04-2019	Thursday	-39%	25%	38%	69%	84%	rang campared to last we						
	,						Avg cost for two increased						
16-07-2019	Tuesday	-59%	10%	40%	73%	84%	by 23%						
	,						Recovered by last week						
23-07-2019	Tuesday	128%	24%	40%	75%	78%	same day						
	,						Packaging charge increase						
11-08-2019	Sunday	-54%	22%	33%	33%	74%	by 45%						
	,					-	Recovered by last week						
18-08-2019	Sunday	100%	21%	33%	65%	78%	same day						
	,						Out of stock iteam						
14-09-2019	Saturday	-51%	21%	15%	67%	74%	increased by 80%						
	, , , , , , , , , , , , , , , , , , , ,	3270	22/0	25%	3770	. 170	Recovered by last week						
21-09-2019	Saturday	114%	20%	34%	65%	75%	same day						
		11-170	20/0	3470	3370	. 370	Out of stock iteam						
17-11-2019	Sunday	-54%	21%	14%	71%	77%	increased by 220%						
2, 11 2010		3470	21/0	1170	7170	. , , , ,	Recovered by last week						
24-11-2019	Sunday	124%	21%	34%	66%	76%	same day						

This above table has Positive and Negative conversion rate, and funnel break into 4 parts which is L2M, M2C, C2P, P2O.

- First, I will see conversion change %, then go to funnel break down to identify in which part of funnel the issue occurred.
- After that I will see what Is the reason behind, with the help of supporting data.

Let's have a look in each date and identify the reason, at a time I will take two dates to compared.

		i		<u>i </u>		
Date	Days	Conversion change %	L2M	/I2C	C2P	P20
29-01-2019	Tuesday	-52%	12%	42%	72%	80%
· ·	Average Conversation			38%	71%	81%
05-02-2019	Tuesday	115%	26%	40%	71%	80%
				i		

- On 29-01-2019, By analysing above table, we get to know the issue occurs in L2M funnel.
- On 05-02-2019, this date showing 115% conversion change, but it's recovered from last week same day fluctuation.

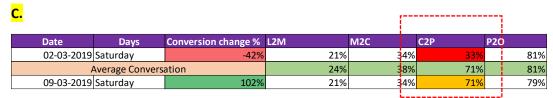


- Orange line and blue line is tolerance line +-20%.
- The reason behind change in L2M, as you see the above chart we get to know count of restaurant crossed orange tolerance line.
- Due to decrease in count of restaurants L2M decrease till 12%.

							i		
Date	Days	Conversion change %	L2M		M2C	C2F	,	P2O	
19-02-2019	Tuesday	-54%		26%	17%		77%	85%	
A	Average Convers	ation		24%	38%		71%	81%	
26-02-2019	Tuesday	116%		24%	41%		74%	81%	



- On 19-02-2019 conversion change is -54%, as per above table its clear there is issue in M2C funnel.
- On 26-02-2019 conversion change is 116%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in M2C traffic, as per above chart all reasons are under tolerance line.
- It means there is external reason behind the drop, it may be customer got good discount, less delivery time etc on competitors' platform.





- On 02-03-2019 conversion change is -42%, As per above table its look like there is change in C2P.
- On 09-03-2019 conversion change is 102%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in C2P traffic, as per above chart "average delivery charges" crossed tolerance line.
- So, the reason is due to increase in delivery charge by 100% most of the customer leave the platform.

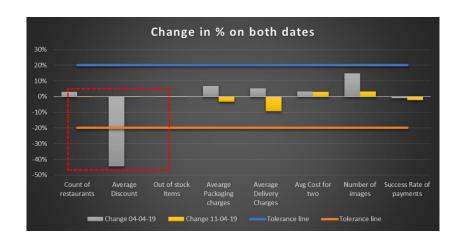
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Date	Days	Conversion change %	L2M	M2C	C2P		P20
19-03-2019	Tuesday	-47%	26%	42%	76	6%	39%
P	Average Conversation			38%	7:	1%	81%
26-03-2019	Tuesday	87%	24%	40%	72	2%	85%
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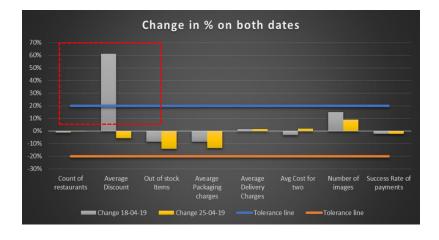
- On 19-03-2019 conversion change is -47%, As per above table its look like there is change in
- On 26-03-2019 conversion change is 87%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in P2O traffic, as per above chart "Success Rate of Payments" crossed tolerance line.
- Decrease in Success Rate of Payments by 30%, Reason behind may be Bank Server down, Issue in app while payment, Limited payment option etc.

Date	Days	Conversion change %	L2M	M2C	C2P	P2O
04-04-2019	Thursday	-53%	26%	20%	69%	78%
Ä	Average Conversation			38%	71%	81%
11-04-2019	Thursday	107%	25%	39%	76%	80%



- On 04-04-2019 conversion change is -53%, As per above table its look like there is change in M2C.
- On 11-04-2019 conversion change is 107%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in M2C traffic, as per above chart "Average Discount" crossed tolerance line.
- So, the reason behind change in M2C is "Average Discount" decrease by 45%.

F.									
	Date	Days	Conversion change %	L2M		M2C	C2P		P2O
	18-04-2019 Thursday		57%		4%	67%		73%	79%
	A	Average Convers	ation	2	4%	38%		71%	81%
	25-04-2019	Thursday	-39%	2	5%	38%		69%	84%



- On 18-04-2019 conversion change is 57%, As per above table its look like there is change in M2C.
- On 25-04-2019 conversion change is -39%, it's just come to average range from last week same day Increase.
- Reason behind change in M2C traffic, as per above chart "Average Discount" crossed tolerance line.
- So, due to increase in "Average Discount" by 60%. M2C is increase by 67%.

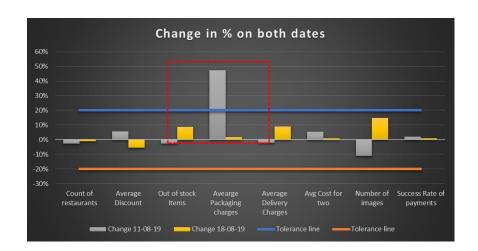
<mark>G.</mark>

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Date	Days	Conversion change %		L2M	N	12C	C2P	P20
16-07-2019	Tuesday	-59%	%	10%		40%	73%	84%
A	Average Conversation			24%		38%	71%	81%
23-07-2019	Tuesday	128	%	24%		40%	75%	78%
					ī			



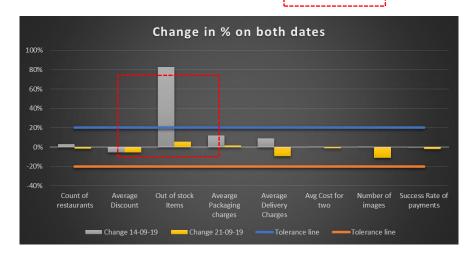
- On 16-07-2019 conversion change is -59%, As per above table its look like there is change in 12M
- On 23-07-2019 conversion change is 128%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in L2M traffic, as per above chart "Average cost for two" crossed tolerance line.
- So, due to change in "Aver cost for two" by 23% that's why there is change in conversion by 59%.

Date	Days	Conversion change %	L2M	M2C	C2P	P2 <mark>O</mark>	
11-08-2019	Sunday	-54%	22%	33%	33%	74	4%
A	Average Conversation			38%	71%	8:	1%
18-08-2019	Sunday	100%	21%	33%	65%	78	8%
			•				



- On 11-08-2019 conversion change is -54%, As per above table its look like there is change in C2P.
- On 18-08-2019 conversion change is 100%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in C2P traffic, as per above chart "Average Packaging charges" crossed tolerance line.
- So, due to change in "Average packaging charges" by 45%. C2P is decrease by 33% from average.

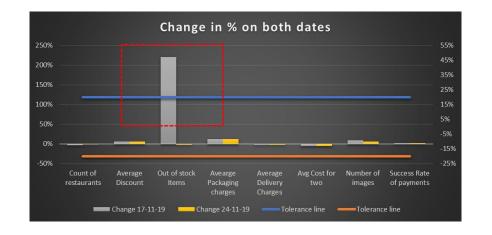
Date	Days	Conversion change %	L2M	M2C	C2 <mark>P</mark>	P2O
14-09-2019	Saturday	-51%	21%	15%	67%	74%
A	Average Conversation			38%	71%	81%
21-09-2019	Saturday	114%	20%	34%	65%	75%



- On 14-09-2019 conversion change is -51%, As per above table its look like there is change in M2C
- On 21-09-2019 conversion change is 114%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in M2C traffic, as per above chart "Out of Stock Items" crossed tolerance line.
- So, due to change in "Out of Stock Items" by 80%. M2C decrease to 15% from 38% average.

J.

Date	Days	Conversion change %	L2M		M2C	C2F		P20
17-11-2019	Sunday	-54%	21	%	14%		71%	77%
A	Average Conversation			%	38%		71%	81%
24-11-2019	Sunday	124%	21	%	34%		66%	76%



- On 17-11-2019 conversion change is -54%, As per above table its look like there is change in M2C.
- On 24-11-2019 conversion change is 124%, it's just a recovery week recovered from last week same day drop.
- Reason behind change in M2C traffic, as per above chart "Out of Stock Items" crossed tolerance line.
- So, due to change in "Out of Stock Items" by 210%. M2C decrease to 14% from 38% average.

CONCLUTION

As per above study/analysis, Reason which is affecting overall conversion ratio is **Count of restaurant**, **Average discount**, **out of stock**, **Average packaging charge**, **Average delivery charge**, **Average cost for two**, **Number of images**, **Success rate of payments**. Out of stock and Average discount are two main reasons behind decrease in overall conversion, we should more focus on these two points to sustain the overall conversion ratio.

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