



Capstone Project



ElecKart Market Mix Modeling

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Eleckart Market Mix Model



ElecKartMarket Mix Modeling Objective:

- To identify the most useful revenue streaming channels for marketing spends.
- To identify other factors having positive impact on sales of three product categories –Camera Accessory, Gaming Accessory & Home Audio Accessory.

Data Understanding:

- Data provided for a period of one year from July 2015 to June 2016.
- The available data is at an order level for all the purchases made in the said time duration.
- The monthly advertising spends on the different marketing channels is also provided.
- The NPS (or brand perception) data is also available together with the holiday list for the year.



Approach



1.Data Preparation

1. Missing Values and Outlier Treatment
2. Variable Transformation
3. Unified date format

2.Data Exploration

1. Variable Identification
2. Univariate and Bi-Variate Analysis

3.Feature Engineering

1. Feature Extraction and Selection
2. Variable Interaction
3. Feature Creation

4.Model Building

1. Basic linear Model
2. Multiplicative Model
3. Koyck Model
4. Distributed Lag Model
5. Multiplicative + Distributed Lag Model

5.Model Evaluation and Recommendation

Source Code Details

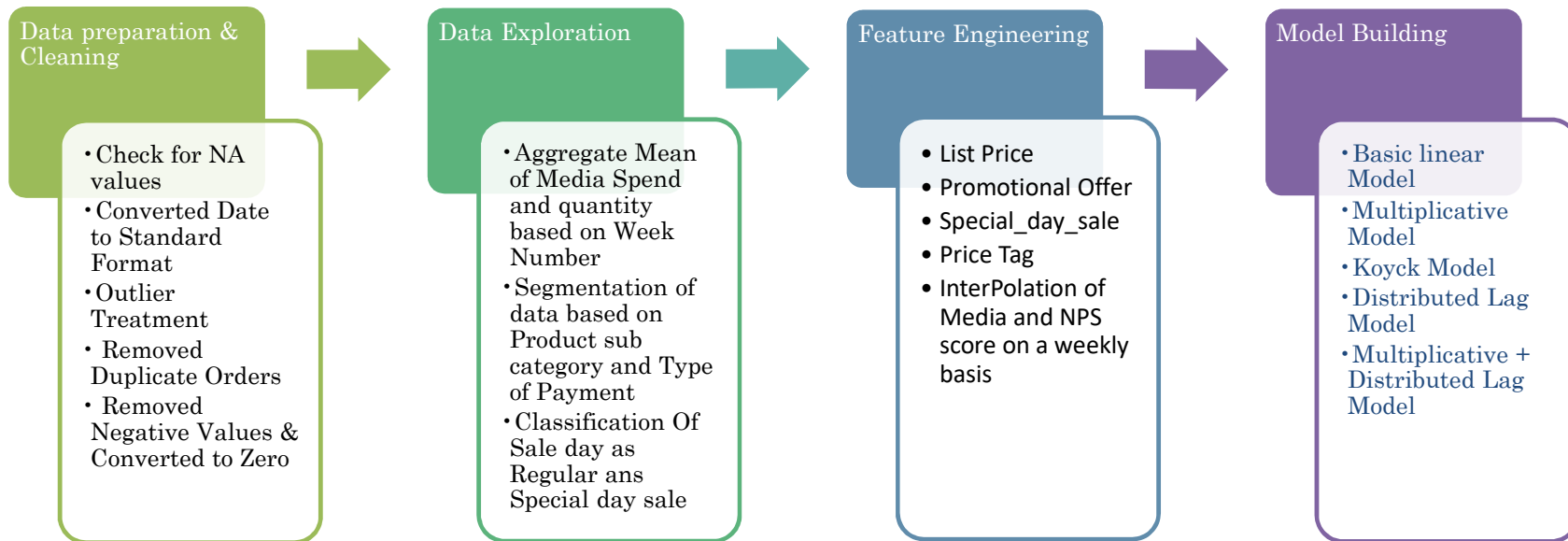
We have segregated our code into three parts.

1. Data Preparation and Model building is available under the filename "Ecommerce_Capstone_Model.R"
2. Data Preparation and EDA is available under "EDA_Capstone.R"

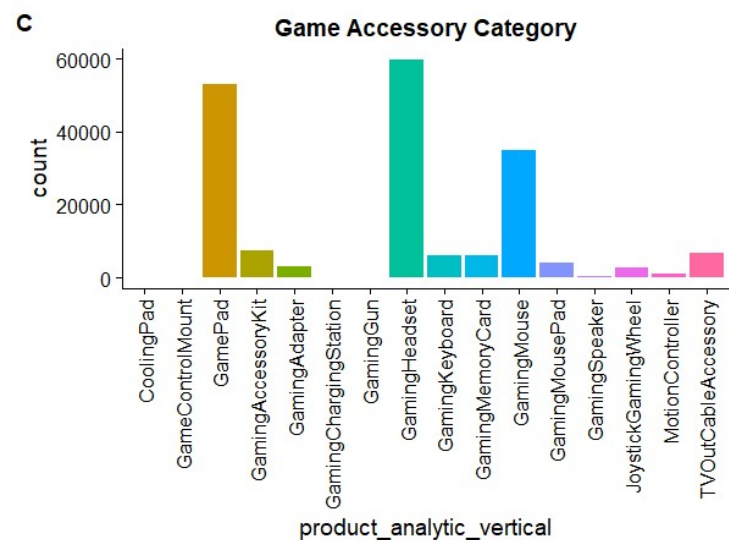
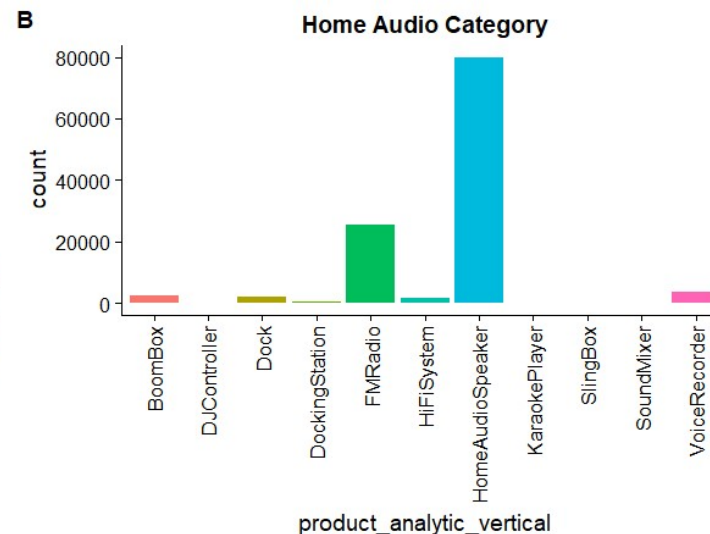
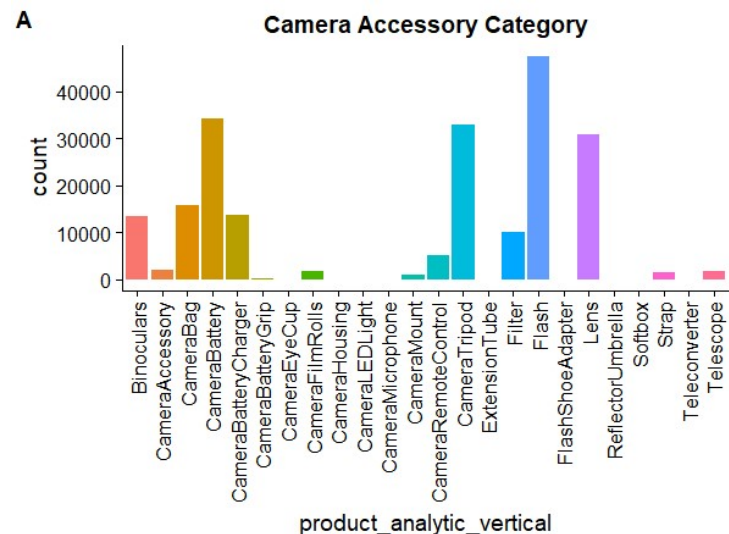
Note:

Data preparation is same under both files. Due to naming conventions used inside, we have segregated the files.

Problem Solving approach

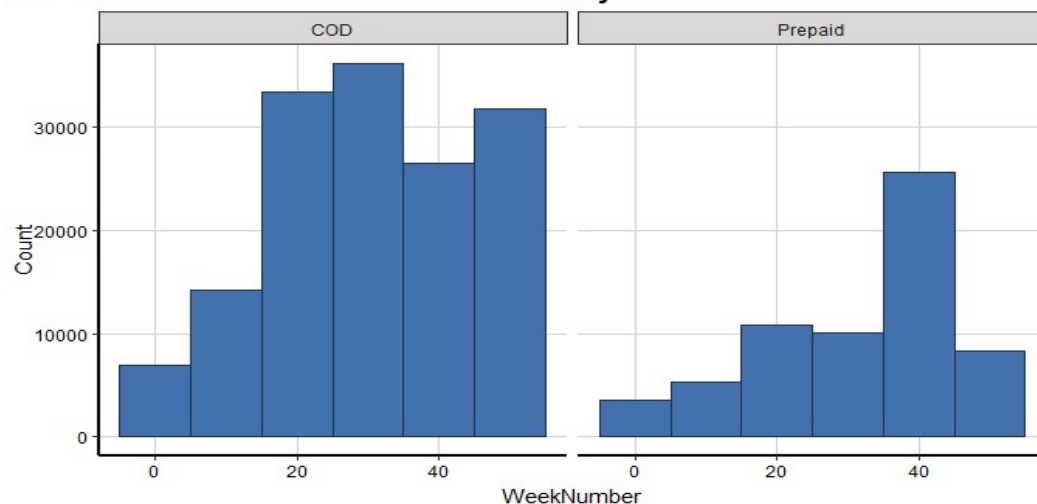


EDA for Different Product Analytic Category

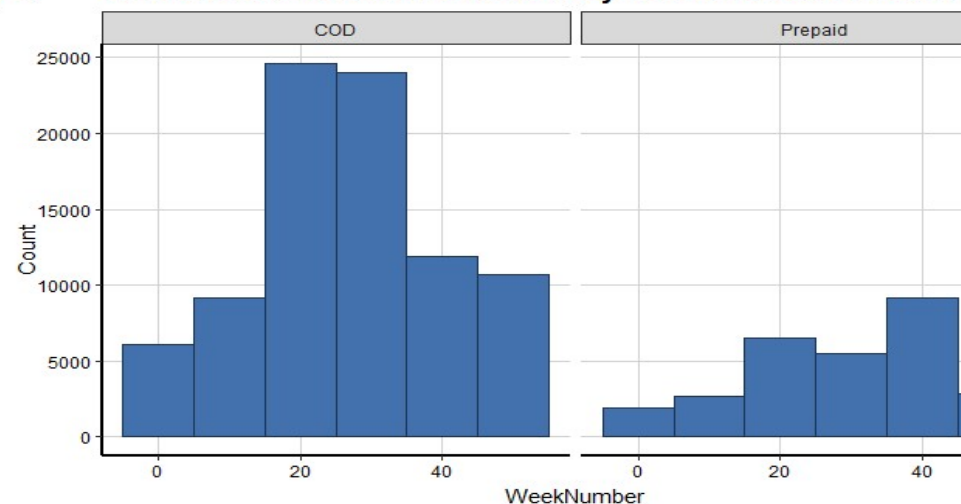


- Under Camera Accessories we found Camera Lens, Flash , Tripod are high selling categories , where as CamerLED Light, Extension tube and Camera Housing are low selling Categories.
- Under Home Audio we found Home Audi Speakers and FM Radio are high selling products,where as SingBox , Sound Mixer and DJ Controller does not have any demand.
- Under Gaming Accessories “Gaming Headset”, “GamePad” and GameMouse are high selling categories where as Cooling Pad , GameControl Mount and Gaming Charging Station got zero demand.

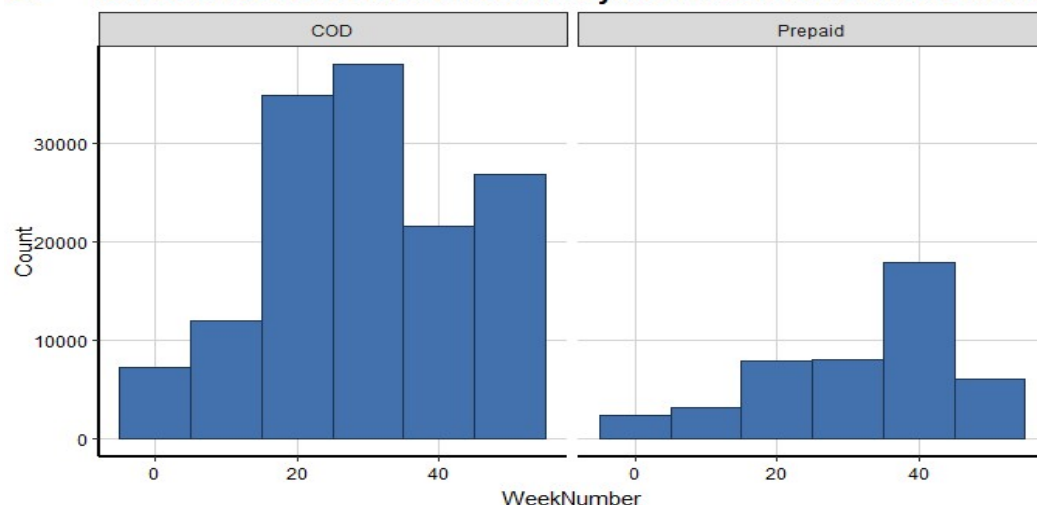
A No. of orders based on different Payment Methods for camera accessories



B No. of orders based on different Payment Methods for Home Audio

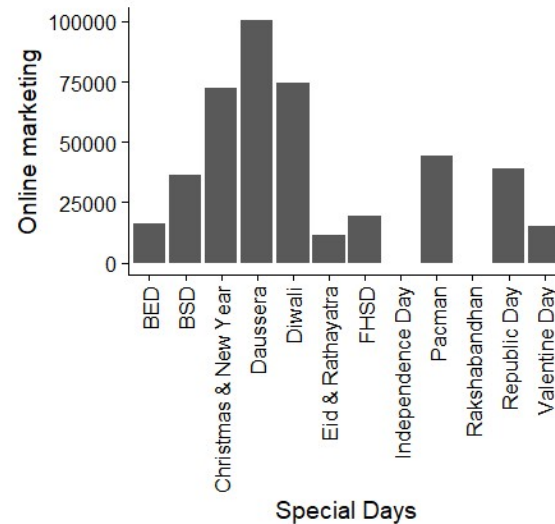
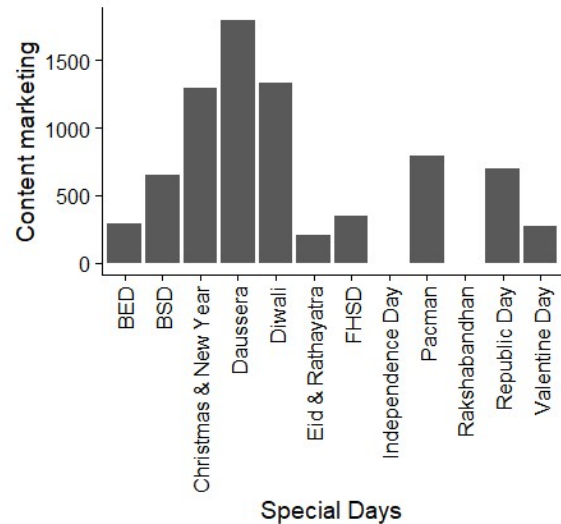
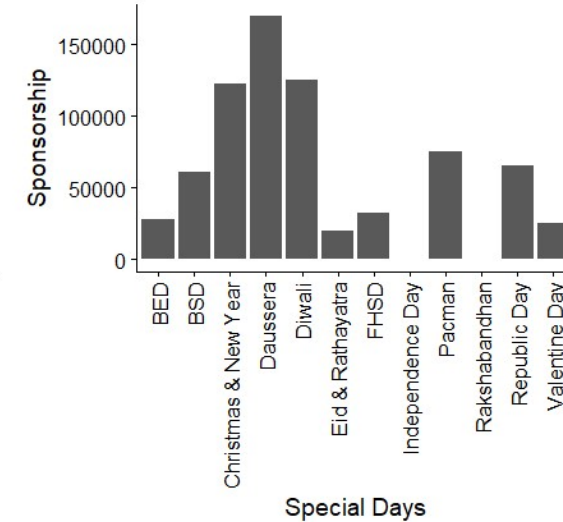
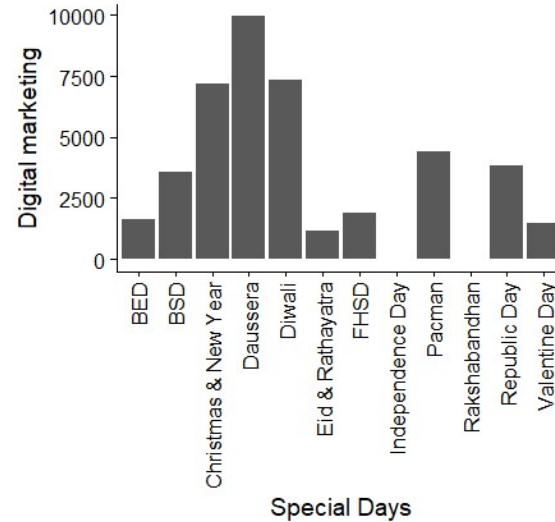
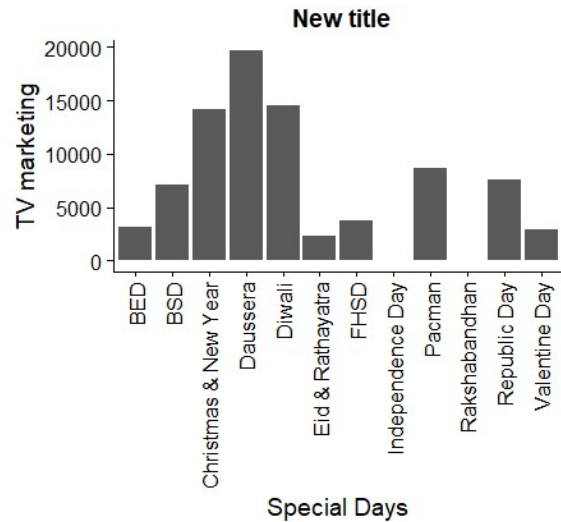


C No. of orders based on different Payment Methods for Home Audio



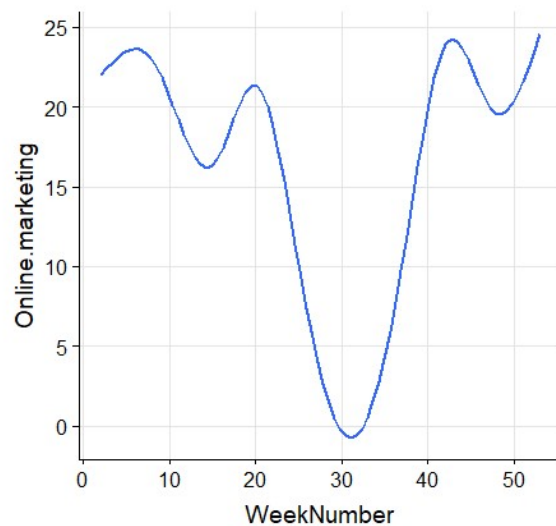
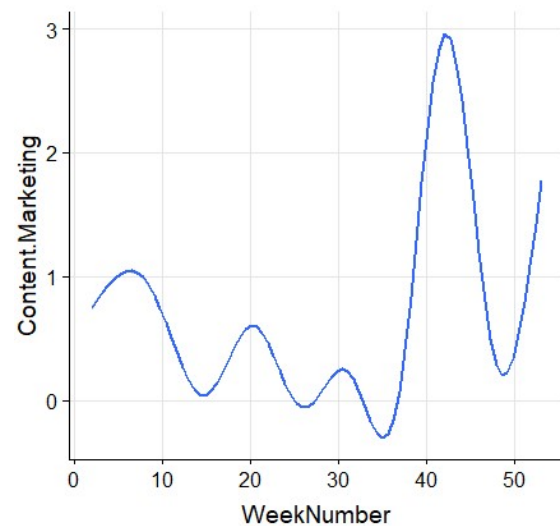
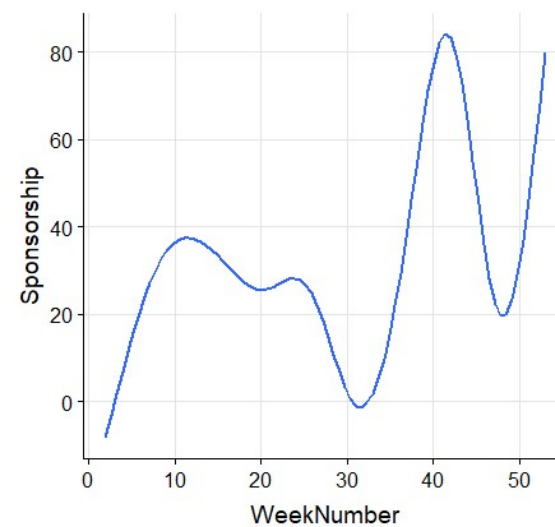
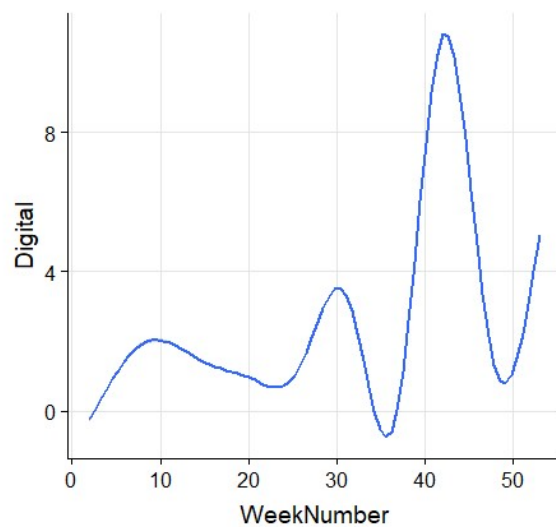
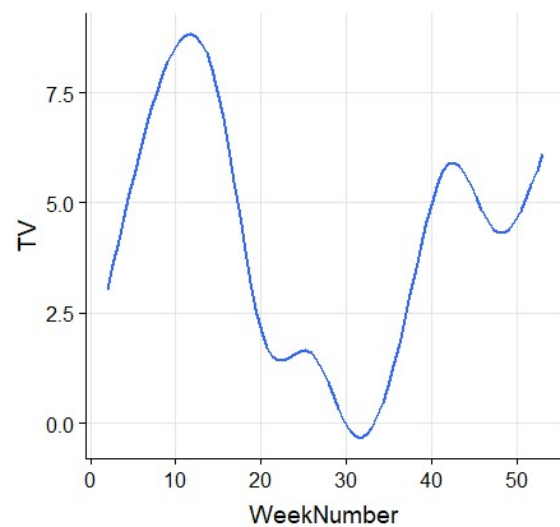
Between week 25 and 35 there is huge demand on COD Payment type across all three segments.

Between week 35 and 40 there is also relatively a small surge in prepaid (Online) Payment type

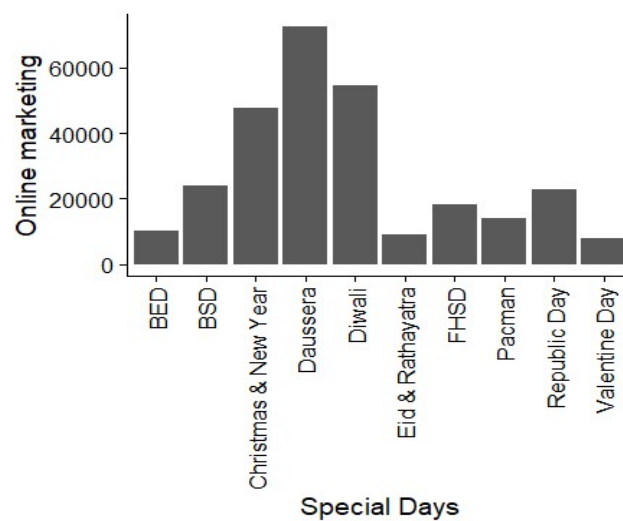
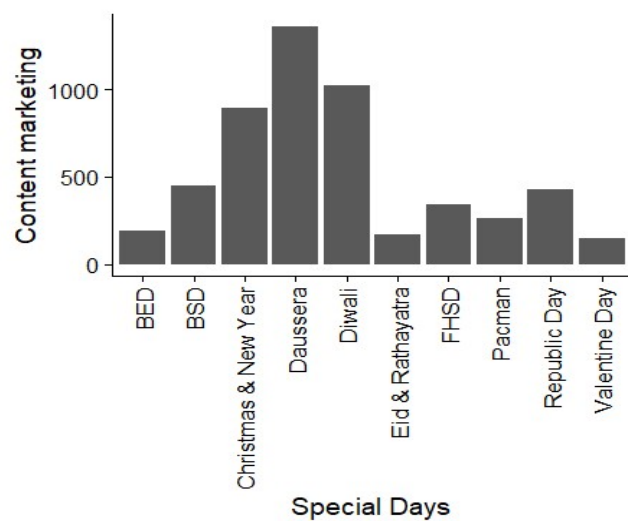
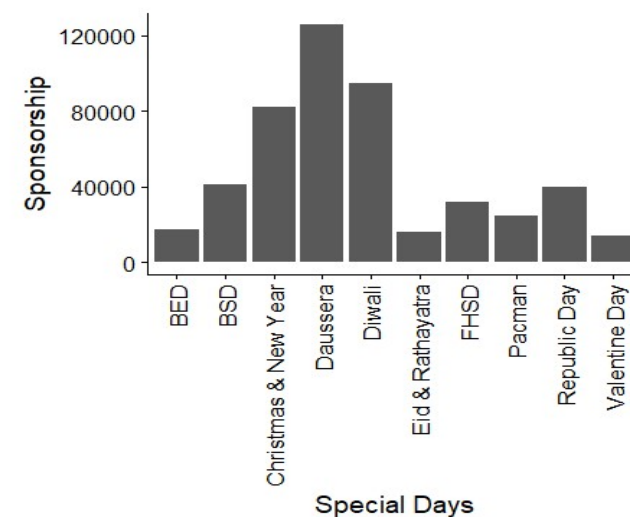
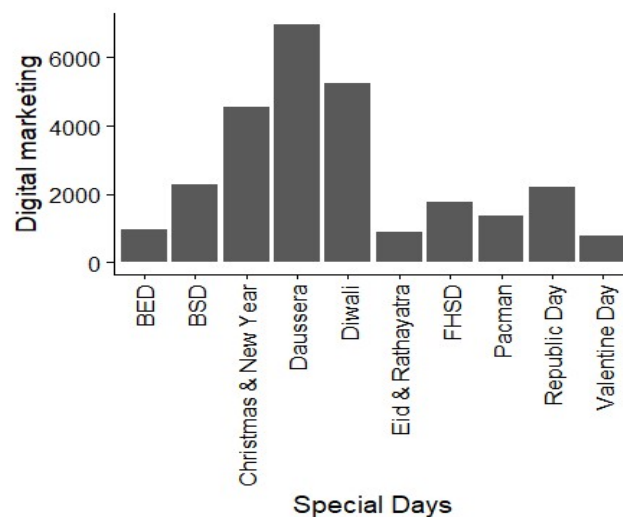
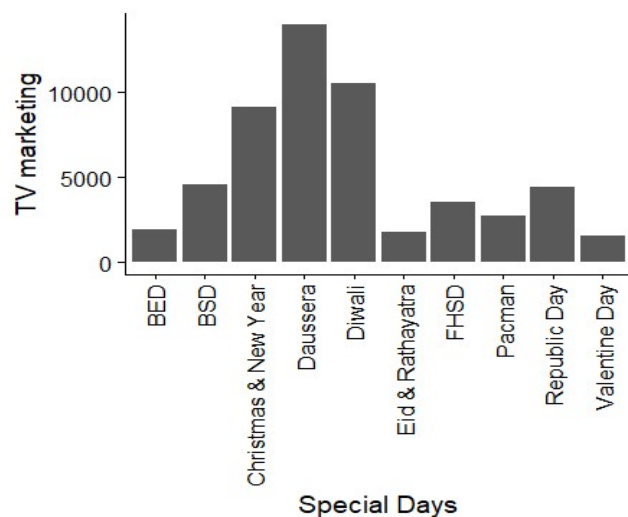


- On Independence Day and Rakshabandhan Day there is no spend on media advertisement. They should try leveraging their media spend on these days too.

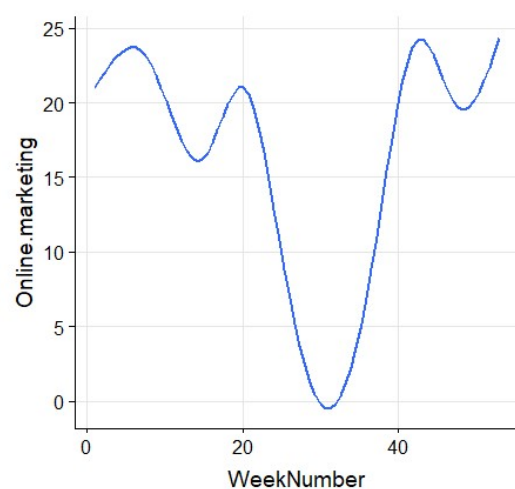
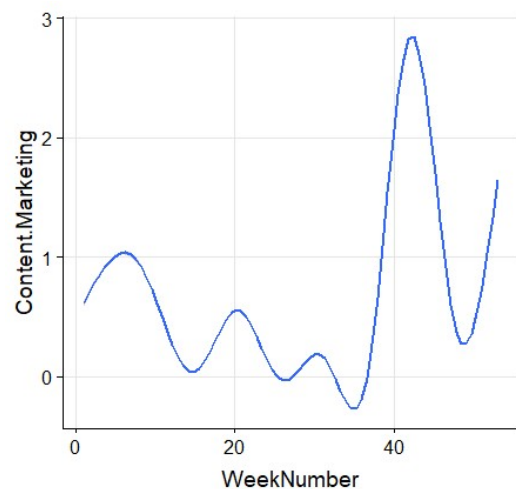
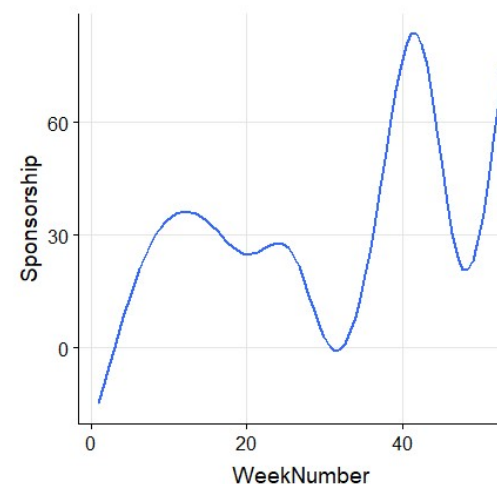
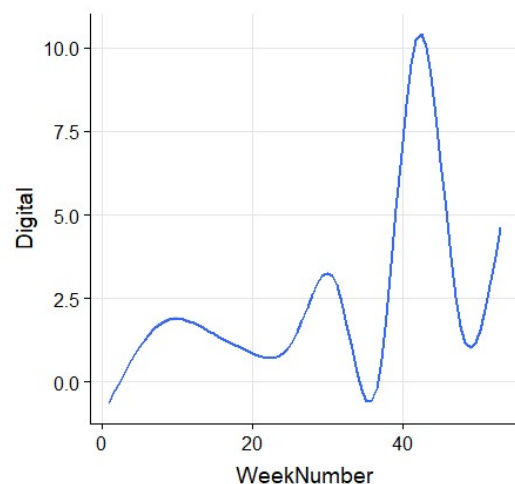
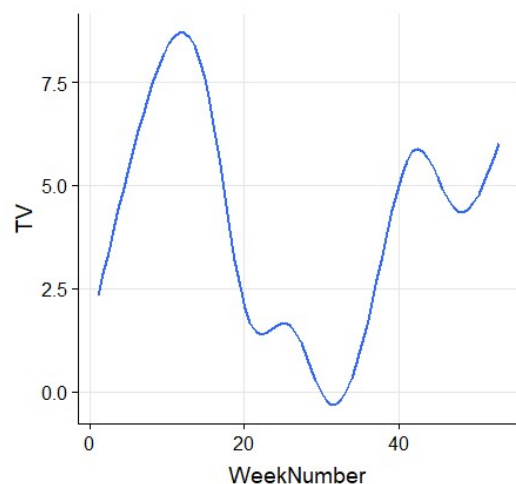
Media Marketing Spend on Regular Days for Camera Accessory



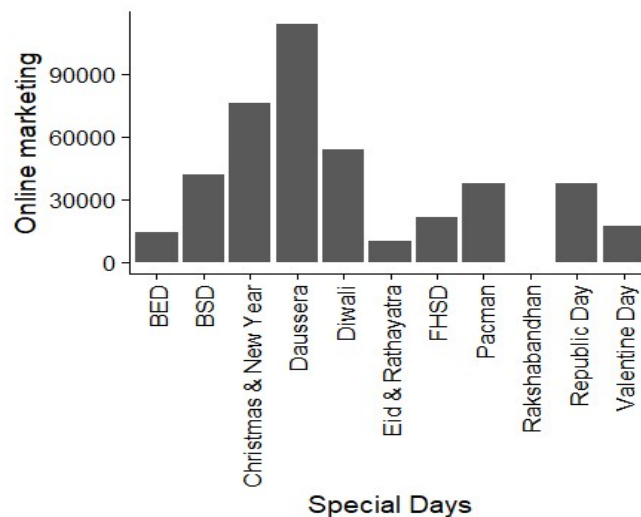
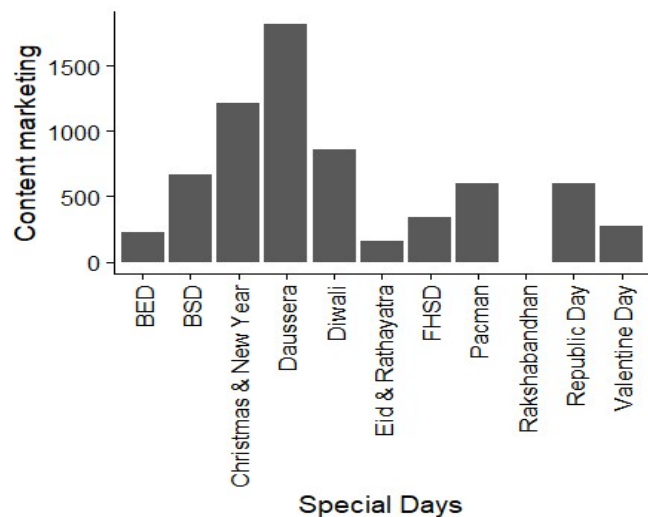
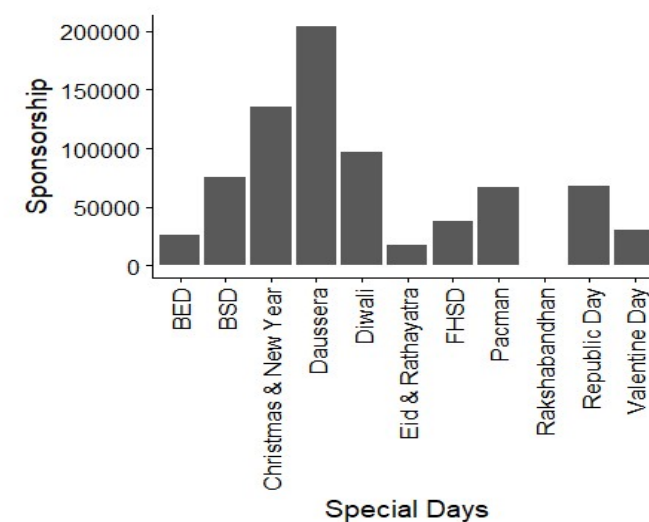
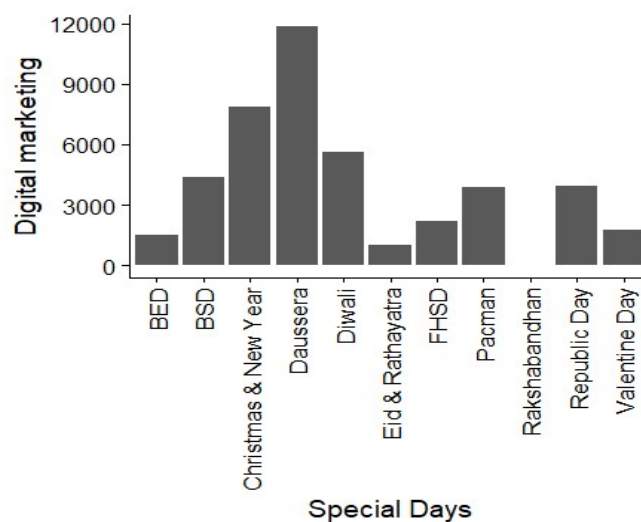
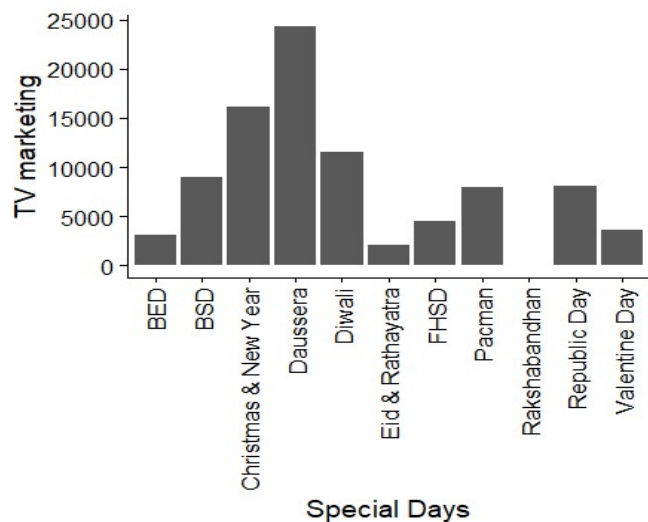
Media Marketing Spend on Special Day for Home Audio accessory



Media Marketing Spend on Regular Days for Home Audio Accessory



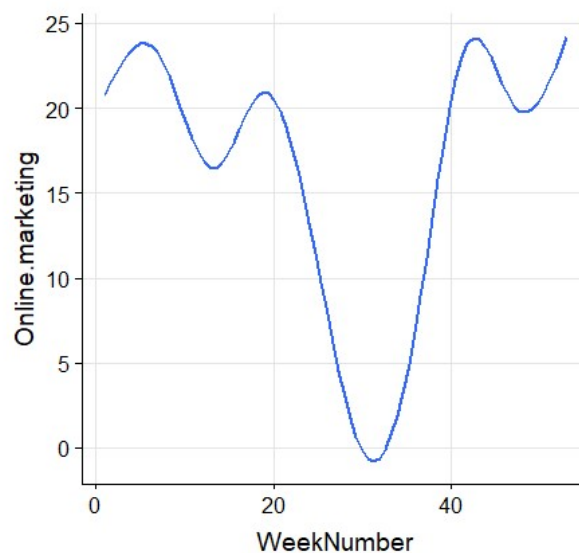
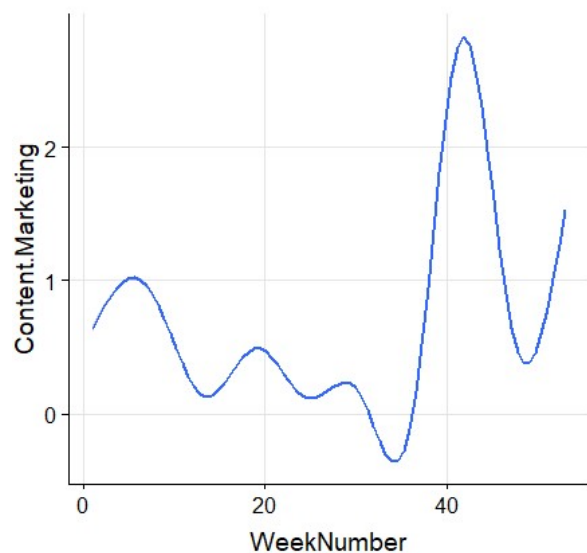
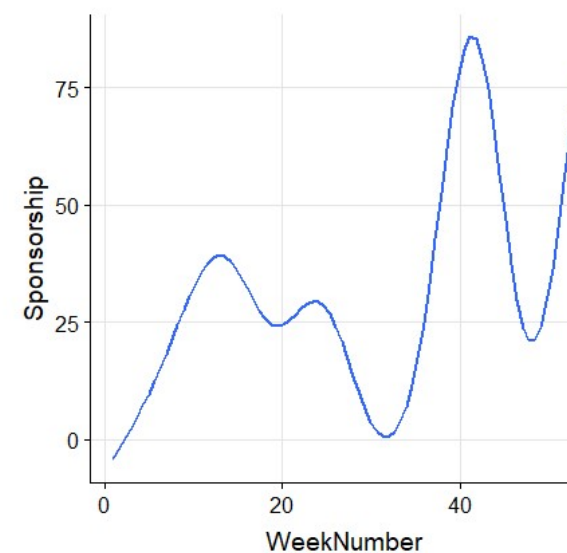
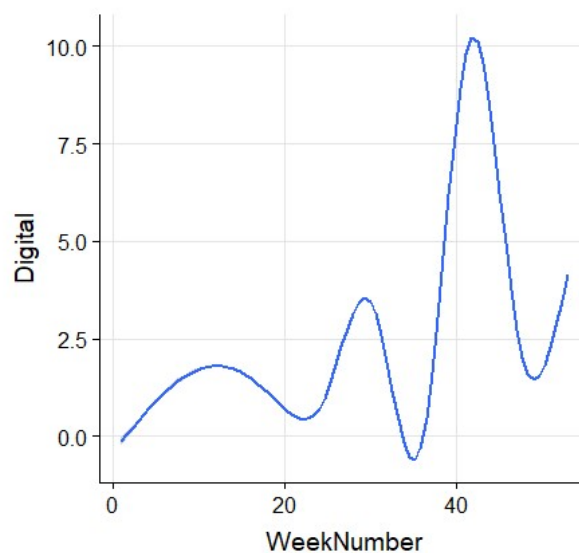
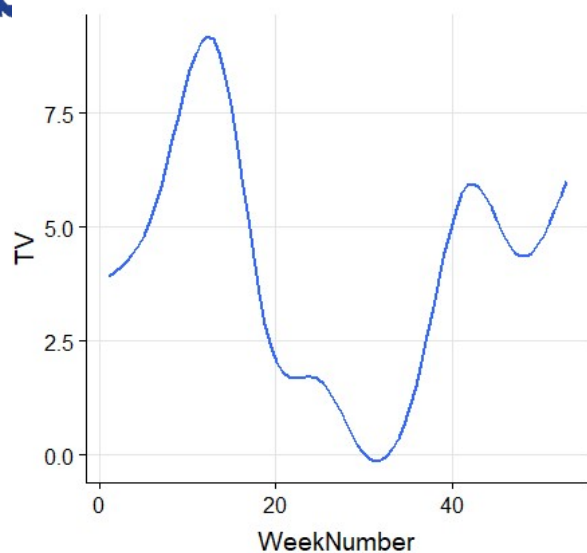
Media Marketing Spend on Specials Days for Gaming Accessory



On Rakshabandan Day there is no spend on media advertisement for Gaming accessory. They should try leveraging their media spend on these days too.

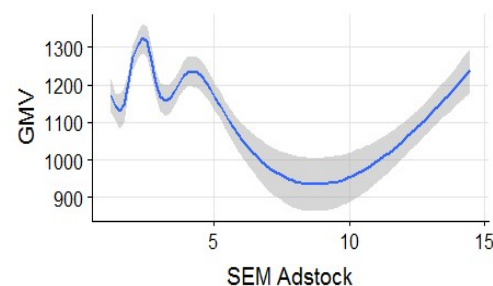
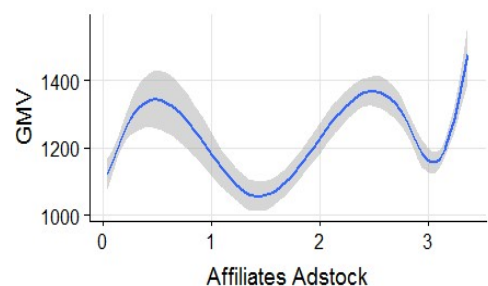
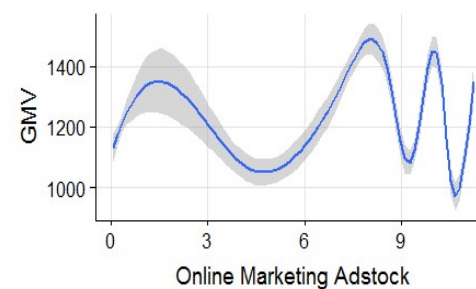
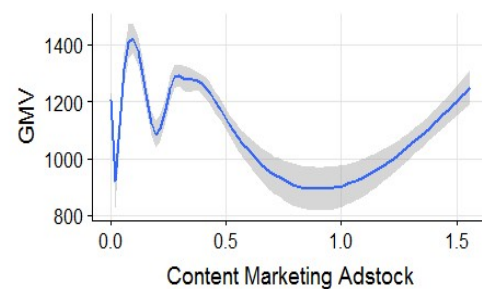
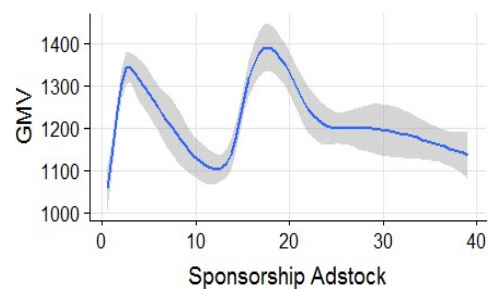
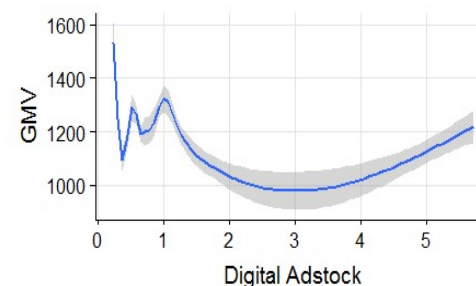
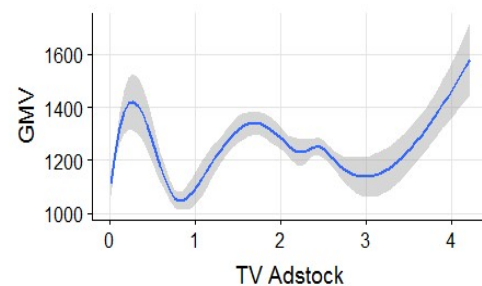


Media Marketing Spend on Regular Days for Gaming Accessory



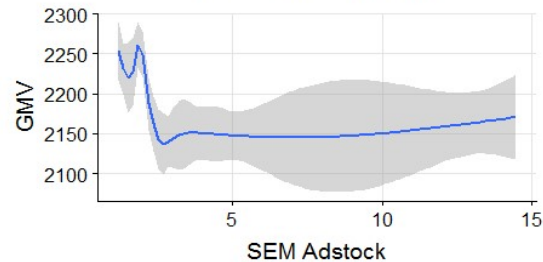
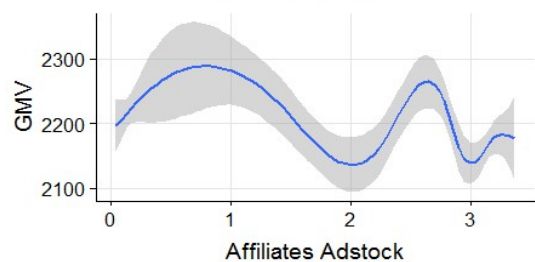
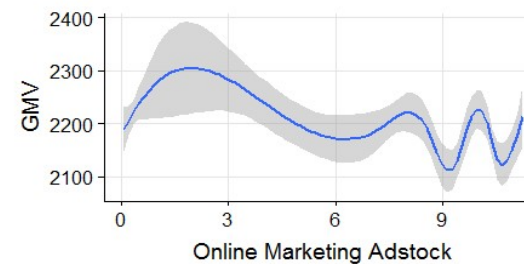
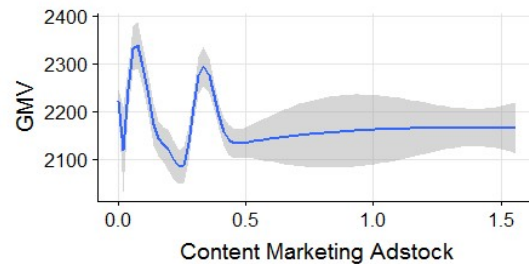
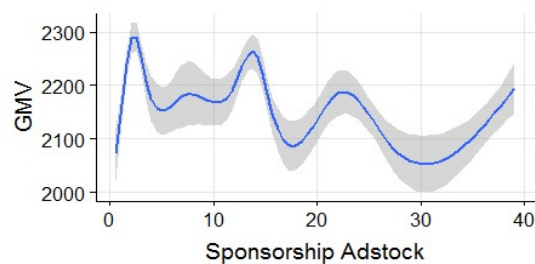
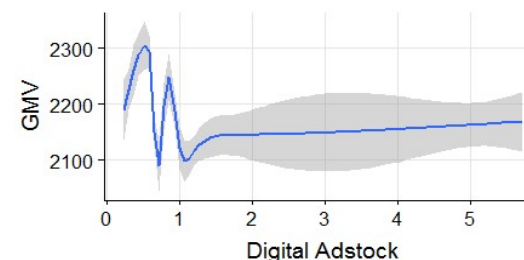
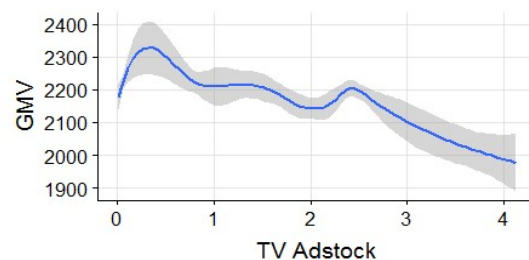
Impact of Adstock on Sales for Camera Accessory

Camera Accessory

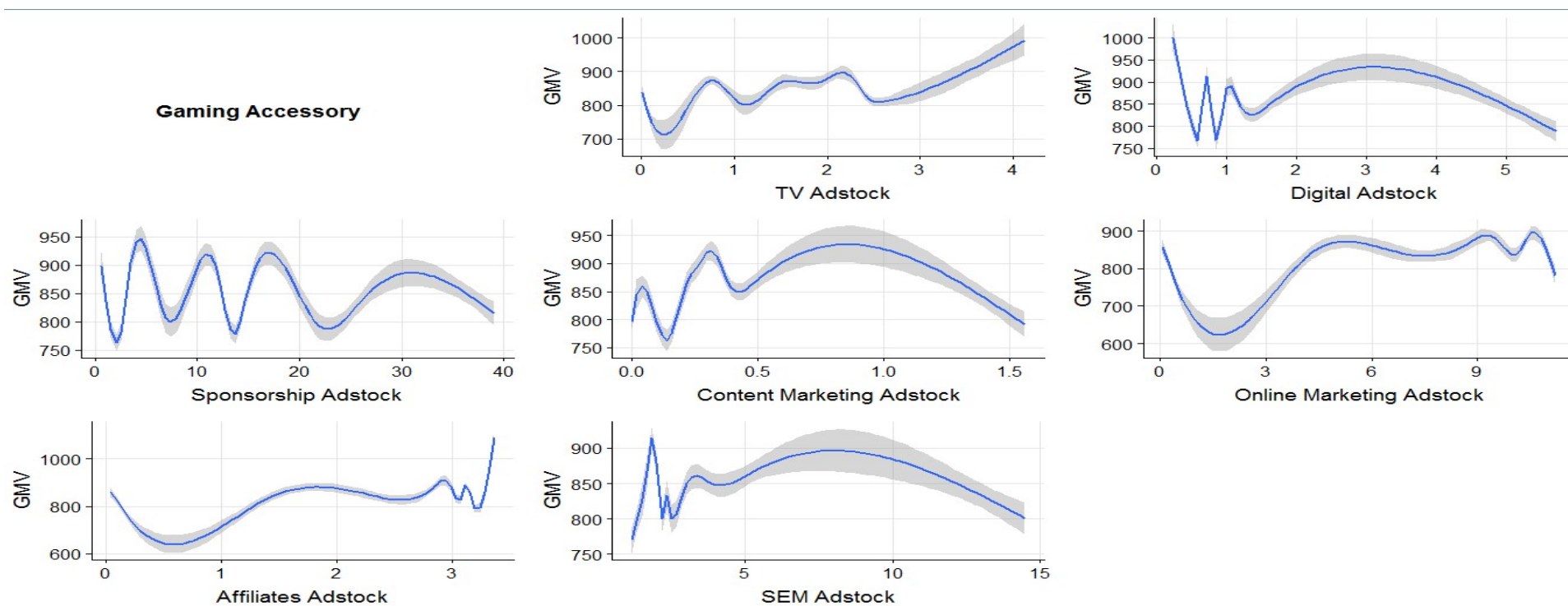


Impact of Adstock on Sales for HomeAudio Accessories

HomeAudio Accessory



Impact of Adstock on Sales for Gaming Accessory



Model Building

Following models are built to facilitate marketing team to make strategic decisions on their media spends:-

- Simple linear Model- Build the Basic Linear Model with all the KPI
- Multiplicative Model- Build the multiplicative model using the log of the individual KPIs
- Koyck Model- Build the Koyck model using the lag of the dependent variable
- Distributed Lag Model- Build the distributed lag model using the past lags of both the dependent and the independent variables



Linear Model for Camera Linear Model, HomeAudio and Gaming



```
Camera Linear Model: lm(formula = gmv ~ deliverycdays + promotional_offer + sla +  
  product_procurement_sla + is_special_week + nps_score + Digital_Adstock +  
  Online_marketing_Adstock + SEM_Adstock + price_tagMass_Product,  
  data = lm_data_camera_1)  
Home Audio Linear Model: lm(formula = gmv ~ promotional_offer + is_special_week + SEM_Adstock +  
  price_tagMass_Product, data = lm_data_homeaudio_1)  
Gaming Linear Model: lm(formula = gmv ~ deliverycdays + is_special_week + nps_score +  
  price_tagAspiring_Product + price_tagMass_Product, data = lm_data_gaming_1)
```

	Camera Linear Model		Home Audio Linear Model		Gaming Linear Model	
(Intercept)	2463062.446***	(74826.480)	1672187.654***	(107898.289)	1000513.195***	(36560.100)
deliverycdays	434283.506***	(106245.832)			196835.166***	(38480.544)
promotional_offer	723986.950***	(171902.170)	624530.220***	(133586.790)		
sla	339068.797***	(89563.259)				
product_procurement_sla	540118.282***	(115679.338)				
is_special_week	353133.424***	(85073.142)	393887.897***	(110223.508)	169777.805***	(37968.267)
nps_score	-676874.220***	(193158.301)			-309988.913***	(37392.653)
Digital_Adstock	1743190.818**	(607824.496)				
online_marketing_Adstock	434007.380**	(157946.971)				
SEM_Adstock	-2010264.978**	(678740.723)	447306.371***	(109392.493)		
price_tagMass_Product	-987582.214***	(172475.844)	1590227.128***	(132748.239)	938510.436***	(43366.777)
price_tagAspiring_Product					416714.718***	(43173.790)
R-squared	0.589		0.738		0.782	
sigma	755710.401		1294779.472		452223.464	
F	13.022		97.688		105.460	
adj. R-squared	0.543		0.730		0.775	



Multiplicative Model for Camera Linear Model, HomeAudio and Gaming

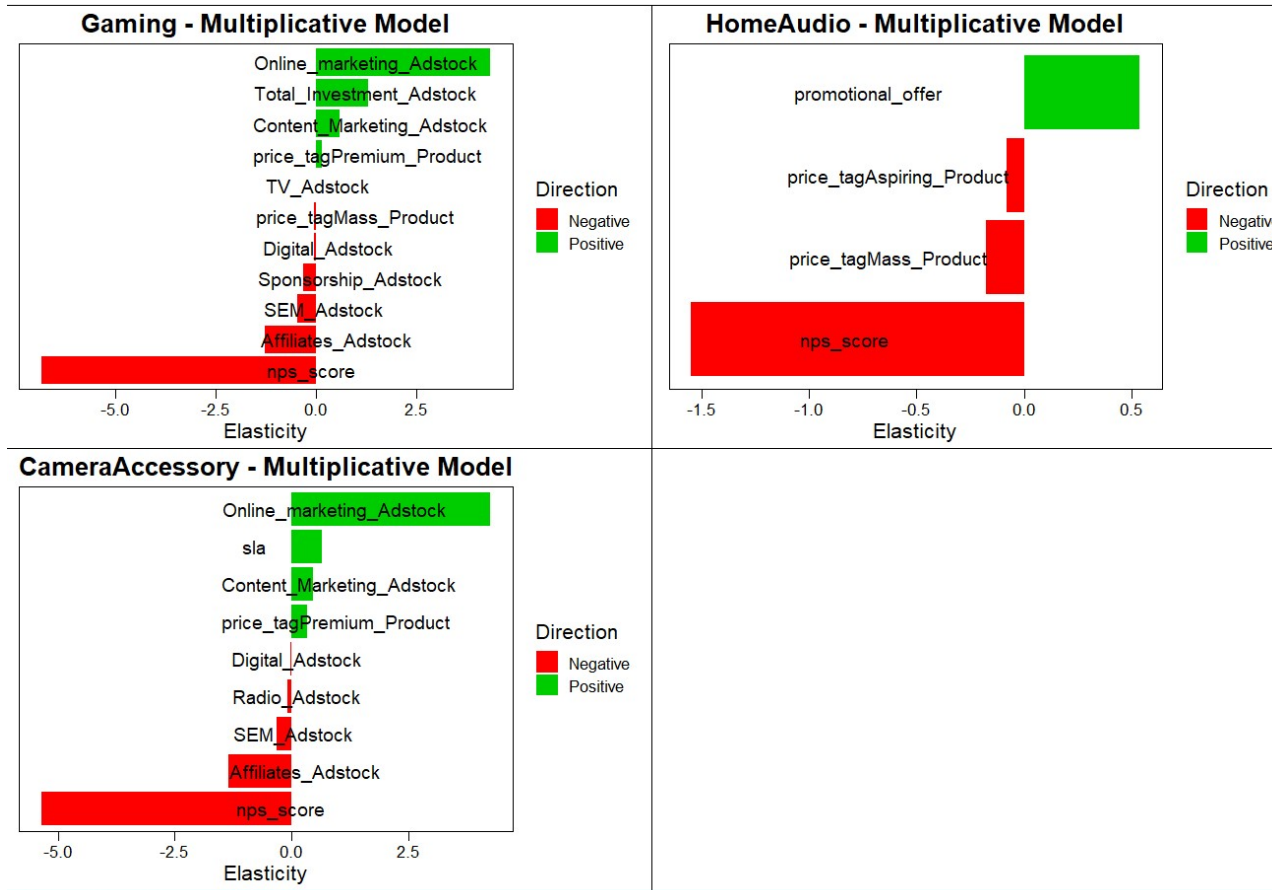


Calls:

```
Camera Multiplicative Model: lm(formula = gmv ~ sla + nps_score + Digital_Adstock + Content_Marketing_Adstock +
  Online_marketing_Adstock + Affiliates_Adstock + SEM_Adstock +
  Radio_Adstock + price_tagPremium_Product, data = data_CameraAccessory_Multi_1)
Home Audio Multiplicative Model: lm(formula = gmv ~ promotional_offer + price_tagMass_Product +
  nps_score + price_tagAspiring_Product, data = data_HomeAudio_Multi_2)
Gaming Multiplicative Model: lm(formula = gmv ~ nps_score + Total_Investment_Adstock + TV_Adstock +
  Digital_Adstock + Sponsorship_Adstock + Content_Marketing_Adstock +
  Online_marketing_Adstock + Affiliates_Adstock + SEM_Adstock +
  price_tagMass_Product + price_tagPremium_Product, data = data_Gaming_Multi_2)
```

	Camera Multiplicative Model		Home Audio Multiplicative Model		Gaming Multiplicative Model	
(Intercept)	34.988*	(14.685)	29.383***	(4.930)	47.081***	(13.546)
sla	5.125***	(0.557)				
nps_score	-19.701***	(3.694)	-5.163***	(1.139)	-22.441***	(3.436)
Digital_Adstock	3.303***	(0.408)			3.677***	(0.422)
Content_Marketing_Adstock	-2.929***	(0.276)			-3.129***	(0.267)
Online_marketing_Adstock	34.865***	(3.643)			32.525***	(3.761)
Affiliates_Adstock	-30.525***	(3.459)			-27.803***	(3.827)
SEM_Adstock	-4.205***	(0.573)			-5.593***	(0.686)
Radio_Adstock	0.603***	(0.124)				
price_tagPremium_Product	-2.077***	(0.343)			-1.106***	(0.069)
promotional_offer			1.992***	(0.423)		
price_tagMass_Product			1.526***	(0.093)	0.363***	(0.067)
price_tagAspiring_Product			0.677***	(0.079)		
Total_Investment_Adstock					4.996***	(0.908)
TV_Adstock					-1.360***	(0.400)
Sponsorship_Adstock					-1.743***	(0.300)
R-squared	0.826		0.825		0.860	
sigma	0.760		0.817		0.787	
F	48.579		163.564		78.494	
adj. R-squared	0.809		0.820		0.849	

Elasticity of the different variables w.r.t the overall sales for Multiplicative Model



- Thus going with the Multiplicative Model,
 - For gaming Accessory ElecKart should focus more on the Online channel ,Content Marketing and decrease its spending on Sponsorships w.r.t the sales.
 - For Home Accessory ElecKart should focus more on the Promotion
 - For Camera Accessory ElecKart should focus more on the Online channel ,Content Marketing and decrease its spending on Radio channel, Affiliates w.r.t the sales.

Koyck Model

calls:

```
Camera kyock Model: lm(formula = gmv ~ deliverycdays + promotional_offer + sla +
  product_procurement_sla + is_special_week + nps_score + Digital_Adstock +
  SEM_Adstock + price_tagMass_Product, data = kyock_data_camera_1)
```

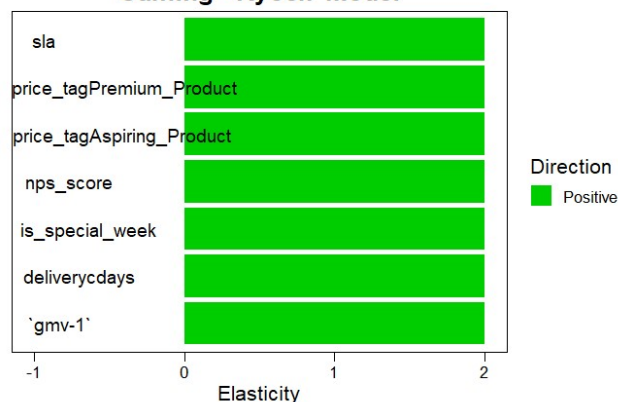
```
Home Audio kyock Model: lm(formula = gmv ~ promotional_offer + is_special_week + SEM_Adstock +
  price_tagAspiring_Product + price_tagPremium_Product, data = kyock_data_homeaudio_1)
```

```
Gaming kyock Model: lm(formula = gmv ~ deliverycdays + sla + is_special_week + nps_score +
  `gmv-1` + price_tagPremium_Product + price_tagAspiring_Product,
  data = kyock_data_gaming_1)
```

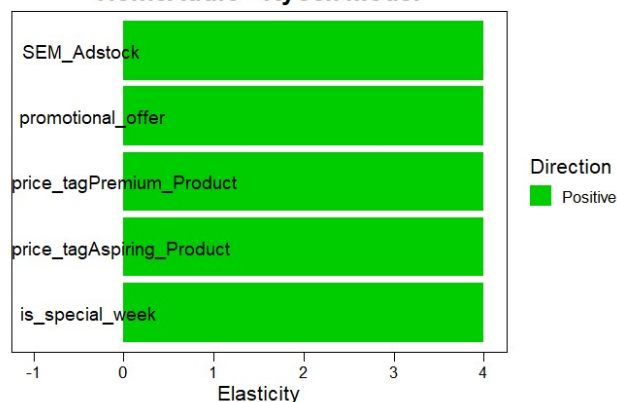
	Camera kyock Model		Home Audio kyock Model		Gaming kyock Model	
(Intercept)	2463062.446***	(77444.514)	1672187.654***	(108023.684)	637772.913***	(95616.346)
deliverycdays	430231.080***	(109952.575)			166287.015***	(39550.584)
promotional_offer	644570.746***	(175384.060)	668719.538***	(144115.282)		
sla	326218.357***	(92570.450)			112336.126**	(40840.011)
product_procurement_sla	431131.709***	(112469.178)				
is_special_week	412691.589***	(85143.995)	393231.337***	(110354.488)	123041.258**	(38382.067)
nps_score	-926812.636***	(176363.097)			-259928.092***	(37137.489)
Digital_Adstock	1210336.085*	(596215.640)				
SEM_Adstock	-1433652.573*	(668071.885)	449409.555***	(109549.428)		
price_tagMass_Product	-887870.873***	(174514.838)				
price_tagAspiring_Product			-1611324.188***	(136792.398)	-436134.016***	(57811.298)
price_tagPremium_Product			-1465775.252***	(163635.254)	-1168779.486***	(65932.432)
`gmv-1`					0.363***	(0.089)
R-squared	0.555		0.739		0.814	
sigma	782151.253		1296284.209		420996.469	
F	12.724		78.105		90.435	
adj. R-squared	0.511		0.729		0.805	

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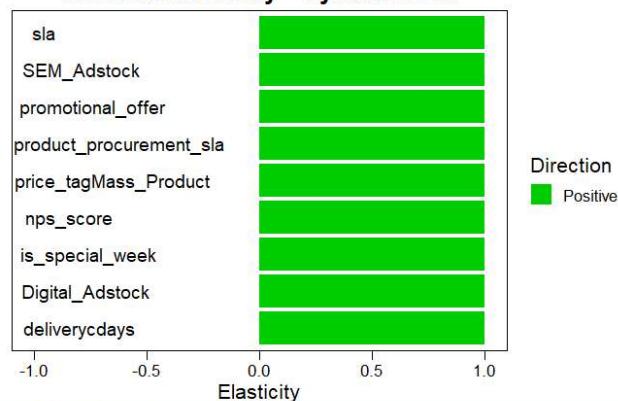
Gaming - Kyock Model



HomeAudio - Kyock Model



CameraAccessory - Kyock Model



- Positive elasticity means that increasing the value of the KPI would lead to increase in the sales figure
- Thus going with the Kyock Model
 - For gaming Accessory ElecKart should focus more on Special Weeks
 - For Home Accessory ElecKart should focus more on the Promotion, Special Weeks
 - For Camera Accessory ElecKart should focus more on the Promotion, Special Weeks, Digital Marketing, Product Procurement SLA

+xlab("variables")

Camera Accessories

- Adjusted R Square figures are based on the performance of the model on the training data.
- The SSE figures are based on the 5 fold cross validation again on the training data
- Multiplicative model has a decent Adj. R sq figure, but in other three models error is more due to high SSE value and less adj R.square value
- So, we select the Multiplicative linear model on the account of higher Adj. R Sq values

Model	Variables	Adj. R Square	Cross Validation
Linear Model	deliverycdays+promotional_offer+sla+product_procurement_sla+is_special_week+nps_score+Digital_adstock+Online_marketing_adstock+SEM_adstock+Price_tagMassProduct+Price_tagAspiringProduct	0.589	2938108
Multiplicative Model	sla+nps_score+digital_adstock+Content_Marketing_adstock+Online_marketing_adstock+Affiliates_adstock + SEM_adstock + Radio_adstock+Price_tagPremiumProduct	0.826	0.29
Koyck Model	deliverycdays+promotional_offer+sla+product_procurement_sla+is_special_week+nps_score+Digital_adstock+SEM_adstock +Price_tagMassProduct	0.55	741644
Distributed Lag Model	lm(formula = gmv ~ deliverycdays + prepaid_percentage + promotional_offer + sla + product_procurement_sla + deliverybdays + nps_score + `deliverycdays-3` + `promotional_offer-3` + `sla-3` + `product_procurement_sla-2` + `product_procurement_sla-3` + `deliverybdays-2` + `gmv-3`, data = data_dlag_camera)	0.60	52891



Home Audio Accessories

- Adjusted R Square figures are based on the performance of the model on the training data.
- The SSE figures are based on the 5 fold cross validation again on the training data
- Multiplicative model has a decent Adj. R square figure, but in other three models error is more due to high SSE value and less adj. R square value
- So, we select the Multiplicative linear model on the account of higher Adj. R Square values

Model	Variables	Adj. R Square	Cross Validation
Linear Model	promotional_offer+is_special_week+SEM_adstock+Price_tagMassProduct	0.738	4581898
Multiplicative Model	nps_score+promotional_offer+Price_tagMassProduct+Price_tagAspirinProduct	0.825	0.72
Koyck Model	promotional_offer+is_special_week+SEM_adstock+Price_tagMassProduct+Price_tagAspirinProduct	0.739	900556
Distributed Lag Model	lm(formula = gmv ~ deliverycdays + promotional_offer + product_procurement_sla + deliverybdays + nps_score + Digital_Adstock + Online_marketing_Adstock + SEM_Adstock + Radio_Adstock + `deliverycdays-1` + `deliverycdays-3` + `prepaid_percentage-1` + `promotional_offer-2` + `sla-1` + `sla-3` + `product_procurement_sla-1` + `deliverybdays-1` + `deliverybdays-2` + `deliverybdays-3` + `gmv-1` + `gmv-2` + `gmv-3`, data = data_dlag_home_audio)	0.74	780012

Gaming Accessories

- Adjusted R Square figures are based on the performance of the model on the training data.

- The SSE figures are based on the 5 fold

cross validation again on the training data

- Multiplicative model has a decent Adj. R sq figure, but in other three models error is more due to high SSE value and less add R.square value

- So, we select the Multiplicative linear model on the account of higher Adj. R Square values

Model	Variables	Adj. R Square	Cross Validation on 5 fold
Linear Model	deliverycdays+is_special_week+nps_score+Price_tagMassProduct+Price_tagAspiringProduct	0.782	2507662
Multiplicative Model	nps_score+digital_adstock+Content_Marketing_adstock+Online_marketing_adstock+Affiliates_adstock + SEM_adstock+Price_tagPremiumProduct+Price_tagAspiringProduct+TV_adstock+Total Investment_adstock+ TV_adstock	0.860	0.4
Koyck Model	deliverycdays+sla+is_special_week+nps_score+Price_tagMassProduct+Price_tagAspiringProduct+gmV-1	0.814	455814
Distributed Model	lm(gmv ~ prepaid_percentage + sla + deliverybdays + nps_score + is_special_week + TV_Adstock + Digital_Adstock + Sponsorship_Adstock + `prepaid_percentage-2` + `prepaid_percentage-3` + `promotional_offer-1` + `promotional_offer-2` + `promotional_offer-3` + `sla-1` + `sla-2` + `sla-3` + `product_procurement_sla-1` + `product_procurement_sla-2` + `deliverybdays-3` + `is_special_week-2` + `is_special_week-3` + `nps_score-3` + `gmV-1` + `gmV-3`,data_dlag_GamingAccessory)	0.5924	1888542



Then what should we do to improve Sale 3X?

Multiplicative model is performing well in all three categories . So we choose multiplicative model elasticity for improving their sales.

Thus going with the Multiplicative Model,

- I. For gaming Accessory ElecKart should focus more on the Online channel ,Content Marketing and decrease its spending on Sponsorships w.r.t the sales.
- II. For Home Accessory ElecKart should focus more on the Promotion.
- III. For Camera Accessory ElecKart should focus more on the Online channel ,Content Marketing and decrease its spending on Radio channel, Affiliates w.r.t the sales.



Thank You