

# STRUCTURAL EQUATION MODELING WITH LAVAAN TO EXPLAIN THE EFFECTS OF CO-CREATION ON CONSUMER BEHAVIOR

Rocío Alarcón López (Universidad de Murcia) Inés López López (Universidad de Murcia) \*Salvador Ruiz de Maya (Universidad de Murcia)





#### Introduction

- Traditionally, firms developed their products and consumers accepted them passively. Nowadays, consumers adopt a more active role, especially in the virtual environment (Grönroos, 2011).
- Firms are increasingly integrating customers in their innovation processes to better suit consumers' needs (Ostrom et al., 2015).
- Diverse companies have implemented co-creation activities in recent years (Kristal et al., 2016).

## threadless

























## What would be yummy as a chip? Create it and you could win!





#### Co-creation

 Co-creation: Joint value creation through a process in which providers and customers systematically interact, share information, learn and integrate resources (Prahalad and Ramaswamy, 2004).

### Goal of the study

 To analyze the effects of failed (vs. successful) co-creation experiences on consumer responses.



### Hypotheses

H1: Participation in co-creation activities leads to more positive emotions when the co-creation outcome is successful than when it is failed.

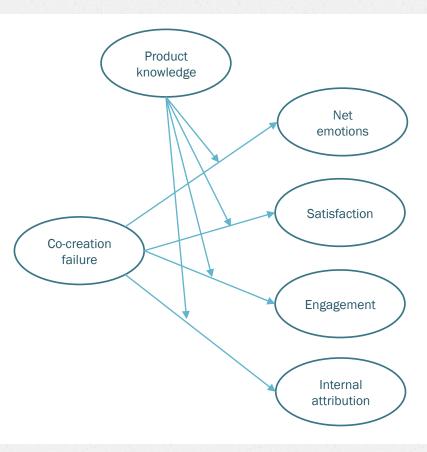
H2: Participation in co-creation activities leads to more satisfaction when the co-creation outcome is successful than when it is failed.

H3: Participation in co-creation activities leads to more consumer engagement when the co-creation outcome is successful than when it is failed.

H4: Participation in co-creation activities leads to more internal attributions when the co-creation outcome is successful than when it is failed.



### **Study:** Hypotheses







- Between-subjects experimental design where we manipulated the co-creation outcome (successful vs. failed vs. control condition)
- 144 participants
- Data were collected using a self-administered online survey





- Dependent variables:
  - Satisfaction with the co-creation outcome
  - Positive and negative emotions felt while exposed to the outcome
  - Consumer engagement
  - Internal and external attributions
- Control variable:
  - Product knowledge
- Service performance was introduced to check the manipulation of the co-creation outcome





#### CO-CREATION CONDITIONS

Nos gustaría que imaginases el siguiente escenario:

LA CROQUETERÍA GOURMET es una nueva croquetería que ha abierto en Murcia. Está buscando ideas creativas para sus nuevas croquetas. Le gustaría incluir en su carta una croqueta sabrosa, original y novedosa que cautive a los consumidores.

Para recibir las propuestas de los consumidores ha creado un **concurso de croquetas en Facebook.** La croqueta gourmet con más likes será la ganadora y obtendrá un premio de 250 euros.







#### CO-CREATION CONDITIONS

| Para ser realmente original en tu propuesta también puedes elegir otros ingredientes como frutos secos, especias o frutas tropicales. |  |
|---|--|
| Cualquier combinación de ingredientes es posible, pero recuerda que puedes elegir un máximo de TRES ingredientes.                     |  |
| A continuación, escribe los TRES ingredientes de tu croqueta.   |  |
| Por último, escribe un NOMBRE para tu croqueta gourmet.   |  |



SUCCESSFUL CO-CREATION OUTCOME











#### CONTROL CONDITION

Nos gustaría que imaginases el siguiente escenario:

LA CROQUETERÍA GOURMET es un nuevo restaurante de croquetas que ha abierto en Murcia. Imagina que planeas ir hoy a esta croquetería y puedes anticipar el pedido por Internet. Puedes elegir un máximo de tres croquetas diferentes para hacer tu pedido.

--> Haz clic para ver la carta de croquetas







#### CONTROL CONDITION

Señala las croquetas de LA CROQUETERÍA GOURMET que quieres probar.

Recuerda que puedes elegir un máximo de tres croquetas distintas.

Croqueta de berenjena

Croqueta de queso parmesano

Croqueta de atún

Croqueta de jamón de bellota

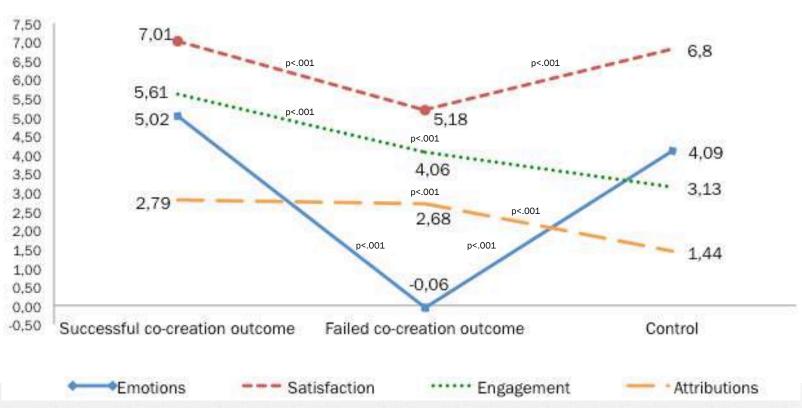
Croqueta de secreto ibérico







ANCOVA results. Covariate: Product knowledge



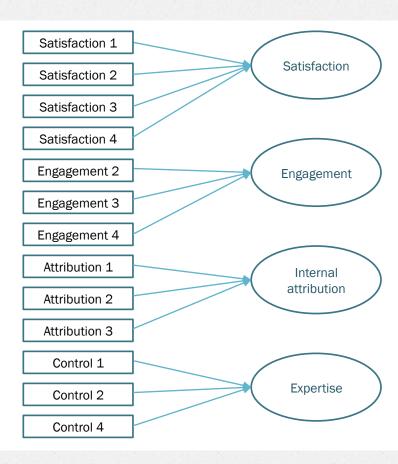


- Can we better understand the relationship among the variables?
- First we check reliability and validity through Confirmatory Factor Analysis (CFA)
- Yves Rosseel (2012). lavaan: An R Package for Structural Equation Modeling. Journal of Statistical Software, 48(2), 1-36

```
CFAmodel <- '
# measurement
satisfaction =~ satis1 + satis2 + satis3 + satis4
engagement =~ engage2 + engage3 + engage4
attribution =~ attribution1 + attribution2 + attribution3
expertise =~ control1 + control2 + control4
'
fitCFA <- cfa(CFAmodel, data = mydata)
summary(fitCFA, fit.measures = TRUE)
fitMeasures(fitCFA)
inspect (fitCFA, "cor.ov")
standardizedSolution(fitCFA)
MI <- modificationIndices(fitCFA)
subset(MI, mi > 10)
```



### **Study: Model**



Reliability and validity through Confirmatory Factor Analysis (CFA)

| tonability and validit                   | . y CI II O OI |
|--|----------------|
| Number of observations                   | 144            |
| Estimator                                | ML             |
| Model Fit Test Statistic                 | 98.043         |
| Degrees of freedom                       | 59             |
| P-value (Chi-square)                     | 0.001          |
| Model test baseline model:               |                |
| Minimum Function Test Statistic          | 1669.684       |
| Degrees of freedom                       | 78             |
| P-value                                  | 0.000          |
| User model versus baseline model:        |                |
| Comparative Fit Index (CFI)              | 0.975          |
| Tucker-Lewis Index (TLI)                 | 0.968          |
| Loglikelihood and Information Criteria:  |                |
| Loglikelihood user model (H0)            | -3582.655      |
| Loglikelihood unrestricted model (H1)    | -3533.634      |
| Number of free parameters                | 32             |
| Akaike (AIC)                             | 7229.310       |
| Bayesian (BIC)                           | 7324.344       |
| Sample-size adjusted Bayesian (BIC)      | 7223.088       |
| Root Mean Square Error of Approximation: |                |
| RMSEA                                    | 0.068          |
| 90 Percent Confidence Interval           | 0.043 0.091    |
| P-value RMSEA <= 0.05                    | 0.110          |
| Standardized Root Mean Square Residual:  |                |
| SRMR                                     | 0.060          |

|                 | Estimate | Std.Err | z-value | P(> z )  |
|-----------------|----------|---------|---------|----------|
| satisfaction =~ |          |         |         | 50% Sand |
| satis1          | 1.000    |         |         |          |
| satis2          | 0.937    | 0.055   | 17.145  | 0.000    |
| satis3          | 0.997    | 0.048   | 20.896  | 0.000    |
| satis4          | 1.071    | 0.046   | 23.429  | 0.000    |
| engagement =~   |          |         |         |          |
| engage2         | 1.000    |         |         |          |
| engage3         | 0.968    | 0.039   | 24.795  | 0.000    |
| engage4         | 1.010    | 0.044   | 23.041  | 0.000    |
| attribution =~  |          |         |         |          |
| attribution1    | 1.000    |         |         |          |
| attribution2    | 1.210    | 0.224   | 5.398   | 0.000    |
| attribution3    | 1.623    | 0.341   | 4.765   | 0.000    |
| expertise =~    |          |         |         |          |
| control1        | 1.000    |         |         |          |
| control2        | 0.887    | 0.070   | 12.634  | 0.000    |
| control4        | 0.810    | 0.083   | 9.749   | 0.000    |



Reliability and validity through Confirmatory Factor Analysis (CFA)

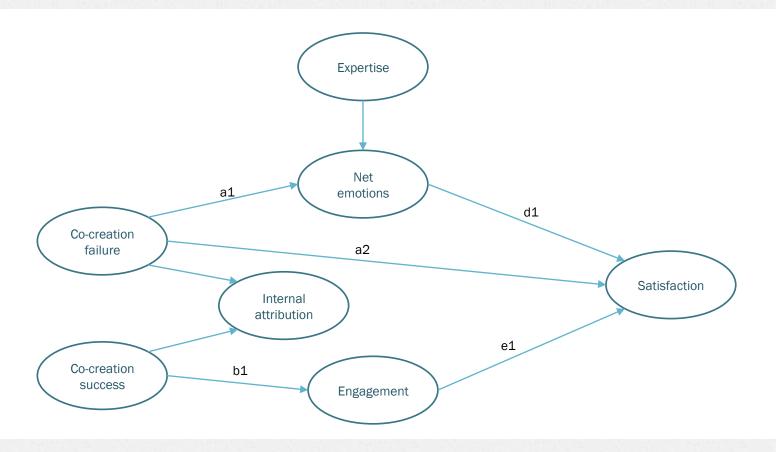
 $AVE = \frac{\sum \lambda^2}{\left[\sum \lambda^2 + \sum (1 - \lambda^2)\right]}$ 

```
# CR and AVE
                       sl <- standardizedSolution(fitCFA)</pre>
                       sl <- sl$est.std[sl$op == "=~"]
                        sl # These are the standardized factor loadings for each item
                        # summaryzed calculations
                        crsatisfaction \leftarrow sum(sl[c(1:4)])^2 / (sum(sl[c(1:4)])^2 + sum(1 - sl[c(1:4)]^2))
                        crsatisfaction
                        avesatisfaction <- sum(sl[c(1:4)]^2) / (sum(sl[c(1:4)]^2) + sum(1 - sl[c(1:4)]^2))
                        avesatisfaction
                        crengagement < sum(s[c(5:7)])^2 / (sum(s[c(5:7)])^2 + sum(1 - s[c(5:7)]^2))
CR = \frac{\left(\sum \lambda\right)^2}{\left[\left(\sum \lambda\right)^2 + \sum (1 - \lambda^2)\right]}
                        crengagement
                        aveengagement <- sum(s[c(5:7)]^2) / (sum(s[c(5:7)]^2) + sum(1 - s[c(5:7)]^2))
                        aveengagement
                        crattribution <- sum(sl[c(8:10)])^2 / (sum(sl[c(8:10)])^2 + sum(1 - sl[c(8:10)]^2))
                        crattribution
                        aveattribution <- sum(s[c(8:10)]^2) / (sum(s[c(8:10)]^2) + sum(1 - s[c(8:10)]^2))
                        aveattribution
                        crexpertise <- sum(sl[c(11:13)])^2 / (sum(sl[c(11:13)])^2 + sum(1 - sl[c(11:13)])^2))
                        crexpertise
                        aveexpertise \leftarrow sum(sl[c(11:13)]^2) / (sum(sl[c(11:13)]^2) + sum(1 - sl[c(11:13)]^2))
                        aveexpertise
```



- Estimate the model by Structural Equation Modeling (SEM) with lavaan
- Check for total effects
- Use bootstrapping to get confidence intervals for the estimations
- Set random seed so results can be reproduced
- Center the binary independent variables (Kraemer and Blasey, 2004)
- 1 0 0 -> 2/3 -1/3 -1/3

### **Study:** Model



- SEM estimation
- Including total effects

```
SEMmodelFBI <- '
# measurement
satisfaction =~ satis1 + satis2 + satis3 + satis4
engagement =~ engage2 + engage3 + engage4
attribution =~ attribution1 + attribution2 + attribution3
expertise =~ control1 + control2 + control4
diffemotions =~ diffem3
failure =~ failurecentered
success =~ successcentered
# regressions
diffemotions ~ a1*failure + expertise
satisfaction ~ a2*failure + d1*diffemotions
attribution ~ failure + success
engagement ~ a3*failure + b1*success + e1*satisfaction
# total effect of failure on satisfaction
failuresatisf := a2 + (a1*d1)
# total effect of failure on engagement
failureengage := a3 + (a1*d1*e1) + (a2*e1)
# total effect of success on engagement
successengage := b1
fitSEMFBI <- sem(SEMmodelFBI, data = mydata,</pre>
                 se = "bootstrap", bootstrap = 10000)
summary(fitSEMFBI, fit.measures = TRUE, standardized = TRUE, rsquare=TRUE,
        estimates = TRUE, ci = TRUE)
parameterEstimates(fitSEMFBI, boot.ci.type="bca.simple")
fitMeasures(fitSEMFBI)
inspect (fitSEMFBI, "cor.lv")
```





lavaan 0.6-3 ended normally after 161 iterations

Optimization method NLMINB Number of free parameters 42

Number of observations 144

Estimator ML
Model Fit Test Statistic 166.865
Degrees of freedom 94
P-value (Chi-square) 0.000

Model test baseline model:

Minimum Function Test Statistic 1906.456
Degrees of freedom 120
P-value 0.000

User model versus baseline model:

Comparative Fit Index (CFI) 0.959
Tucker-Lewis Index (TLI) 0.948

Loglikelihood and Information Criteria:

Loglikelihood user model (H0) -4105.474
Loglikelihood unrestricted model (H1) -4022.041

Number of free parameters 42
Akaike (AIC) 8294.947
Bayesian (BIC) 8419.680
Sample-size adjusted Bayesian (BIC) 8286.781

Root Mean Square Error of Approximation:

RMSEA 0.073
90 Percent Confidence Interval 0.055 0.091
P-value RMSEA <= 0.05 0.021

Standardized Root Mean Square Residual:

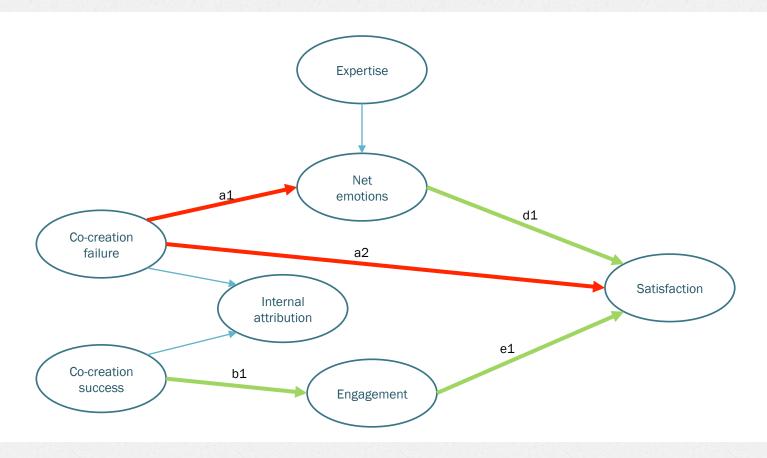
SRMR 0.073

#### : Results

|                   | Estimate | Std.Err | z-value | P(> z ) | ci.lower | ci.upper | Std.lv | Std.all |
|-------------------|----------|---------|---------|---------|----------|----------|--------|---------|
| attribution ~     |          |         |         |         |          |          |        |         |
| failure           | 1.016    | 0.344   | 2.949   | 0.003   | 0.358    | 1.707    | 0.384  | 0.38    |
| success           | 0.859    | 0.434   | 1.981   | 0.048   | 0.088    | 1.784    | 0.323  | 0.32    |
| engagement ~      |          |         |         |         |          |          |        |         |
| success (b1)      | 1.888    | 0.473   | 3.995   | 0.000   | 0.974    | 2.817    | 0.318  | 0.31    |
| netemotions ~     |          |         |         |         |          |          |        |         |
| failure (a1)      | -4.518   | 0.715   | -6.324  | 0.000   | -5.896   | -3.094   | -0.499 | -0.49   |
| expertise         | 0.396    | 0.131   | 3.020   | 0.003   | 0.129    | 0.648    | 0.226  | 0.22    |
| satisfaction ~    |          |         |         |         |          |          |        |         |
| failure (a2)      | -0.839   | 0.348   | -2.413  | 0.016   | -1.541   | -0.166   | -0.190 | -0.19   |
| engagemnt (e1)    | 0.153    | 0.069   | 2.208   | 0.027   | 0.014    | 0.286    | 0.204  | 0.20    |
| netemotns (d1)    | 0.183    | 0.049   | 3.696   | 0.000   | 0.089    | 0.284    | 0.374  | 0.37    |
|                   |          |         |         |         |          |          |        |         |
| efined Parameters |          |         |         |         |          |          |        |         |
| 2.742             | Estimate | Std.Err | z-value |         |          | ci.upper | Std.lv | Std.al  |
| failuresatisf     | -1.664   | 0.370   | -4.494  | 0.000   | -2.415   |          | -0.377 | -0.37   |
| successatisf      | 0.288    | 0.141   | 2.044   | 0.041   | 0.026    | 0.583    | 0.065  | 0.06    |



### **Study: Model**





#### Conclusion

- Co-creation can be an effective strategy to enhance consumer's engagement and satisfaction
- But co-creation with negative output favors negative emotions and less satisfaction

### **THANKS!**

**QUESTIONS?**