# Project in applied econometrics Report

# Lucas Javaudin, Robin Le Huérou-Kérisel, Rémi Moreau

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#### Abstract

This project has aimed at reproducing Moretti's 2011 paper on social learning effects in movie sales with R. We also blabla. Main results:

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## 1 Intuitions and detailed presentation of the model

- 1.1 Some intuitions
- 1.2 Presentation of the model

bonjour je m'appelle Rémi

### 2 Analysis and main results

#### 2.1 Identification of the surprises

Surprises consist in the residuals of the regression of the log-number of sales in the first week on the log-number of screens available (opened by theaters). This definition of surprises holds because we suppose that theaters are profit-maximizing agents and make use of all the available information to predict the success of a movie. If this definition is correct, we should expect log-number of screens opened by theaters first week to be a good indicator of knowledge available on the movie quality before it is released. In the Table 1 we reproduce Moretti's regression of log\_sales\_first\_we on log\_screens\_first\_week. Each column is the result of the regression when we control with some variables (film genre, rating available, cost, distributor, weekday, month, week, year). The fact that adding control variables doesn't change the robustness of the regression proves Moretti's point which is that theaters take into account these factors when deciding their number of available screens.

Table 1: Regression of first-weekend sales on number of screens

		$Dependent\ variable:$						
			log	_sales_first	_we			
-	(1)	(2)	(3)	(4)	(5)	(6)	(7)	
log_screens_first_week	0.893*** (0.004)	0.896*** (0.005)	0.883*** (0.005)	0.871*** (0.005)	0.803*** (0.006)	0.806*** (0.006)	0.813*** (0.006)	
$R^2$	0.907	0.909	0.910	0.912	0.932	0.936	0.938	
Adjusted $R^2$	0.907	0.908	0.910	0.912	0.928	0.931	0.933	

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

In fact, the aters perform more than what we could do using all the data available in the data set. Regressing first week-end sales on our control variables gives us a  $\mathbb{R}^2$  of .7, which is smaller than .9 performed by the aters only.

Table 2: Regression of first-weekend sales on control variables

	Dependent variable:
	log_sales_first_we
Observations	4,992
$\mathbb{R}^2$	0.699
Adjusted R <sup>2</sup>	0.674
Note:	*p<0.1: **p<0.05: ***p<0.01

We have performed the same kind of regression on France data from 2004 to 2008 and find quite similar results (Table 3 for France data and Table 4 for Paris data only<sup>1</sup>).

FRANCE

PARIS

is this useful?

nope, you only need the significa tivity of the coefficient

<sup>&</sup>lt;sup>1</sup>Data available for Paris are richer of 600 movies than France.

Table 3: Regression of first-week entries on number of screens for France

		$Dependent\ variable:$						
			1	og_entree_t	fr			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	
log_seance_fr	1.208*** (0.009)	1.237*** (0.010)	1.237*** (0.010)	1.279*** (0.014)	1.282*** (0.014)	1.287*** (0.014)	1.196*** (0.014)	
Observations	2,046	2,046	2,046	2,046	2,046	2,046	2,046	
$\mathbb{R}^2$	0.893	0.899	0.900	0.917	0.924	0.925	0.943	
Adjusted R <sup>2</sup>	0.893	0.898	0.898	0.910	0.915	0.916	0.935	

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table 4: Regression of first-week entries on number of screens for Paris only

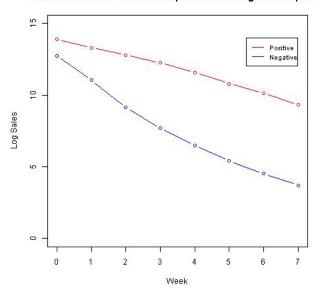
	$Dependent\ variable:$						
			log	g_entree_pa	aris		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
log_seance_paris	1.342*** (0.010)	1.336*** (0.011)	1.337*** (0.011)	1.281*** (0.014)	1.281*** (0.014)	1.284*** (0.014)	1.152*** (0.014)
Observations R <sup>2</sup>	2,701 0.875	2,701 0.880	2,701 0.881	2,701 0.901	2,701 0.908	2,701 0.909	2,701 0.927
Adjusted $\mathbb{R}^2$	0.875	0.879	0.880	0.892	0.897	0.898	0.918

Note:

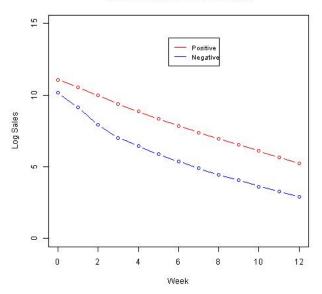
\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Figure 1: We find the same graph as Moretti

#### Decline in sale for movies with positive and negative surprises



#### Decline in sales for french data



#### Decline in sales for Paris data only

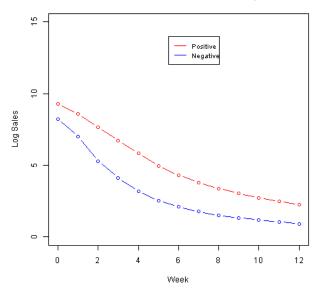


Table 5: Decline in box-office sales by opening week surprise

		Dependen	t variable:	
		$\log\_$	sales	
	(1)	(2)	(3)	(4)
t	$-0.952^{***}$ $(0.007)$	$-0.952^{***}$ $(0.006)$	$-1.289^{***}$ $(0.009)$	
t:surprise		0.475*** (0.009)		
$t:positive\_surprise$			0.640*** (0.013)	
$I(t *bottom\_surprise)$				$-1.353^{***}$ $(0.011)$
$I(t * middle\_surprise)$				$-1.011^{***}$ (0.011)
$I(t *top\_surprise)$				$-0.491^{***}$ $(0.011)$
Observations $R^2$ Adjusted $R^2$	39,936 0.772 0.739	39,936 0.788 0.758	39,936 0.787 0.756	39,936 0.790 0.760
Note:			<0.1; **p<0.0!	

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Table 6: Precision of the prior

	Depende	nt variable:
	$\log_{-}$	_sales
	(1)	(2)
t	-1.291***	$-1.267^{***}$
	(0.010)	(0.087)
t:positive_surprise	0.654***	-0.061
	(0.013)	(0.121)
t:sequel	0.037	
	(0.038)	
t:positive_surpriseTRUE:sequel	-0.225***	
. – .	(0.053)	
t:var_surprise		-0.045
-		(0.174)
t:positive_surpriseTRUE:var_surprise		1.416***
. – . – .		(0.243)
Observations	39,936	39,936
$\mathbb{R}^2$	0.787	0.787
Adjusted $R^2$	0.756	0.757
Note:	*p<0.1; **p<	0.05; ***p<0.01

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Table 7: Decline in box-office sales by opening week surprise

		Dependen	t variable:	
		log_en	tree_fr	
	(1)	(2)	(3)	(4)
t	$-0.526^{***}$ $(0.002)$	$-0.526^{***}$ $(0.002)$	$-0.571^{***}$ $(0.003)$	
t:surprise		0.076*** (0.004)		
t:positive_surprise			0.087*** (0.004)	
$t: bottom\_surpriseFALSE$				$-0.459^{***}$ $(0.004)$
$t:bottom\_surprise$				$-0.574^{***}$ $(0.004)$
$t:middle\_surprise$				-0.088*** $(0.005)$
Observations	26,598	26,598	26,598	26,598
$\mathbb{R}^2$	0.851	0.853	0.853	0.854
Adjusted R <sup>2</sup>	0.838	0.841	0.841	0.841

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table 8: Precision of the prior

	De	pendent varia	ble:
		log_entree_fr	•
	(1)	(2)	(3)
t	-0.570***	-0.698***	-0.678***
	(0.003)	(0.013)	(0.004)
t:positive_surprise	0.105***	0.109***	0.009
	(0.005)	(0.018)	(0.006)
t:saga	-0.027		
	(0.016)		
t:positive_surpriseTRUE:saga	-0.145***		
	(0.019)		
t:var_surprise		0.370***	
_ :		(0.035)	
t:positive_surpriseTRUE:var_surprise		-0.062	
· · · · · · · · · · · · · · · · · · ·		(0.050)	
t:art essai			0.259***
			(0.006)
t:positive_surpriseTRUE:art_essai			0.066***
. – . –			(0.008)
Observations	26,598	26,546	26,598
$\mathbb{R}^2$	0.855	0.854	0.880
Adjusted R <sup>2</sup>	0.843	0.842	0.870
Note:	*p<	<0.1; **p<0.05	5; ***p<0.01

Table 9: Decline in box-office sales by opening week surprise

		Dependen	t variable:	
		$\log_{entr}$	ree_paris	
	(1)	(2)	(3)	(4)
t	$-0.583^{***}$ $(0.002)$	$-0.583^{***}$ $(0.002)$	$-0.564^{***}$ $(0.003)$	
t:surprise		$-0.032^{***}$ $(0.004)$		
t:positive_surprise			$-0.039^{***}$ $(0.005)$	
$t: bottom\_surpriseFALSE$				$-0.594^{***}$ $(0.004)$
$t:bottom\_surprise$				$-0.541^{***}$ (0.004)
$t: middle\_surprise$				$-0.021^{***}$ (0.006)
Observations	35,113	35,113	35,113	35,113
R <sup>2</sup>	0.810	0.810	0.810	0.811
Adjusted R <sup>2</sup>	0.794	0.794	0.794	0.795
Note:		*p<	<0.1; **p<0.05	5; ***p<0.01

Table 10: Precision of the prior

	<i>De</i>	pendent varia	ble:
	lo	og_entree_par	ris
	(1)	(2)	(3)
t	$-0.560^{***}$ $(0.003)$	$-0.772^{***}$ $(0.017)$	$-0.616^{***}$ $(0.005)$
t:positive_surprise	$-0.030^{***}$ $(0.005)$	$-0.213^{***}$ $(0.024)$	
t:saga	$-0.118^{***}$ $(0.017)$		
$t:positive\_surpriseTRUE:saga$	-0.022 (0.020)		
t:var_surprise		0.576*** (0.045)	
$t:positive\_surpriseTRUE:var\_surprise$		0.480*** (0.065)	
t:art_essai			0.087*** (0.006)
$t:positive\_surpriseTRUE: art\_essai$			0.156*** (0.009)
Observations	35,113	35,074	35,113
$\mathbb{R}^2$	0.811	0.814	0.819
Adjusted R <sup>2</sup>	0.795	0.798	0.804

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

2.2 Precision of the prior

#### 2.3 Size of the Social Network

Consumers with a larger social network receive more feedbacks from their peers and thus they are able to evaluate more precisely the quality of the movie.

Table 11: Precision of peers' signal

	Dependent variable: log_entree_fr		
	(1)	(2)	
$\overline{t}$	-0.663***	-0.451***	
	(0.007)	(0.005)	
$t \times \text{positive surprise}$	0.061***	0.076***	
	(0.010)	(0.006)	
$t \times \text{tout public}$	0.115***		
· · · · · · · · ·	(0.008)		
$t \times positive\_surprise \times tout\_public$	0.031***		
	(0.011)		
$t \times \text{seance}$ fr first week		-0.033***	
		(0.001)	
$t \times \text{positive\_surprise} \times \text{seance\_fr\_first\_week}$		0.011***	
. – . – –		(0.001)	
Observations	26,598	26,598	
$\mathbb{R}^2$	0.856	0.867	
Adjusted R <sup>2</sup>	0.844	0.856	
Note:	*p<0.1; **p<	(0.05; ***p<0.01	

## 2.4 Does Learning Decline Over Time?

Table 12: Convexity of the sales profile

	$Dependent\ variable:$
	log_entree_fr
$\overline{t}$	-0.978***
	(0.011)
$t^2$	0.034***
	(0.001)
$t \times \text{positive\_surprise}$	0.393***
	(0.016)
$t^2 \times \text{positive\_surprise}$	-0.026***
1	(0.001)
Observations	26,598
$R^2$	0.861
Adjusted R <sup>2</sup>	0.850
Note:	*p<0.1; **p<0.05; ***p<0.01

# 3 Conclusion: some comments