

Case Study 3

Best Buy Warranty Sales Project

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Outline

- ✓ *Intro & Data Summary*
- ✓ *Model Verification*
- ✓ *Results Interpretation*
- ✓ *Proposal & Limitations*



Introduction



Background Recap

Warranty, is a Geek Squad protection plan purchased along with a product. The Geek Squad protection plan aims to insure customers after the manufacturer's warranty expires. Warranty sale becomes an important marketing strategy and offers substantial profit for Best Buy.



Research Purpose

*To identify customers who have a high propensity to purchase warranty;
To make a data-driven proposal for the potential marketing strategy.*



Data Snapshot

<i>personid</i>	<i>age</i>	<i>hisp</i>	<i>PriceCategory</i>	<i>married</i>	<i>MyBestBuy</i>	<i>hhincome</i>	<i>appliances</i>	<i>Warranty</i>	<i>familysize</i>	<i>productgeneration</i>	<i>newcustomer</i>	<i>weekend</i>
54963010	62	0	12	0	0	0	0	0	2	7	1	1
21629010	59	0	12	0	0	0	0	1	2	7	1	1
20421010	60	0	13	0	1	0	1	0	1	8	0	0
38784010	62	0	10	0	0	0	0	1	2	6	1	1
55630030	54	0	9	0	0	0	0	1	2	5	0	0
15893020	62	0	12	1	1	0	1	0	4	7	0	0
86263010	59	0	5	1	0	0	0	1	3	4	1	0
51270010	59	0	11	0	0	0	0	0	2	7	1	0

3,206 transactions, 13 Variables, made in March 2017 from Santa Clara Best Buy stores.



Variables of Theoretical Model



Dependent Variables

- **Warranty** = 1 if customer i purchased a Geek Squad protection plan along with the product and 0 otherwise



Variables of Interest

- **PriceCategory** = categorical variable between 0 and 17 that defines the value of the purchased product.

Other Variables of Interest

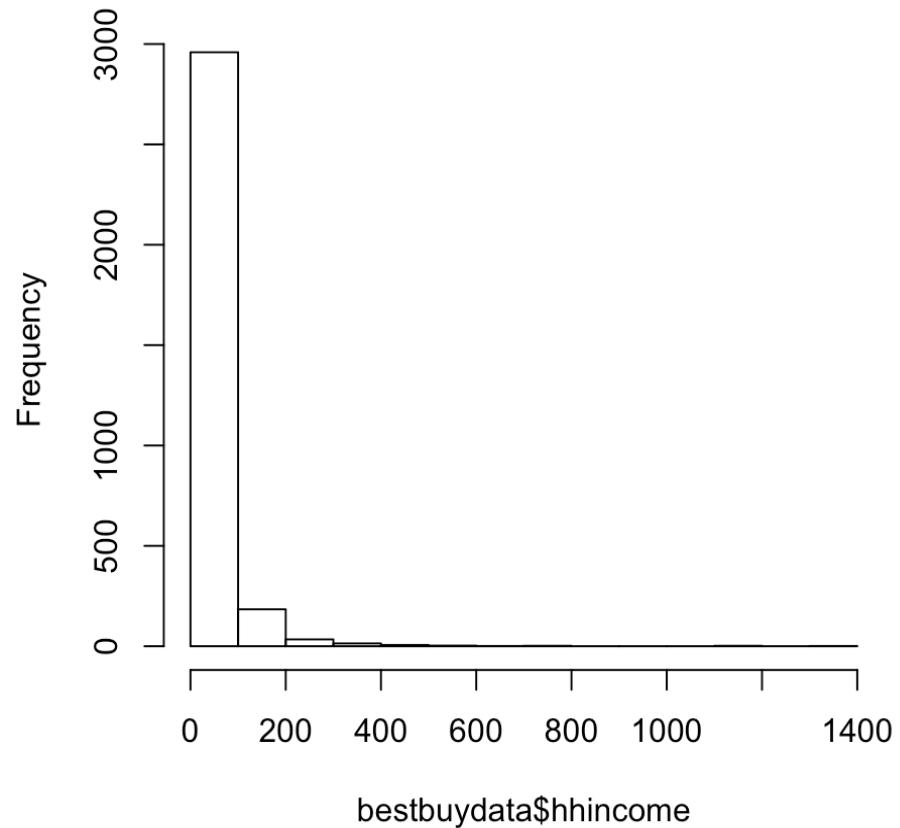


- **MyBestBuy** = 1 if customer i has MyBestBuy (i.e. company) credit card and 0 otherwise
- **Appliances** = 1 if customer i purchased a product from the “home appliances” category and 0 otherwise
- **ProductGeneration** = indicates the generation of a product within its product domain. A high number indicates a new generation.
- **Demographic: Age, Hisp, Married, Hhincome, Familysize**

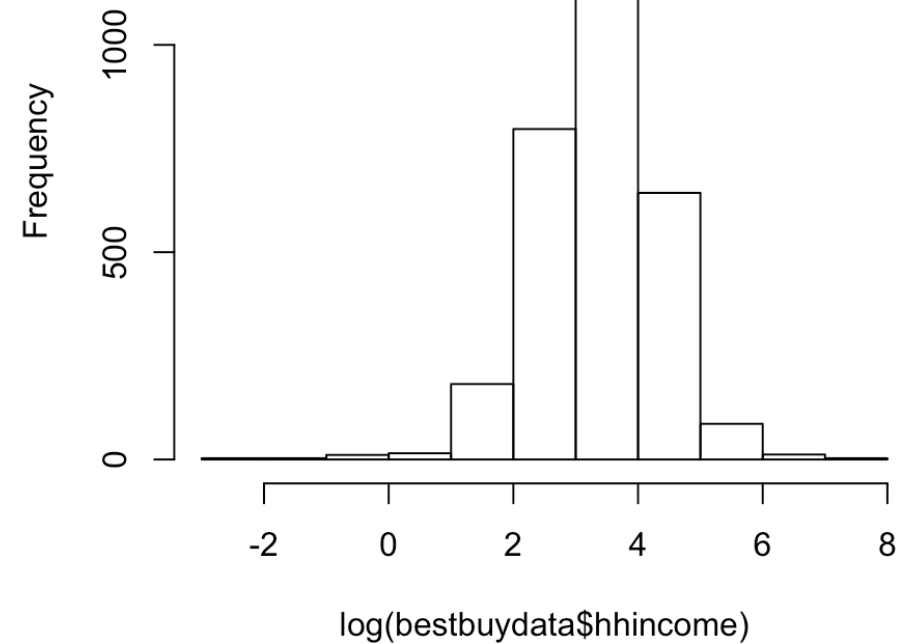


Log-transformation

Histogram of bestbuydata\$hhincome



Histogram of log(bestbuydata\$hhincome)



The Theoretical Model

✓ *Logit*

Warranty_i

$$\begin{aligned} &= \beta_0 + \beta_1 \text{PriceCategory}_i + \beta_2 \text{Appliances}_i + \beta_3 \log(1 + \text{hhincome})_i \\ &+ \beta_4 \text{ProductGeneration}_i + \beta_5 \text{MyBestBuy}_i + \beta_6 \text{Hisp}_i + \beta_7 \text{Age}_i \\ &+ \beta_8 \text{Married}_i + \beta_9 \text{FamilySize}_i + \varepsilon_i \end{aligned}$$



Model Verification



Multicollinearity

```
> vif(df1)
```

	Variables	VIF
1	bestbuydata.Warranty	1.093986
2	bestbuydata.PriceCategory	9.622327
3	bestbuydata.productgeneration	9.228072
4	bestbuydata.MyBestBuy	1.136930
5	bestbuydata.appliances	1.192873
6	bestbuydata.age	1.169291
7	bestbuydata.married	4.577358
8	bestbuydata.hhincome1	1.670485
9	bestbuydata.hisp	1.160839
10	bestbuydata.familysize	4.221931

```
> vif(df3)
```

	Variables	VIF
1	bestbuydata.Warranty	1.086805
2	bestbuydata.PriceCategory	1.444348
3	bestbuydata.MyBestBuy	1.135174
4	bestbuydata.appliances	1.192517
5	bestbuydata.age	1.166522
6	bestbuydata.familysize	1.231464
7	bestbuydata.hhincome1	1.564901
8	bestbuydata.hisp	1.160560



Interaction term

Call:

```
glm(formula = Warranty ~ age + hisp + PriceCategory + MyBestBuy +  
    hhincome1 + appliances + familysize + PriceCategory:appliances,  
    family = binomial, data = bestbuydata)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.0344	-1.1836	0.7358	0.9121	2.1384

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.45039	0.78212	-1.854	0.0637 .
age	-0.01844	0.01139	-1.619	0.1055
hisp	-1.39394	0.16442	-8.478	< 2e-16 ***
PriceCategory	0.13194	0.02327	5.670	1.43e-08 ***
MyBestBuy	0.15049	0.08420	1.787	0.0739 .
hhincome1	0.25545	0.05432	4.703	2.57e-06 ***
appliances	2.74865	0.31664	8.681	< 2e-16 ***
familysize	0.34960	0.04062	8.606	< 2e-16 ***
PriceCategory:appliances	-0.24913	0.02758	-9.033	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 4255.7 on 3205 degrees of freedom
Residual deviance: 3909.7 on 3197 degrees of freedom
AIC: 3927.7

Number of Fisher Scoring iterations: 4

Theory:

- *More expensive appliances come with longer warranties*
- *Less likely that consumers will buy Best Buy warranty*



Interaction term tests

```
> AIC(logit2,logit3)
```

```
      df      AIC
```

```
logit2  8 4014.789
```

```
logit3  9 3927.695
```

```
> BIC(logit2,logit3)
```

```
      df      BIC
```

```
logit2  8 4063.371
```

```
logit3  9 3982.350
```

```
> with(logit2, pchisq(null.deviance - deviance, df.null  
- df.residual, lower.tail = FALSE))
```

```
[1] 9.177856e-52
```

```
> with(logit3, pchisq(null.deviance - deviance, df.null  
- df.residual, lower.tail = FALSE))
```

```
[1] 6.322924e-70
```

```
> anova(logit2, logit3, test="Chisq")
```

```
Analysis of Deviance Table
```

```
Model 1: Warranty ~ age + hisp + PriceCategory + MyBestBuy + hhincome1 +  
  appliances + familysize
```

```
Model 2: Warranty ~ age + hisp + PriceCategory + MyBestBuy + hhincome1 +  
  appliances + familysize + PriceCategory:appliances
```

	Resid. Df	Resid. Dev	Df	Deviance	Pr(>Chi)
1	3198	3998.8			
2	3197	3909.7	1	89.094	< 2.2e-16 ***

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```



Heteroskedasticity

- *There is heteroskedasticity*
- *The coefficients and p-values do not change when we include robust standard errors*

```
> gqtest(logit3) # p-value is insignificant
```

Goldfeld-Quandt test

```
data: logit3
GQ = 0.97535, df1 = 1594, df2 = 1594, p-value = 0.6908
alternative hypothesis: variance increases from segment 1 to 2
```

```
> bptest(logit3) # p-value is significant
```

studentized Breusch-Pagan test

```
data: logit3
BP = 53.831, df = 8 p-value = 7.445e-09
```

```
> coeftest(logit3, vcov = vcovHC(logit3, "HC1"))
```

z test of coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-1.450389	0.788276	-1.8400	0.06578	.
age	-0.018438	0.011505	-1.6026	0.10901	
hisp	-1.393938	0.160750	-8.6715	< 2.2e-16	***
PriceCategory	0.131937	0.022821	5.7814	7.408e-09	***
MyBestBuy	0.150488	0.084918	1.7722	0.07637	.
hhincome1	0.255453	0.054002	4.7304	2.240e-06	***
appliances	2.748647	0.312145	8.8057	< 2.2e-16	***
familysize	0.349603	0.040773	8.5744	< 2.2e-16	***
PriceCategory:appliances	-0.249134	0.027057	-9.2078	< 2.2e-16	***

Signif. codes:	0	****	0.001	***	0.01 ** 0.05 '.' 0.1 ' ' 1



Final Model

Logit

$$\text{Warranty}_i = -1.45 + 0.13\text{PriceCategory}_i + 2.75\text{Appliances}_i + 0.26\log(1 + \text{hhincome})_i + 0.15\text{MyBestBuy}_i - 1.39\text{Hisp}_i - 0.02\text{age}_i + 0.35\text{Familysize}_i - 0.25\text{PriceCategory}_i * \text{Appliances}_i$$

Probit

$$\text{Warranty}_i = -0.89 + 0.08\text{PriceCategory}_i + 1.66\text{Appliances}_i + 0.16\log(1 + \text{hhincome})_i + 0.09\text{MyBestBuy}_i - 0.85\text{Hisp}_i - 0.01\text{age}_i + 0.21\text{Familysize}_i - 0.15\text{PriceCategory}_i * \text{Appliances}_i$$



Probit Model

Call:

```
glm(formula = Warranty ~ age + hisp + PriceCategory + MyBestBuy +  
    hhincome1 + appliances + familysize + PriceCategory:appliances,  
    family = binomial(link = "probit"), data = bestbuydata)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.0591	-1.1836	0.7373	0.9160	2.1812

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.888531	0.471974	-1.883	0.0598 .
age	-0.011061	0.006886	-1.606	0.1082
hisp	-0.851901	0.098374	-8.660	< 2e-16 ***
PriceCategory	0.079961	0.013718	5.829	5.58e-09 ***
MyBestBuy	0.092168	0.051111	1.803	0.0713 .
hhincome1	0.155066	0.032879	4.716	2.40e-06 ***
appliances	1.658999	0.188129	8.818	< 2e-16 ***
familysize	0.213400	0.024731	8.629	< 2e-16 ***
PriceCategory:appliances	-0.150419	0.016286	-9.236	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

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Null deviance: 4255.7 on 3205 degrees of freedom
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Number of Fisher Scoring iterations: 4



Final Logit Model

$$\begin{aligned} &Warranty_i \\ &= -1.45 + 0.13PriceCategory_i + 2.75Appliances_i \\ &+ 0.26\log(1 + hhincome)_i + 0.15MyBestBuy_i - 1.39Hispanic_i \\ &- 0.02age_i + 0.35Familysize_i + \varepsilon_i \end{aligned}$$



Final Model

Call:

```
glm(formula = Warranty ~ age + hisp + PriceCategory + MyBestBuy +  
    hhincome1 + appliances + familysize + PriceCategory:appliances,  
    family = binomial, data = bestbuydata)
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Deviance Residuals:

Min	1Q	Median	3Q	Max
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Interpretation of results

The likelihood ratio test results shows that our model has a good fit!



```
> with(logit7,null.deviance-deviance)
[1] 346.0453
> with(logit7,df.null-df.residual)
[1] 8
> with(logit7,pchisq(null.deviance-deviance,df.null-df.residual,lower.tail = FALSE))
[1] 6.322924e-70
```



Interpretation of results

Our model has a 67% correct Classification rate!

```
> pred<-predict(logit7,data=BestBuy,type="response")  
> prediction<-ifelse(pred>=0.5,1,0)  
> error<-mean(prediction!=BestBuy$Warranty)  
> 1-error  
[1] 0.6746725
```

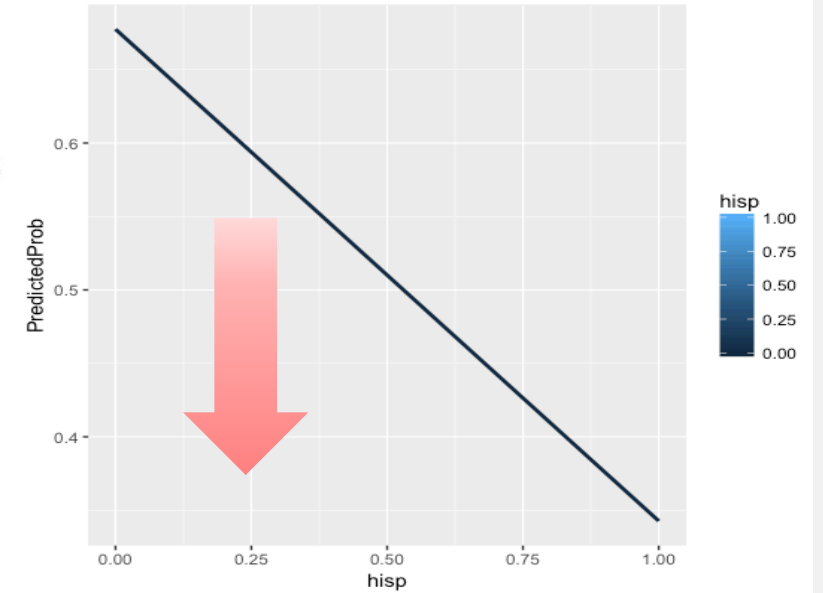
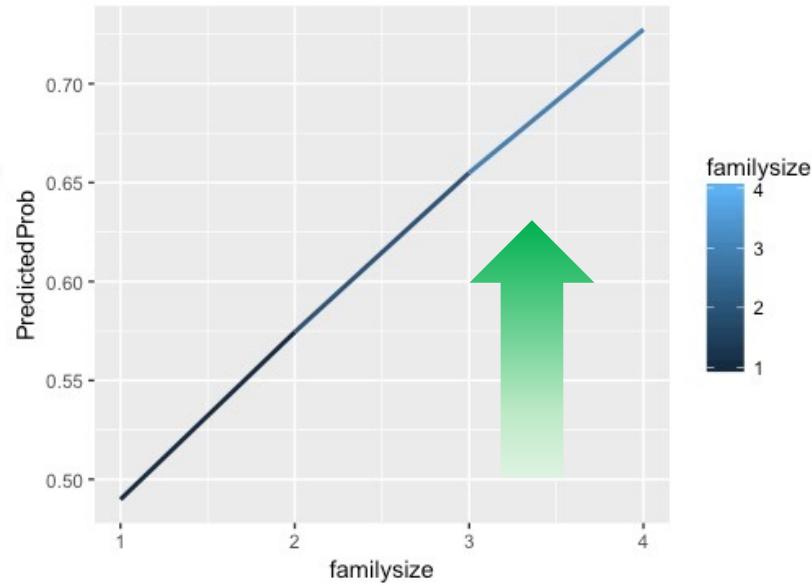
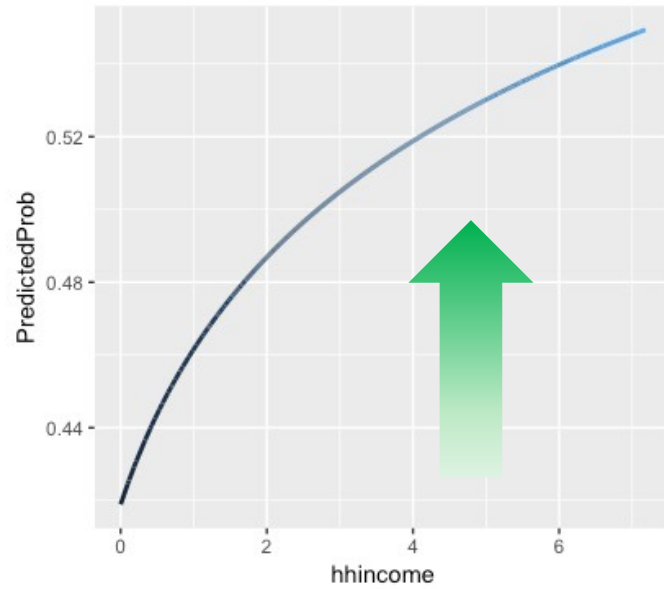


Interpretation of results(individual variable)

<i>Variables</i>	<i>Relationship with buying warranty</i>	β	e^{β}	<i>Marginal Effects</i>
Household Income	Positive	0.255(log)	1.290(log)	0.059%
Family size	Positive	0.350	1.419	8%
Hispanic	Negative	-1.394	0.248	-33%
Price Category	Interaction	See next slide		
Appliances	Interaction			
Age	unknown	NA		
MyBestBuy	unknown	NA		



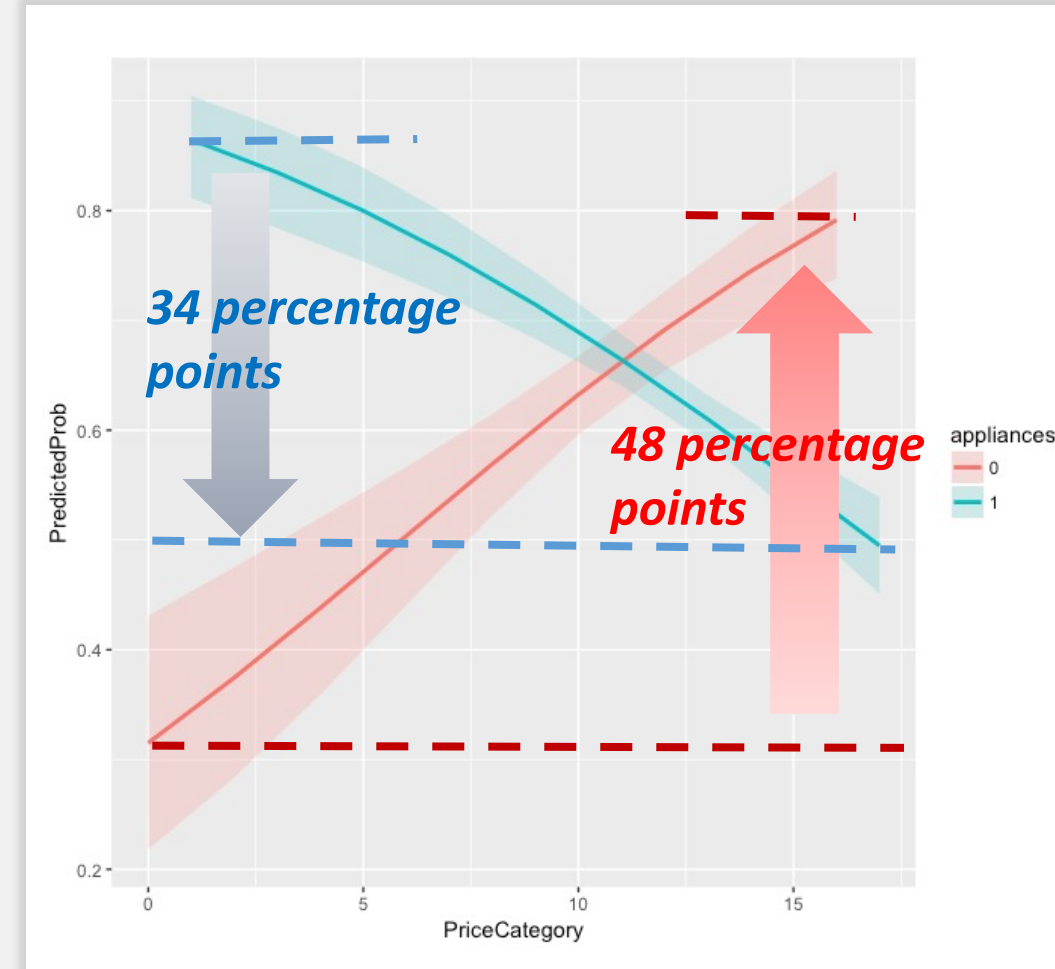
Interpretation of results(individual variable)



Interpretation of results(interaction term)

Holding other variables constant...

Price Category	Non-Appliances	Appliances
β	0.132	-0.117(=0.132-0.249)
e^{β} (odds ratio)	1.141	0.890
Marginal Effects	3%	-2%



Interpretation of results(individual variable)

<i>Variables</i>	<i>Relationship with buying warranty</i>	β	e^{β}	<i>Marginal Effects</i>
Household Income	Positive	0.255(log)	1.290(log)	0.059%
Family size	Positive	0.350	1.419	8%
Hispanic	Negative	-1.394	0.248	-33%
Price Category	Interaction	See next slide		
Appliances	Interaction			
Age	unknown	NA		
MyBestBuy	unknown	NA		



Implications for Marketing Strategy

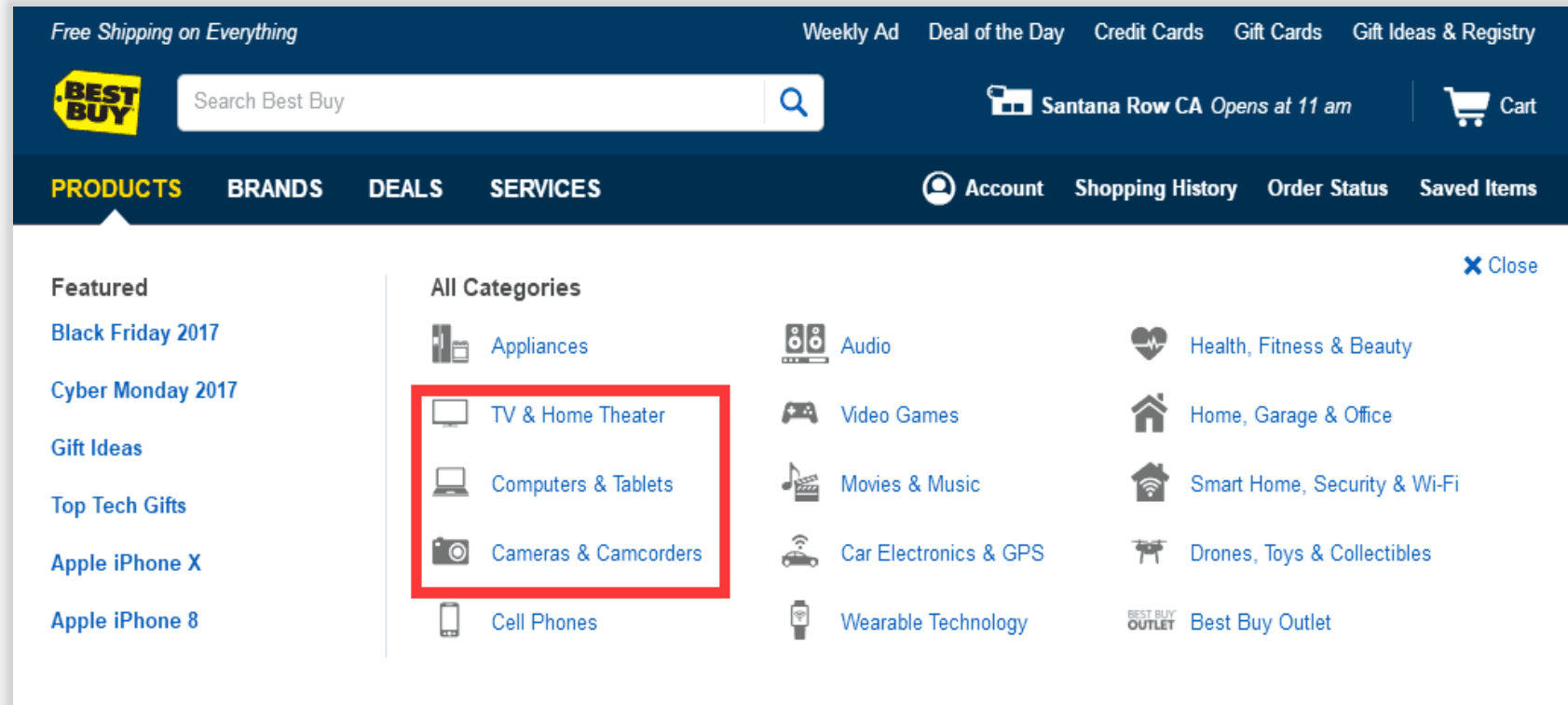
✓ *The important things to decide our target*

- *No. 1: Non-appliances with higher price*
- *No. 2: Appliances with lower price*
- *No. 3: Non-Hispanic*
- *No. 4: Larger Family size*
- *No. 5: Higher income*



Implications for Marketing Strategy

- *No. 1: Non-appliances with higher price*



Target Customers: Likely to decide company supplies, buy new house, want tech for recreation.




Implications for Marketing Strategy

■ No. 2: Appliances with lower price

Major Kitchen Appliances

>

From crisp and clean to sleek and modern, create the look of luxury in your kitchen.



Refrigerators

Dishwashers

Freezers & Ice Makers

Microwaves

Ranges, Cooktops & Ovens

Kitchen Appliance Packages


Range Hoods

Appliance Parts & Accessories

Small Kitchen Appliances

>

Enhance your cooking skills with small appliances that simplify your life and style up your countertop.



Coffee, Tea & Espresso

Pots & Pans

Blenders & Juicers

Slow Cookers, Crock Pots & Roaster Ovens

Mixers

Food Processors

Toasters

Shop all

Target Customers: High tendency to make decisions for buying small appliances and home improvement, focused on convenience.



Implications for Marketing Strategy

- *No. 3: Non-Hispanic*

Target Customers: English speakers

- *No. 4: Larger Family size*

Target Customers: Family-oriented

- *No. 5: Higher income*

Target Customers: Business-oriented



Marketing Plan Proposal



- *Sending follow-up emails for TV/Home Theater, Computer, Camera, Small kitchen appliances buyers*



- *Increase Social Media marketing through influencers for family-oriented customers*



- *Releasing ads at transportation such as airport for Business-oriented customers*



- *Won't increase budget for Spanish-language promotion, only increase English ads*



Online



Offline



Limitation

- 1. Some variables are not convincing.*
- 2. Narrow consumer data scale.*
- 3. Limitation of demographic info.*



Thanks!



Correlation

```
> cor(mydata)
```

	personid	age	hisp	PriceCategory	married
personid	1.000000000	-0.0330527947	0.0967204647	-0.048153525	-0.03817971
age	-0.033052795	1.000000000	-0.0006346689	0.039916952	0.15756214
hisp	0.096720465	-0.0006346689	1.000000000	-0.328141800	-0.02658656
PriceCategory	-0.048153525	0.0399169525	-0.3281418001	1.000000000	0.07494838
married	-0.038179713	0.1575621439	-0.0265865550	0.074948382	1.00000000
MyBestBuy	-0.032355211	0.3115567007	-0.0659347592	0.135125725	0.12202874
hhincome	-0.018325405	0.0652357371	-0.1052860258	0.290427681	0.20854885
appliances	-0.032671960	0.1967221171	-0.1163422707	0.309767740	0.09605320
Warranty	-0.021868583	0.0129989178	-0.1847920169	0.071934974	0.22135089
familysize	-0.036133232	0.1456522428	-0.0147721428	0.064004686	0.87308567
productgeneration	-0.027111898	0.0542402527	-0.3080954399	0.944045453	0.07437216
newcustomer	-0.002354919	0.0120054043	0.0134669523	-0.005186286	0.01522605
weekend	-0.014707124	-0.0083913126	-0.0149751267	-0.011490043	0.03270994

	MyBestBuy	hhincome	appliances	Warranty	familysize
personid	-0.032355211	-0.01832540	-0.032671960	-0.02186858	-0.03613323
age	0.311556701	0.06523574	0.196722117	0.01299892	0.14565224
hisp	-0.065934759	-0.10528603	-0.116342271	-0.18479202	-0.01477214
PriceCategory	0.135125725	0.29042768	0.309767740	0.07193497	0.06400469
married	0.122028744	0.20854885	0.096053202	0.22135089	0.87308567
MyBestBuy	1.000000000	-0.01204500	0.128009299	0.05936840	0.10869629
hhincome	-0.012045000	1.000000000	0.155125760	0.10770128	0.17158996
appliances	0.128009299	0.15512576	1.000000000	0.04482788	0.06402652
Warranty	0.059368397	0.10770128	0.044827879	1.000000000	0.19979141
familysize	0.108696288	0.17158996	0.064026525	0.19979141	1.000000000
productgeneration	0.133728765	0.28408757	0.293973195	0.06093647	0.06164029
newcustomer	-0.047584715	0.01221961	-0.002667570	0.01075128	0.02360174
weekend	0.005156923	0.01489175	0.007693935	0.02250629	0.01707987



Correlation

	productgeneration	newcustomer	weekend
personid	-0.02711190	-0.002354919	-0.014707124
age	0.05424025	0.012005404	-0.008391313
hisp	-0.30809544	0.013466952	-0.014975127
PriceCategory	0.94404545	-0.005186286	-0.011490043
married	0.07437216	0.015226049	0.032709938
MyBestBuy	0.13372877	-0.047584715	0.005156923
hhincome	0.28408757	0.012219606	0.014891748
appliances	0.29397319	-0.002667570	0.007693935
Warranty	0.06093647	0.010751281	0.022506286
familysize	0.06164029	0.023601740	0.017079870
productgeneration	1.00000000	-0.011593103	-0.014526542
newcustomer	-0.01159310	1.00000000	0.010458271
weekend	-0.01452654	0.010458271	1.00000000



Probit model predict accuracy rate

```
> pred<-predict(probit1,data=BestBuy,type="response")  
> predic<-ifelse(pred>=0.5,1,0)  
> error<-mean(predic!=BestBuy$Warranty)  
> 1-error  
[1] 0.6752963
```

