

Exploring Sales on Poshmark

Modeling monthly net earnings collected over three years from a boutique seller on Poshmark.



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STAT 560
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Introduction

What is Poshmark?

Poshmark is a social commerce marketplace established in 2011 where people from United States can buy and sell women's, men's and kids fashion items. The appeal of Poshmark is that you can create your own boutique and shop great deals without leaving the comfort of your own couch!

Motivation

Listing Date	Order ID	Listing Title	Department	Category	Subcategory	Brand	Color	Size	Bundle	Offer	Order Number	Net Price	Seller Price	Net Earnings	Buyer Name	Buyer State	Buyer Zip	Buyer Use Sales Tax
9/8/2015	2/5/2016	56b554d0 Pleated cc	Women	Dresses	Forever 21	S	N	N	N			\$7.00	\$0.00	\$4.05	FL	33936	diagonalley	
9/8/2015	4/7/2016	5707242a Joe FINAL	Women	Jackets & Coats	Honey Pur	Black	S	N	N	Y		\$7.00	\$0.00	\$4.05	ND	58576	fashonvictm33	
*****	*****	56cfc5b2 Joe FINAL	Women	Dresses	Christina Love	S	N	N	N			\$5.00	\$0.00	\$2.05	NY	11557	littledeer64	
*****	2/4/2016	56b39eea Black body	Women	Dresses	Long Sleeve	Black,Gray	S	N	N	N		\$9.00	\$0.00	\$6.05	NC	27705	cmnoblitt	
*****	5/2/2016	57283258 Joe FINAL	Women	Tops	Tees - Sho	Charlotte Blue	S	N	N	N		\$6.00	\$0.00	\$3.05	OK	74743	marygrom17	
*****	*****	57efb5cb Dress with	Women	Dresses	Mini	Charlotte Blue,Green	XS	Y	N	N		\$5.95	\$0.00	\$4.48	GU	96932	marysonn	
*****	*****	56ac2c3d Dress	Women	Dresses	Mini	New York Purple,W	S	N	Y	N		\$6.00	\$0.00	\$3.05	KY	41189	haleybeth79	
*****	2/2/2016	56b13a8f Dress	Women	Dresses	High Low	Mossimo Purple,Pu	XS	N	N	Y		\$12.00	\$0.00	\$9.05	NC	27104	unberlym81	
*****	*****	568ae285 Upper arm	Women	Jewelry	Bracelets	Gold	OS	N	N	Y		\$5.00	\$0.00	\$2.05	OH	43201	kyallen95	
*****	*****	56c2af61 Joe FINAL	Women	Dresses	Long Sleeve	Gray	XS	N	Y	N		\$5.00	\$0.00	\$2.05	CT	6040	mellihg	
*****	*****	56c3fe53 NWT Den	Women	Dresses	Derek Lam	Blue,Black	XS	N	N	Y		\$8.00	\$0.00	\$5.05	FL	33912	skrewbolli	
*****	*****	56d4a1f1 New York	Women	Dresses	New York	Black,Blue	XS	Y	N	N		\$7.65	\$0.00	\$6.12	GA	30346	waddiekins	
*****	3/5/2016	56db8209 NEW Mac	Women	Pants	Boot Cut	Macy's Black	O	N	Y	Y		\$7.00	\$0.00	\$4.05	PA	16046	calibrynn	
*****	5/5/2016	572b4eaa Joe FINAL	Women	Bags	Crossbody	Mossimo Green	OS	N	N	N		\$6.00	\$0.00	\$3.05	MD	20746	tesuniquediva	
*****	*****	575af790 Button-do	Women	Dresses	High Low	Pink,Red	M	N	N	N		\$6.00	\$0.00	\$3.05	VA	22041	amerna1987	
*****	2/1/2016	56af9c94 Red leath	Women	Jackets & Coats	Forever 21	Red	S	N	Y	N		\$10.00	\$0.00	\$7.05	CA	91786	jojoxblvd	
*****	*****	56a83ack Red coat	Women	Jackets & Coats	Trench Co	Macy's Red	M	N	N	N		\$17.00	\$0.00	\$13.60	TX	75110	momabear04	
*****	2/6/2016	56b6961e Dark blue	Women	Dresses	Long Sleeve	Charlotte Green,Bl	S	Y	N	N		\$9.00	\$0.00	\$6.05	CA	92260	jaimemarn1031	
*****	*****	56d4a1f1 Charlotte	Women	Dresses	Long Sleeve	Charlotte Green,Bl	S	Y	N	N		\$7.65	\$0.00	\$6.12	GA	30346	waddiekins	
*****	*****	57196f30 Joe FINAL	Women	Dresses	Long Sleeve	Nanette Li Green,Br	S	N	N	N		\$8.00	\$0.00	\$5.05	CO	80528	kolko65	
*****	5/1/2016	57263249 NWT L'am	Women	Dresses	Nanette Li	Green	XS	N	Y	Y		\$5.00	\$0.00	\$2.05	NY	10014	smbyden94	

Poshmark sales data was made available as of last winter of 2018 to each individual seller. Data such as merchandise category, subcategory, location of purchase by state, and much more are readily available in the sales report. It is left to each Poshmark seller to analyze their data independently. However, the average seller would be overwhelmed with the amount of information the data provides. As a result, most sellers do not take advantage of the data set.

For this analysis, the order date of the when the merchandise was purchased and the net earnings from the sold merchandise were the two variables of interest. The variables were aggregated by months such that monthly net earnings were observed over the months from December 2015 to March 2019. The sales report was provided by Manal Saad, a Poshmark boutique seller (closet handle @stylesforu).

The goal is to see if there is any underlying trends in monthly sales (e.g. seasonal sales). Additionally, it would be in the best interest to find models that give the client an idea of what to expect in the future sales. These models will be constructed by the Locally Estimated Scatterplot Smoothing (LOESS) method and the Thin-plate Smoothing (TPS) Spline method.

Preliminary Analysis

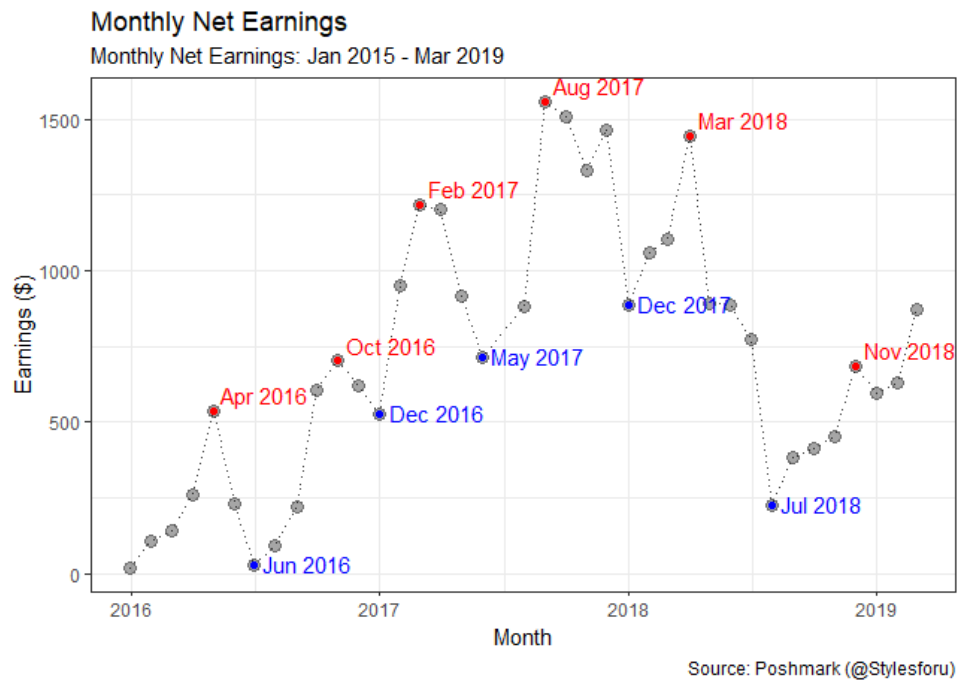


Figure 1: Monthly Net Earnings with lows and highs highlighted in blue and red, respectively.

The scatterplot of Manal's monthly net earnings were plotted against the months from January 2015 to March 2019. It can be seen that earnings were volatile; there are several extrema that showcase the buying and selling seasons. Therefore, it would be desirable to have a fitted model that captures the periodicity of monthly earnings.

Comparison Between Models

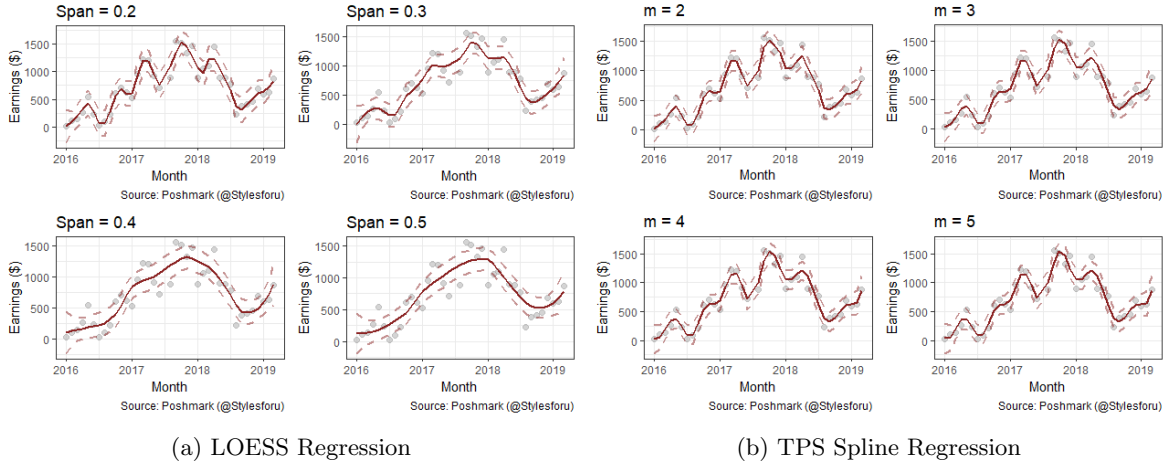


Figure 2: Comparison between the two fitted methods.

As it can be seen in Figure 2(a), the LOESS regression produced better results when quadratic functions were fitted versus linear functions. However, the optimal smoothing parameter level was more subject to interpretation. Varies levels of smoothing parameter levels either captured monthly trends (span = 0.2) or yearly trends (span = 0.5).

The different TPS spline curves are displayed in Figure 2(b). The TPS spline regression had little to no change in curvature as the degree of smoothness increased. The curvature of the TPS spline curve is very similar to the LOESS curve with a 20% smoothing parameter level.

Conclusion

It is difficult to see whether there is a underlying trend monthly sales based on the fitted curves. This can be due to the variability in the seller's sales activity. It for this reason the choice of curve used to predict future sales should be selected by an expert in merchandise sales.

Appendix A: SAS Code and Relevant Output

Here is the code to produce a LOESS regression of degree 1:

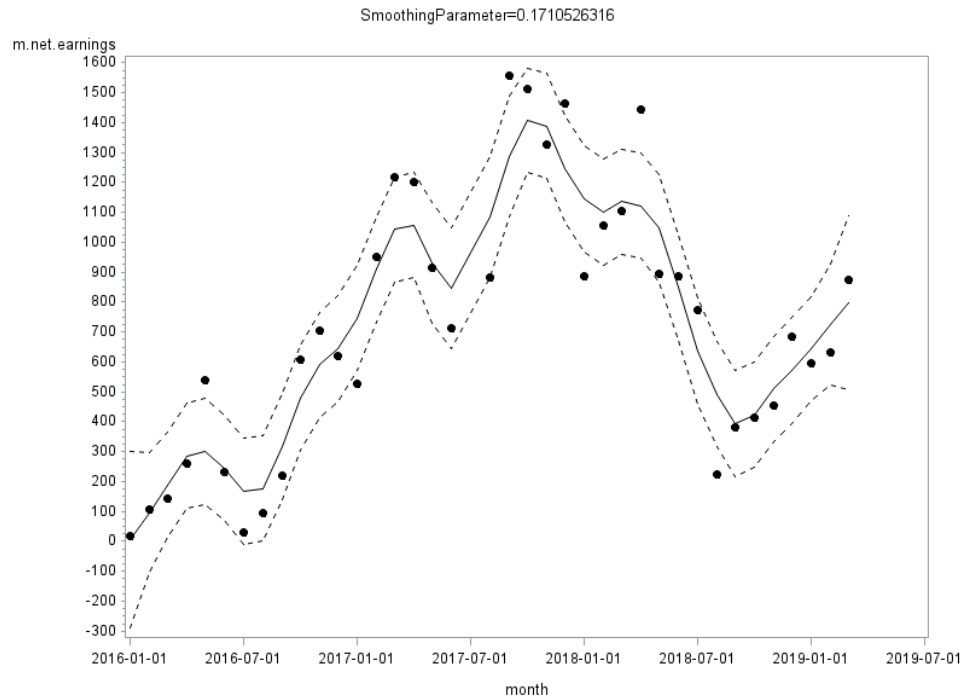
```
/* LOESS regression with degree 1 */

proc loess data=rdata;
  model m_net_earnings=month/clm alpha=0.05;
  ods output OutputStatistics=results
run;

symbol1 color=black value=dot;
symbol2 color=black value=none interpol=join line=1;
symbol3 color=black value=none interpol=join line=2;
symbol4 color=black value=none interpol=join line=2;

proc gplot data=results;
  by SmoothingParameter;
  plot (DepVar Pred LowerCL UpperCL)*month/ overlay;
run;
```

Here is the output:



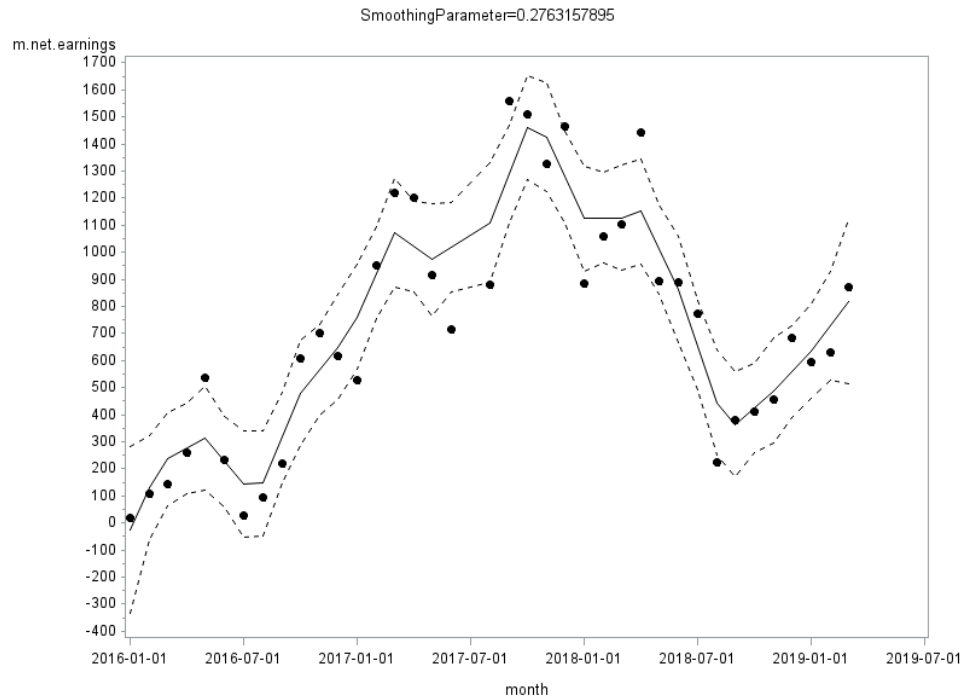
Here is the code to produce a LOESS regression of degree 2:

```
/* LOESS regression with degree 2 */

proc loess data=rdata;
  model m_net_earnings=month/degree=2 clm alpha=0.05;
  ods output OutputStatistics=results;
run;

proc gplot data=results;
  by SmoothingParameter;
  plot (DepVar Pred LowerCL UpperCL)*month/ overlay;
run;
```

Here is the output:



Here is the code to produce a LOESS regression of degree 2 with different smoothing parameters:

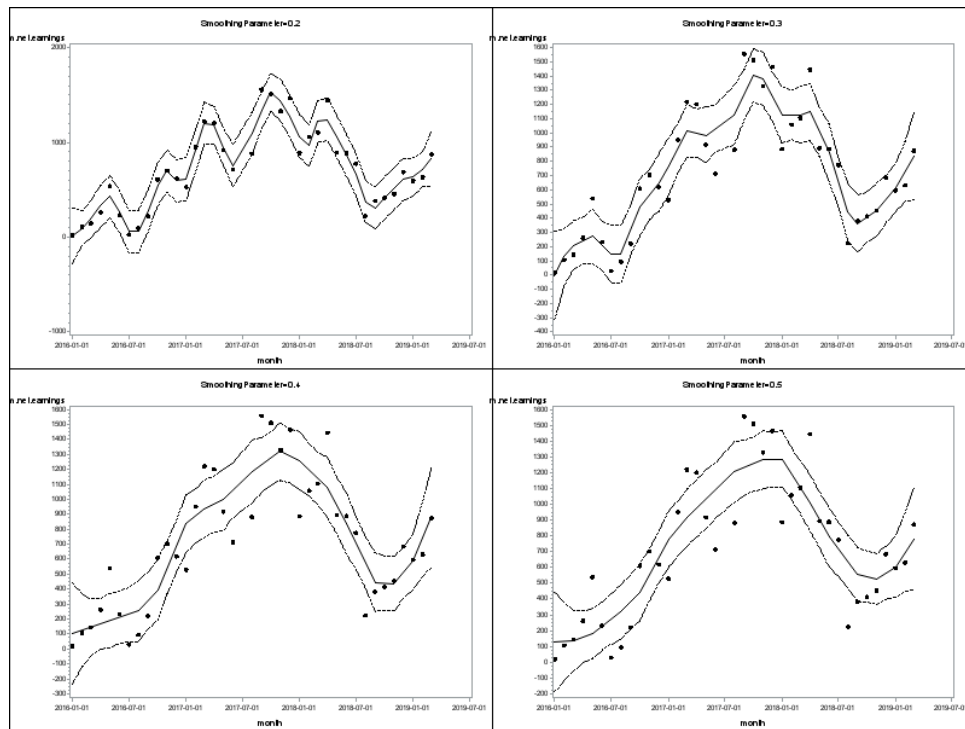
```
/* LOESS regressions (degree 2) with different smoothing parameters */

proc loess data=rdata;
  model m_net_earnings=month/degree=2 clm smooth =0.2 0.3 0.4 0.5;
  ods output OutputStatistics=results;
run;

proc gplot data=results;
  by SmoothingParameter;
  plot (DepVar Pred LowerCL UpperCL)*month/ overlay name='graph';
run;

goption display;
proc greplay nofs tc=sashelp.templt template=l2r2;
  igout gseg;
  treplay 1:graph 2:graph2 3:graph1 4:graph3;
run;
quit;
```

Here is the output:



Here is the code to produces TPS spline regressions with different degrees of smoothness:

```
/*SPLINE ESTIMATION PROCEDURE , M=2*/

proc tpspline data=rdata;
  model m_net_earnings=(month)/m=2;
  output out=result pred lclm uclm;
run;

title 'm=2';
symbol1 color=black value=dot;
symbol2 color=black value=none interpol=join line=1;
symbol3 color=black value=none interpol=join line=2;
symbol4 color=black value=none interpol=join line=2;

proc gplot data=result;
  plot (m_net_earnings p_m_net_earnings lclm_m_net_earnings uclm_m_net_earnings)*month/
    overlay name='graph2';
run;

/*SPLINE ESTIMATION PROCEDURE , M=3*/

proc tpspline data=rdata;
  model m_net_earnings=(month)/m=3;
  output out=result pred lclm uclm;
run;

title 'm=3';
proc gplot data=result;
  plot (m_net_earnings p_m_net_earnings lclm_m_net_earnings uclm_m_net_earnings)*month/
    overlay name='graph3';
run;

/*SPLINE ESTIMATION PROCEDURE , M=4*/
```

```

proc tpspline data=rdata;
  model m_net_earnings=(month)/m=4;
  output out=result pred lclm uclm;
run;

title 'm=4';
proc gplot data=result;
  plot (m_net_earnings p_m_net_earnings lclm_m_net_earnings uclm_m_net_earnings)*month/
    overlay name='graph4';
run;

/*SPLINE ESTIMATION PROCEDURE , M=5*/

proc tpspline data=rdata;
  model m_net_earnings=(month)/m=5;
  output out=result pred lclm uclm;
run;

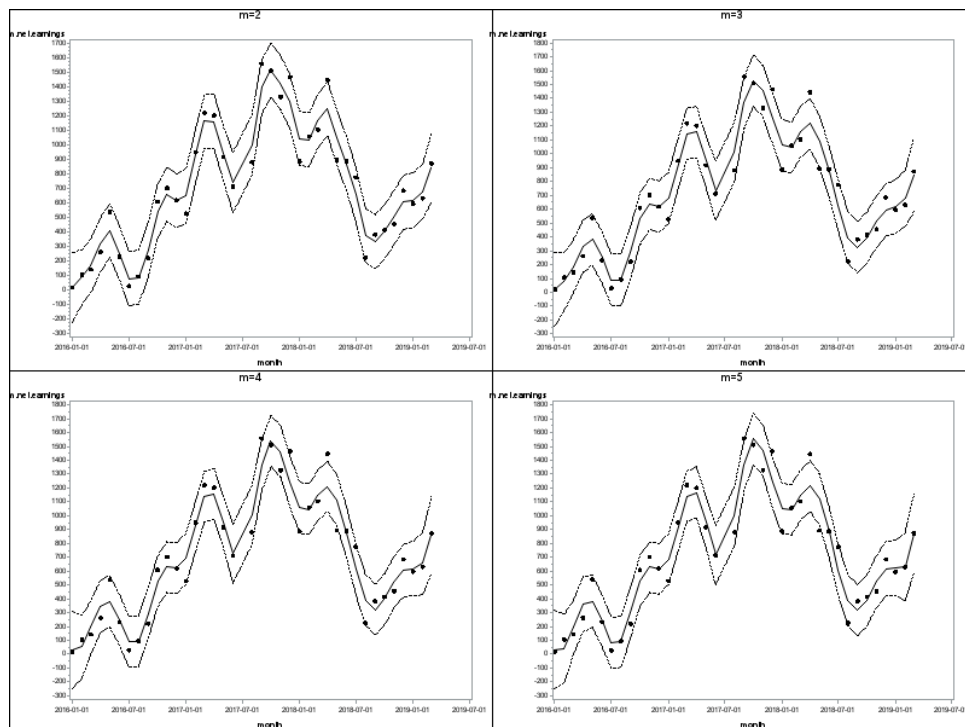
title 'm=5';
proc gplot data=result;
  plot (m_net_earnings p_m_net_earnings lclm_m_net_earnings uclm_m_net_earnings)*month/
    overlay name='graph5';
run;

/* Display the four TPS curves */

goption display;
proc greplay nofs tc=sashelp.templt template=l2r2;
  igout gseg;
  treplay 1:graph2 2:graph4 3:graph3 4:graph5;
run;

```

Here is the output:

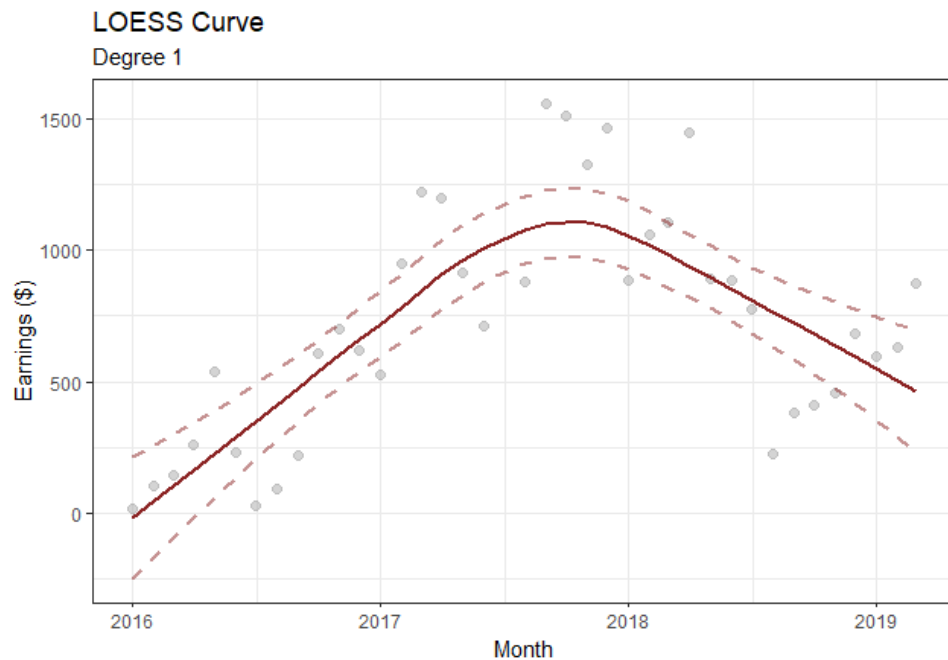


Appendix B: R Code and Relevant Output

Here is the code to produce a LOESS regression of degree 1:

```
1 ## Create LOESS fitted values
2
3 library(ggplot2)
4 library(dplyr)
5 library('fANCOVA')
6
7 p.loess1 = loess.as(dd.m$month, dd.m$m.net.earnings, degree=1, criterion = c("aicc", "gcv")[1],
8   plot = FALSE)
9 aicc_value1 = p.loess1$pars$span
10 fit1.p = predict(loess(m.net.earnings ~ as.numeric(month), degree=1, span = aicc_value1, data=dd.
11   m), se=TRUE)
12 ## Plot LOESS curves (degree 1) with 95% confidence bands
13
14 dd.m %>% mutate(smooth_1 = fit1.p$fit, LB = fit1.p$fit - qt(0.975, fit1.p$df)*fit1.p$sse, UB =
15   fit1.p$fit + qt(0.975, fit1.p$df)*fit1.p$sse) %>%
16   ggplot(aes(month, m.net.earnings)) +
17   theme_bw() +
18   labs(title = "LOESS Curve", subtitle = "Degree 1") +
19   labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
20   geom_point(size = 2, alpha = .5, color = "darkgrey") +
21   geom_line(aes(month, smooth_1), color = "brown4", size = 1) +
22   geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
23   geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
```

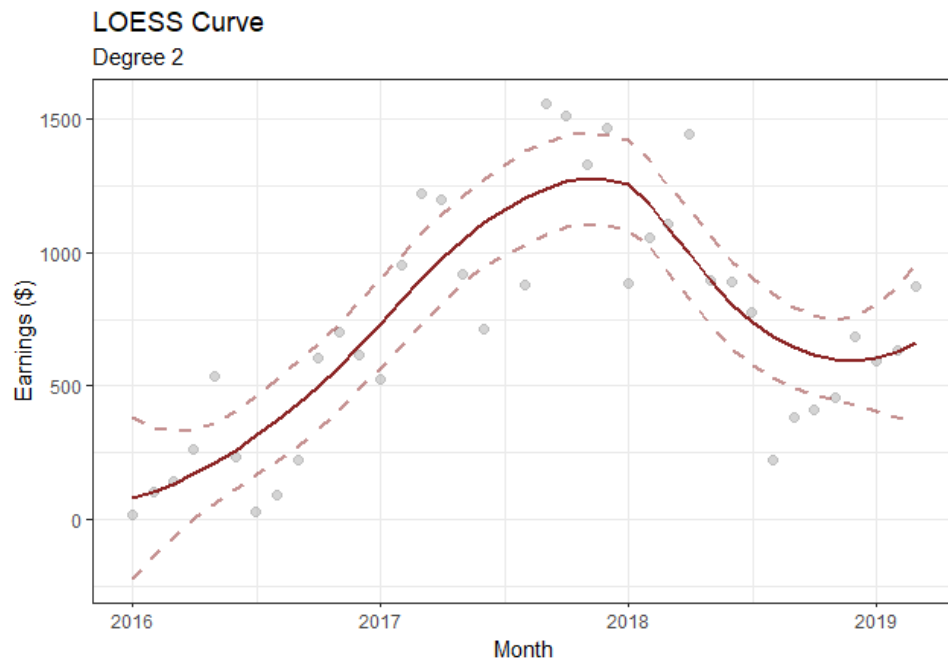
Here is the output:



Here is the code to produce a LOESS regression of degree 2:

```
1 ## Create LOESS fitted values
2
3 library(ggplot2)
4 library(dplyr)
5 library('fANCOVA')
6 p.loess2 = loess.as(dd.m$month, dd.m$m.net.earnings, degree=2, criterion = c("aicc", "gcv")[1],
7   plot = FALSE)
8 aicc_value2 = p.loess2$pars$span
9 fit2.p = predict(loess(m.net.earnings ~ as.numeric(month), degree=2, span = aicc_value2, data=dd.
10   m), se=TRUE)
11 ## Plot LOESS curves (degree 2) with 95% confidence bands
12 dd.m %>% mutate(smooth_2 = fit2.p$fit, LB = fit2.p$fit - qt(0.975, fit2.p$df)*fit2.p$sse, UB =
13   fit2.p$fit + qt(0.975, fit2.p$df)*fit2.p$sse) %>%
14   ggplot(aes(month, m.net.earnings)) +
15   theme_bw() +
16   labs(title = "LOESS Curve", subtitle = "Degree 2") +
17   labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
18   geom_point(size = 2, alpha = .5, color = "darkgrey") +
19   geom_line(aes(month, smooth_2), color = "brown4", size = 1) +
20   geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
21   geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
```

Here is the output:



Here is the code to produce a LOESS regression of degree 2 with different smoothing parameters:

```

1 ## Plot LOESS curves (degree 2) with different span values
2
3
4 # span = 0.2
5
6 plx1 = predict(loess(m.net.earnings ~ as.numeric(month), degree=2, span = 0.2, data=dd.m), se=
  TRUE)
7
8 p1 = dd.m %>% mutate(smooth_2 = plx1$fit, LB = plx1$fit - qt(0.975, plx1$df)*plx1$sse, UB = plx1$
  fit + qt(0.975, plx1$df)*plx1$sse) %>%
9   ggplot(aes(month, m.net.earnings)) +
10  theme_bw() +
11  labs(title = "Span = 0.2") +
12  labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
13  geom_point(size = 2, alpha = .5, color = "darkgrey") +
14  geom_line(aes(month, smooth_2), color = "brown4", size = 1) +
15  geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
16  geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
17
18 # span = 0.3
19
20 plx2 = predict(loess(m.net.earnings ~ as.numeric(month), degree=2, span = 0.3, data=dd.m), se=
  TRUE)
21
22 p2 = dd.m %>% mutate(smooth_2 = plx2$fit, LB = plx2$fit - qt(0.975, plx2$df)*plx2$sse, UB = plx2$
  fit + qt(0.975, plx2$df)*plx2$sse) %>%
23   ggplot(aes(month, m.net.earnings)) +
24   theme_bw() +
25   labs(title = "Span = 0.3") +
26   labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
27   geom_point(size = 2, alpha = .5, color = "darkgrey") +
28   geom_line(aes(month, smooth_2), color = "brown4", size = 1) +
29   geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
30   geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
31
32 # span = 0.4
33
34 plx3 = predict(loess(m.net.earnings ~ as.numeric(month), degree=2, span = 0.4, data=dd.m), se=

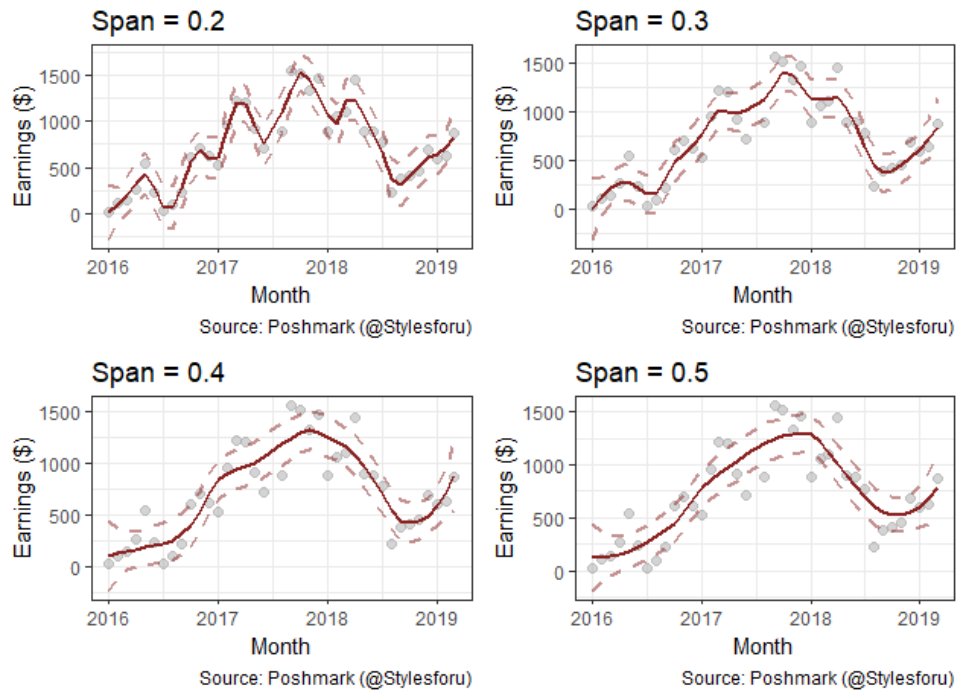
```

```

35 TRUE)
36 p3 = dd.m %>% mutate(smooth_2 = plx3$fit, LB = plx3$fit - qt(0.975, plx3$df)*plx3$se, UB = plx3$
    fit + qt(0.975, plx3$df)*plx3$se) %>%
37 ggplot(aes(month, m.net.earnings)) +
38 theme_bw() +
39 labs(title = "Span = 0.4") +
40 labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
41 geom_point(size = 2, alpha = .5, color = "darkgrey") +
42 geom_line(aes(month, smooth_2), color = "brown4", size = 1) +
43 geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
44 geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
45
46
47 # span = 0.5
48
49 plx4 = predict(loess(m.net.earnings ~ as.numeric(month), degree=2, span = 0.5, data=dd.m), se=
    TRUE)
50
51 p4 = dd.m %>% mutate(smooth_2 = plx4$fit, LB = plx4$fit - qt(0.975, plx4$df)*plx4$se, UB = plx4$
    fit + qt(0.975, plx4$df)*plx4$se) %>%
52 ggplot(aes(month, m.net.earnings)) +
53 theme_bw() +
54 labs(title = "Span = 0.5") +
55 labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
56 geom_point(size = 2, alpha = .5, color = "darkgrey") +
57 geom_line(aes(month, smooth_2), color = "brown4", size = 1) +
58 geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
59 geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
60
61
62 multiplot(p1, p3, p2, p4, cols = 2) # Plots the four graphs into one plane

```

Here is the output:



Here is the code to produces TPS spline regressions with different degrees of smoothness:

```
1 ## Plot TPS curve with different smoothing parameters on Monthly Net Earnings against Monthly
  orders
2
3 library(ggplot2)
4 library(dplyr)
5 library(fields)
6
7
8 # m = 2
9
10 earnings.tps1 = Tps(as.numeric(dd.m$month), dd.m$m.net.earnings, m=2, method="GCV")
11
12 earnings.sel = predictSE(earnings.tps1)
13
14 t1 = dd.m %>% mutate(smooth_1 = earnings.tps1$fitted.values,
15                     LB = earnings.tps1$fitted.values - qt(0.975, earnings.tps1$seff.df)*earnings.sel,
16                     UB = earnings.tps1$fitted.values + qt(0.975, earnings.tps1$seff.df)*earnings.sel)
17
18 ggplot(aes(month, m.net.earnings)) +
19   theme_bw() +
20   labs(title = "m = 2") +
21   geom_point(size = 2, alpha = .5, color = "darkgrey") +
22   geom_line(aes(month, smooth_1), color = "brown4", size = 1) +
23   geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
24   geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
25
26 # m = 3
27
28 earnings.tps2 = Tps(as.numeric(dd.m$month), dd.m$m.net.earnings, m=3, method="GCV")
29
30 earnings.se2 = predictSE(earnings.tps2)
31
32
33 t2 = dd.m %>% mutate(smooth_2 = earnings.tps2$fitted.values,
34                     LB = earnings.tps2$fitted.values - qt(0.975, earnings.tps2$seff.df)*earnings.
35                     se2,
36                     UB = earnings.tps2$fitted.values + qt(0.975, earnings.tps2$seff.df)*earnings.
37                     se2) %>%
38
39 ggplot(aes(month, m.net.earnings)) +
40   theme_bw() +
41   labs(title = "m = 3") +
42   labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
43   geom_point(size = 2, alpha = .5, color = "darkgrey") +
44   geom_line(aes(month, smooth_2), color = "brown4", size = 1) +
45   geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
46   geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
47
48 # m = 4
49
50 earnings.tps3 = Tps(as.numeric(dd.m$month), dd.m$m.net.earnings, m=4, method="GCV")
51
52 earnings.se3 = predictSE(earnings.tps3)
53
54
55 t3 = dd.m %>% mutate(smooth_3 = earnings.tps3$fitted.values,
56                     LB = earnings.tps3$fitted.values - qt(0.975, earnings.tps3$seff.df)*earnings.
57                     se3,
58                     UB = earnings.tps3$fitted.values + qt(0.975, earnings.tps3$seff.df)*earnings.
59                     se3) %>%
60
61 ggplot(aes(month, m.net.earnings)) +
62   theme_bw() +
63   labs(title = "m = 4") +
64   labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Stylesforu)") +
65   geom_point(size = 2, alpha = .5, color = "darkgrey") +
66   geom_line(aes(month, smooth_3), color = "brown4", size = 1) +
67   geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
68   geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
```

```

63
64 # m = 5
65
66 earnings.tps4 = Tps(as.numeric(dd.m$month), dd.m$m.net.earnings, m=5, method="GCV")
67
68 earnings.se4 = predictSE(earnings.tps4)
69
70
71 t4 = dd.m %>% mutate(smooth_4 = earnings.tps4$fitted.values,
72                      LB = earnings.tps4$fitted.values - qt(0.975, earnings.tps4$seff.df)*earnings.
73                      se4,
74                      UB = earnings.tps4$fitted.values + qt(0.975, earnings.tps4$seff.df)*earnings.
75                      se4) %>%
76
77 ggplot(aes(month, m.net.earnings)) +
78 theme_bw() +
79 labs(title = "m = 5") +
80 labs(x = "Month", y = "Earnings ($)", caption="Source: Poshmark (@Styleforum)") +
81 geom_point(size = 2, alpha = .5, color = "darkgrey") +
82 geom_line(aes(month, smooth_4), color = "brown4", size = 1) +
83 geom_line(aes(month, LB), color = "brown4", size = 1, linetype = "dashed", alpha = .5) +
84 geom_line(aes(month, UB), color = "brown4", size = 1, linetype = "dashed", alpha = .5)
85
86 multiplot(t1, t3, t2, t4, cols = 2) # Plots the four graphs into one plane

```

Here is the output:

