

Banco de México
Documentos de Investigación

Banco de México
Working Papers

N° 2018-03

The Effect of Uncertainty on Foreign Direct
Investment: the Case of Mexico

Gabriela López Noria
Banco de México

Juan José Zamudio Fernández
Banco de México

February 2018

La serie de Documentos de Investigación del Banco de México divulga resultados preliminares de trabajos de investigación económica realizados en el Banco de México con la finalidad de propiciar el intercambio y debate de ideas. El contenido de los Documentos de Investigación, así como las conclusiones que de ellos se derivan, son responsabilidad exclusiva de los autores y no reflejan necesariamente las del Banco de México.

The Working Papers series of Banco de México disseminates preliminary results of economic research conducted at Banco de México in order to promote the exchange and debate of ideas. The views and conclusions presented in the Working Papers are exclusively the responsibility of the authors and do not necessarily reflect those of Banco de México.

The Effect of Uncertainty on Foreign Direct Investment: the Case of Mexico*

Gabriela López Noria[†]
Banco de México

Juan José Zamudio Fernández[‡]
Banco de México

Abstract: This paper investigates the effect of uncertainty on FDI flows into the Mexican manufacturing sector during the period 2007-2015. Using a panel of manufacturing subsectors, we estimate a model by System GMM that includes domestic and external factors, as well as idiosyncratic (i.e. that affect manufacturing subsectors in a particular way) and aggregate (i.e. that affect all manufacturing subsectors in general) uncertainty measures as explicative variables. We also perform some simulations to quantify the effect of uncertainty on FDI flows. The main results show that uncertainty discourages FDI flows into the Mexican manufacturing sector. We also find that the idiosyncratic uncertainty measures are more important in explaining FDI flows than the aggregate uncertainty measures, with the exception of the global risk aversion index.

Keywords: Uncertainty, Foreign Direct Investment, Expectations, Manufacturing

JEL Classification: D80, F21, D84, L60

Resumen: Este documento investiga el efecto de la incertidumbre sobre los flujos de IED hacia el sector manufacturero mexicano durante el periodo 2007-2015. Usando un panel de subsectores manufactureros, estimamos un modelo por GMM en Sistema que incluye factores domésticos y externos, así como medidas de incertidumbre idiosincrásicas (i.e. que afectan de forma particular a los subsectores manufactureros) y agregadas (i.e. que afectan de forma general a todos los subsectores manufactureros), como variables explicativas. También llevamos a cabo simulaciones para cuantificar el efecto de la incertidumbre sobre los flujos de IED. Los resultados muestran que la incertidumbre desalienta a los flujos de IED hacia el sector manufacturero mexicano. Asimismo, encontramos que las medidas de incertidumbre idiosincrásicas son más importantes para explicar los flujos de IED que las medidas de incertidumbre agregadas, con la excepción del índice global de aversión al riesgo.

Palabras Clave: Incertidumbre, Inversión Extranjera Directa, Expectativas, Manufactura

*We would like to thank Dr. Daniel I. Chiquiar Cikurel and Dr. Alejandrina Salcedo Cisneros for their helpful comments and suggestions.

[†] Dirección de Medición Económica, Dirección General de Investigación Económica, Banco de México. Email: glopezn@banxico.org.mx.

[‡] Dirección de Medición Económica, Dirección General de Investigación Económica, Banco de México. Email: juan.zamudio@banxico.org.mx.

I. Introduction

The uncertainty-investment relationship has been a core research topic among economists for several decades now. However, its relevance increased following the 2008 global financial crisis (Denis and Kannan, (2013)), when leading economists and monetary authorities from all around the world started to openly express that uncertainty about the global economic outlook was fostering a “wait and see” attitude among economic agents and, therefore, contributing to a decline in spending projects, particularly investment expenditures. The following are some quotes that reflect this situation:

“Uncertainty is largely behind the dramatic collapse in demand. Given the uncertainty, why build a new plant or introduce a new product now? Better to pause until the smoke clears” (Olivier Blanchard, in Bloom (2013)).

“A dark cloud of uncertainty is looming over global growth, particularly around weakening emerging markets and the outcome of the EU referendum, which is chilling some firms’ plans to invest. At present, the economic signals are mixed – we are in an unusually uncertain period” (Caroline Fairbairn, in CBI Press Team (2016)).

"Given the uncertainty in markets, many of our clients are stepping away from trading. Even in wealth management, the uncertain environment means that a lot of the clients sit on cash and avoid long-term decisions and really are not very active and that's a very difficult model to be in....It's a difficult environment still. This year there is huge uncertainty in global markets, major political decisions to be taken in the US (with the presidential election in November) and U.K. (with the referendum on European Union membership), so I think a lot of people are waiting for that dust to settle. So that gives me hope that the second half of the year will see more activity (from clients)...but at the moment the indicators are still that the market is difficult" (UBS Chairman Alex Weber, in Ellyatt and Cutmore (2016)).

“FDI recovery continues along a bumpy road. Particularly of concern is the sharp drop-off in manufacturing investment projects, which play such an important role in

generating badly needed productivity improvements in developing economies. Looking ahead, economic fundamentals point to a potential increase in FDI flows by around 10% in 2017. However, significant uncertainties about the shape of future economic policy developments could hamper FDI in the short term” (United Nations Conference on Trade and Development, UNCTAD, Secretary-General Mukhisa Kituyi, in UNCTAD (2017a)).

The purpose of this paper is to provide some empirical evidence on the effect of uncertainty on foreign direct investment (FDI) flows into the Mexican manufacturing sector. We use a panel of manufacturing sectors with quarterly data over the period 2007 – 2015. Our econometric analysis follows Ghosal and Loungani (1996) in the sense that we estimate an econometric specification where FDI is the dependent variable and, its lag, a proxy for uncertainty and manufacturing cash flow are the main independent variables.¹ However, we augment this econometric specification by adding domestic and external factors that we consider may also encourage or hamper FDI flows into Mexico. These additional independent variables are the following: an insecurity index for Mexico, an interest rate on Mexico’s inflation indexed bonds, the peso United States (US) dollar real exchange rate, Mexico’s exports/GDP ratio, the US industrial production index and the US Federal Funds rate.

The contributions of this paper to the empirical literature on the uncertainty-investment link are three-fold. First, it analyzes the effect of uncertainty on FDI flows into the Mexican manufacturing sector, rather than on fixed investment as most empirical studies on this relationship do (e.g. Episcopos (1995); Ghosal and Loungani (1996, 2000), Bloom et al. (2007)).² Varella-Mollick et al. (2006) and Jordaan (2012) analyze the determinants of FDI flows into Mexico, but they do not focus on the relationship between uncertainty and FDI. Varella-Mollick et al. (2006) study the impact of market size, infrastructure, agglomeration economies, government expenditure on infrastructure, real wages, and labor

¹ Ghosal and Loungani (1996) use sales or cash flow as proxies for Tobin’s q in their analysis. See Section III.2 for more details on these regressors.

² An exception is Lemi and Asefa (2001) who analyze the effect of uncertainty on FDI flows for a sample of African countries.

unions on FDI flows into Mexican states over the period 1994-2001; while Jordaan (2012) analyzes the effect of regional demand, production costs, agglomeration of manufacturing and services, and regional distance to Mexico City and the US on FDI flows into Mexico during the period 1994-1999.

Second, in order to investigate the uncertainty-investment relationship we build some uncertainty measures based on entrepreneurs'/forecasters' expectations about individual manufacturing subsectors' and/or Mexico's economic situation, rather than on the volatility of stock market returns as is usually the practice in the empirical literature (i.e. Episcopos (1995); Leahy and Whited (1996), Bloom et al. (2007), Valencia Herrera and Gándara Martínez (2009), Sharifi-Renani and Mirfatah (2012), among others). To our knowledge, this is the first paper on the uncertainty-investment relationship for the case of Mexico that measures uncertainty based on individuals' expectations. Nonetheless, we also evaluate the importance of uncertainty measures that are more related to the economic and political situation of Mexico and/or of developed countries and, that we consider may affect all manufacturing subsectors in general.

Third, we perform some simulations in order to quantify the effect of uncertainty on FDI flows into the Mexican manufacturing sector. To our knowledge, Bloom et al. (2007) for the case of the United Kingdom is the only previous paper that has performed some simulations to track and quantify the investment response to an uncertainty shock.

Our main findings show that an increase in uncertainty discourages FDI flows into the Mexican manufacturing sector. These results are in line with the theoretical and empirical literature on irreversible investment. The findings also reveal that the idiosyncratic (sectorial) uncertainty measures are more important in explaining FDI flows into the Mexican manufacturing sector than aggregate uncertainty measures, which affect all subsectors in general, with the exception of the global risk aversion index. The econometric simulations show that if uncertainty, proxied for example by the disagreement regarding the firm's economic situation in 12 months time at the subsector level, had been lower by 1 percentage point for each and every quarter during the period 2010 – 2015, the Mexican manufacturing sector would have received, on average, an additional 714 million dollars FDI flows (or an additional 4.6 percent of actual average FDI inflows) per year over the period 2010 - 2015.

The remainder of the paper is organized as follows: Section II reviews the theoretical and empirical literature on the effect of uncertainty on investment; Section III describes the data, the econometric specification, and the estimated results; Section IV presents some simulations in order to quantify the effect of uncertainty on FDI flows; while Section V concludes.

II. Literature Review

Despite the attention the uncertainty-investment link has received over the years, no theoretical consensus has yet emerged on its sign.

On one hand, Hartman (1972, 1976) and Abel (1983, 1984, 1985) show that output price uncertainty raises a competitive and risk-neutral firm's investment if the marginal profitability of capital is convex in prices.³

On the other hand, Pindyck (1988, 1991) and Dixit and Pindyck (1994) show that if an investment is irreversible (i.e. "sunk costs that cannot be recovered should market conditions change adversely" (Caballero and Pindyck (1992))), uncertainty on future demand reduces a risk-neutral firm's investment. An irreversible investment implies not only an investment expenditure but also an opportunity cost since the investor gives up the opportunity of waiting for new information about market conditions to arrive and therefore to make a productive investment. Under irreversibility, the Net Present Value rule becomes invalid. Instead, firms invest if the net present value of investment less the opportunity cost of irreversibility is greater or equal to zero. Therefore, if uncertainty increases, the opportunity cost or "option to defer" as is usually called in the literature also increases, and the likelihood of an economic agent making an investment decreases.

This theoretical puzzle on the uncertainty-investment relationship has prompted the emergence of a considerable amount of empirical studies on this topic. Nonetheless, the majority of them has given support to the negative link. This empirical consensus has been

³ If the marginal profitability of capital is convex in prices, a mean preserving increase in the variance of prices raises the expected return of a marginal unit of capital and, therefore, investment becomes more attractive (Carruth et al. (2000)). See Walter Oi (1961) for more details.

reached by using either aggregate (Ferderer (1993), Episcopos (1995), and Price (1996)) or disaggregated data (Leahy and Whited (1996), Guiso and Parigi (1999), Fuss and Vermeulen (2004), and Bloom et al. (2007)) and, different approaches to measure uncertainty.

The empirical literature has emphasized some advantages of using disaggregated data: 1) it permits to use uncertainty measures that are more related to the idiosyncratic factors which affect industries or firms (Leahy and Whited (1996); Kalckreuth (2000)); and 2) it permits to control for endogeneity problems between uncertainty and investment, as well as for industry/firm heterogeneity, if panel data is available to the researcher (Carruth et al. 2000). In this paper, we use disaggregated data at a manufacturing subsector level, which has enabled us to control for these problems.

As regards the uncertainty measures, there are three main approaches used in the empirical literature to proxy uncertainty.⁴ The most common approach consists on obtaining the volatility of stock prices/returns, output prices, commodity prices and/or exchange rates to proxy uncertainty. For instance, Leahy and Whited (1996) use a panel of 600 US manufacturing firms to show that uncertainty, measured as the variance of firms' daily stock returns, has a negative effect on investment/capital stock over the period 1982 - 1987. Ghosal and Loungani (1996) study the uncertainty-investment relationship using data on 4-digit US manufacturing industries over the period 1958 – 1989. They split their sample into highly and low concentrated industries, based on each industry's four-firm seller concentration ratio in order to investigate if differences across industries affect this link. For each industry, they estimate price equations to obtain uncertainty measures as the standard deviation of the residuals from those regressions. Their main results show that uncertainty depresses investment for low concentrated industries. Bloom et al. (2007) focus on traded British manufacturing firms over the period 1972 - 1991 to analyze their investment response to demand shocks under uncertainty. In this case, uncertainty is proxied as the volatility of the manufacturing firms' daily stock returns and, investment is measured as gross industry investment to the beginning of the period capital stock. The authors give evidence of a non-

⁴ For the case of the US economy, Ferderer (1993) measures uncertainty based on the risk premium included in the term structure of interest rates on long-term bonds. This approach is particularly different from those described in this literature review.

linear effect of real sales growth on investment and, of a weaker effect of sales growth on investment when firms face higher levels of uncertainty. Sharifi-Renani and Mirfatah (2012) investigate the determinants of FDI flows into Iran over the period 1980-2006 and, show that an increase in exchange rate volatility discourages FDI inflows.

A second approach is based on ARCH or GARCH estimates of the conditional variance of prices and/or other type of aggregates (i.e. manufacturing output, exchange rates, wages, among others) to measure uncertainty. Using ARCH estimates of conditional variances of prices, real interest rates, aggregate personal consumption expenditure, stock prices and a composite index of 11 leading indicators, Episcopos (1995) finds that uncertainty depresses fixed private investment for the case of the US during the period 1948 - 1993. Lemi and Asefa (2001) study the impact of uncertainty indicators and of political instability on FDI flows from all source countries into a sample of host countries in Africa, over the period 1987 – 1999, and on US FDI flows to that same sample of host countries in Africa, over the period 1989-1998. Uncertainty is measured as the unconditional standard deviation of the inflation rate and the real exchange rate and/or, as the conditional variance of these two variables, generated from (GARCH) models.⁵ The specification, which is estimated by Tobit random effects and by fixed effects, also controls for the investors' confidence, labor force availability, domestic market size, export sector size, cost of capital, debt burden, among others. Their main results show that the effect of uncertainty indicators on FDI flows from all source countries and from the US as a total is not statistically significant. Political instability and government policy commitment are the only factors having an impact on FDI flows from the US manufacturing sector, whereas inflation and real exchange rate uncertainties have an impact on FDI flows from the US non-manufacturing sector “when they are coupled with political instability and debt burden”.

A third approach relies on data provided by surveys on individual forecasters' expectations about the economic climate for investment decisions and/or on entrepreneurs' expectations about the future demand for their firms' product or output price changes. For the case of Ghanaian manufacturing firms over the period 1994 - 1995, Patillo (1998) tests

⁵ The unconditional standard deviation is obtained as the standard deviation of the monthly series of the inflation rate and/or of the exchange rate for each year.

and finds evidence for the following predictions: 1) firms only invest when the marginal revenue product of capital reaches a hurdle level; 2) uncertainty, which is proxied based on the “entrepreneur’s subjective probability distribution over future demand for the firm’s products”, increases the trigger to a greater extent for firms with irreversible investment and; 3) uncertainty depresses investment (measured as investment in plants and equipment in year t relative to the value of plants and equipment in $t-1$) more severely on firms with irreversible investment. Guiso and Parigi (1999) use the 1993 Italian Survey on Investment in Manufacturing to compute an uncertainty measure based on each entrepreneur’s “subjective probability distribution of future demand for the firm’s product”. They mainly find that uncertainty has a bigger impact on investment (measured as fixed investment in structures, machinery and equipment, and vehicles to capital stock) when firms with more market power and with more irreversible investment are considered. Similarly, Fuss and Vermeulen (2004) study the impact of demand and price uncertainty on investment plans (period 1987 -2000) and realized investments (period 1987 - 1999) for a sample of Belgian manufacturing firms. They use entrepreneurs’ expectations about the future demand for the firm’s products and output price changes, obtained from the Belgian Business Cycle Survey, to build their uncertainty measures.⁶ Their main results show that demand uncertainty has a negative and a statistically significant effect on planned and realized investment, while price uncertainty does not. Finally, Bond et al. (2005) explores the uncertainty-investment relationship for a sample of non-financial British firms over the period 1987 - 2000 using uncertainty proxies based either on stock market returns and/or on monthly analysts’ earnings forecasts. Their findings show that uncertainty is negatively associated to investment and they are robust to the inclusion of other control variables such as sales growth, Tobin’s Q and cash flow.

Valencia-Herrera and Gándara Martínez (2009) is the only paper that, to our knowledge, has investigated the uncertainty-investment relationship for the case of Mexico. The authors focus on a group of 104 firms listed in the Mexican Stock Exchange during the

⁶ In particular, they use the answers to the following two questions: 1) “Do you expect demand for your product, in the next three months (A) to rise, (B) to remain unchanged, (C) to decrease, with respect to its average level at that time of the year?” and 2) “Do you expect the price of your product, in the next three months, (A) to rise, (B) to remain unchanged, (C) to decrease?” to calculate their uncertainty measures as: $[(\%up + \%down) - (\%up - \%down)^2]$.

period 1997 - 2007. They study the effect of the volatility of daily stock market returns on the ratio of net total assets (or net fixed assets) plus amortizations and depreciation to net heritage (or total asset), which is their measure of investment. The analysis is performed by considering large, medium size, and small firms. Their main findings show a negative relationship between uncertainty and investment for the medium size and small firms, while a positive relationship for the large firms.

This paper, which also focuses on the Mexican case, distinguishes from the majority of the empirical literature on the uncertainty-investment link in three main ways. First, it analyzes the effect of uncertainty on FDI flows, rather than on fixed investment. As it is well known, FDI plays a significant role in promoting economic growth and in permitting access to production technology, innovative managerial practices, financial resources, etc. Second, this paper relies on *INEGI's* Monthly Survey of Business Opinion (*INEGI (2017a)*) and on *Banco de México's* Survey of Professional Forecasters (*Banco de México (2017a)*) to build uncertainty measures based on entrepreneurs'/forecasters' expectations about individual manufacturing subsectors' and or Mexico's economic situation, rather than on the volatility of stock market returns. The volatility of stock market returns, which is the most common uncertainty measure used in the empirical literature, may capture several sources of risk, but it may also respond to "extraneous information, reflect irrational behavior and the presence of noise traders, or be dominated by speculative bubbles and subsequent crashes rather than by changes in the firm's fundamentals" (Guiso and Parigi (1999)).⁷ In contrast, the uncertainty measures we used in this paper are closely related to the microeconomic behavior of the manufacturing subsectors analyzed and/or to the idiosyncratic factors that affects them. Nonetheless, we also evaluate the importance of uncertainty measures that are more related to the economic and political situation of Mexico and/or of developed countries. Furthermore, by using directly observable entrepreneurs'/forecasters' expectations, rather than an expectations-formation model to derive them, we are able to avoid measurement problems in our main estimated specification. Third, we perform some simulations in order to quantify the effects of uncertainty on FDI flows. To our knowledge, Bloom et al. (2007)

⁷ Ferderer (1993) and Bloom et al. (2007) also mention that the volatility of stock market returns may be driven by non-fundamental factors, while Fuss and Vermeulen (2004) support the view that this measure may be noisy.

is the only previous paper that has performed some simulations in order to track and quantify the response of investment to an uncertainty shock.

III. Empirical Analysis

III.1 Foreign Direct Investment and Uncertainty Measures

This Section describes our two main variables of interest: FDI and uncertainty.

During the period 2007 - 2015, the Mexican manufacturing sector, which on average accounted for 16.6 percent of Gross Domestic Product (GDP) in these years, attracted an annual average of 46.2 percent of total FDI flows.⁸ In contrast, the services sector, which on average accounted for 59.7 percent of GDP in this same period, attracted an annual average of 38.7 percent of total FDI flows.⁹ The US was responsible for most of these FDI flows into the Mexican manufacturing sector, except for two years: in 2010, the Dutch company Heineken bought the brewery *FEMSA* (*FEMSA* (2017)), and in 2013 the Belgian company Anheuser-Busch InBev acquired the brewery *Modelo* (*Anheuser-Busch InBev* (2017)). See Figures 1a and 1b for more details.¹⁰

⁸ According to the United Nations Conference on Trade and Development (UNCTAD (2017b)), Mexico was ranked position 19 by the FDI flows it received in 2007, while position 16, in 2015. Within Latin America and the Caribbean, Mexico was the second largest recipient of FDI flows both in 2007 and 2015, after Brazil.

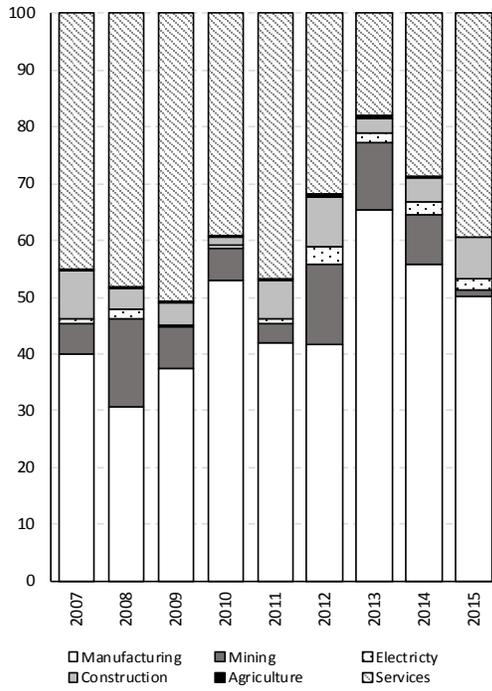
⁹ For more details see the following website belonging to *Secretaría de Economía*:

<<https://www.gob.mx/se/acciones-y-programas/competitividad-y-normatividad-inversion-extranjera-directa?state=published>>

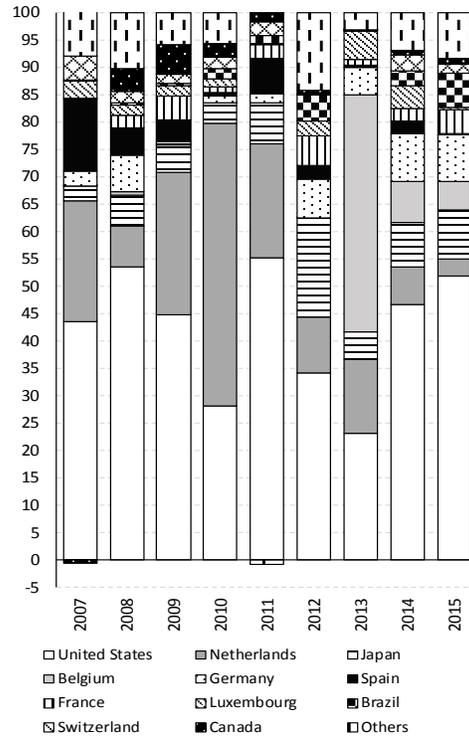
¹⁰ For more details see the following websites belonging to *FEMSA* and Anheuser-Busch InBev, respectively: <<http://ir.femsa.com/mx/results.cfm>> and <<http://www.ab-inbev.com/investors/reports-and-filings.html>>

Figure 1. FDI Flows into the Mexican Manufacturing Sector

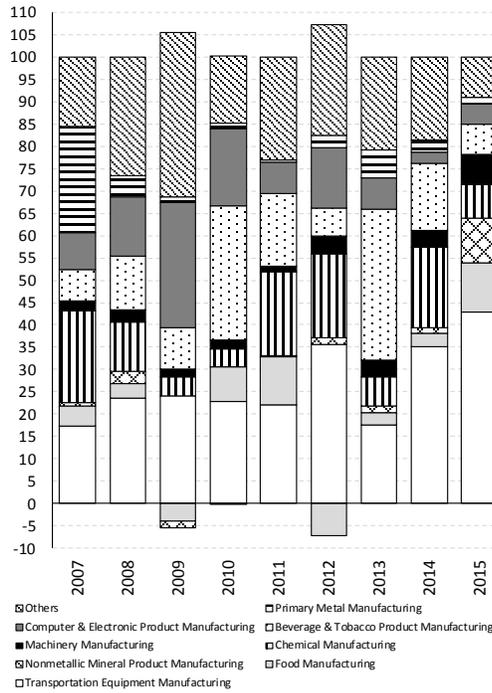
a) By Sector Share



b) By Source Country



c) By manufacturing industry



FDI flows into the Mexican manufacturing sector have been heterogeneous. In 2007, most FDI flows went into the primary metal manufacturing sector, followed by the chemical manufacturing sector and the transportation equipment manufacturing sector. In 2010 and 2013, the beverages and tobacco product manufacturing sector was the biggest recipient of FDI flows, as mentioned before. However, in 2012 and 2015 the transportation equipment manufacturing sector attracted 35.7 percent and 42.8 percent of the FDI flows, respectively. See Figure 1c.

In this paper, we employ data on net FDI flows at a 3-digit manufacturing sector level. These data come from *Banco de México* and cover the period 01/2007 - 04/2015. They include new investments, credits given by the headquarters to their related firms or branches, and the reinvestment of utilities. These data are quarterly and we expressed them in 2010 million pesos.

We also use two surveys to build three uncertainty measures that we consider closely related to the idiosyncratic factors that affect the manufacturing subsectors in Mexico: *INEGI's* Monthly Survey of Business Opinion (*INEGI* (2017a)) and *Banco de México's* Survey of Professional Forecasters (*Banco de México* (2017a)).

From *INEGI's* Survey, we focus on the following questions to build our first two measures of uncertainty:

- 1) How would you expect the economic situation of your firm to be in 12 months time relative to the current situation?
- 2) How would you expect the economic situation of your country to be in 12 months time relative to the current situation?

The survey participants answer these questions by choosing only one of the following qualitative options: “much better”, “better”, “the same”, “worst” or “much worst”. Since these data per survey respondent is not public, *INEGI* provided us, for each question and for each manufacturing subsector and quarter analyzed, with the percentage distribution of manufacturing subsectors that answered these qualitative options. Therefore, we calculated our first two measures of uncertainty, which from now on will be called *unc 1: Firm's economic situation in 12 months time* and, *unc 2: Country's economic situation in 12 months time*, as the standard deviation of the percentage distribution of the answers to the qualitative

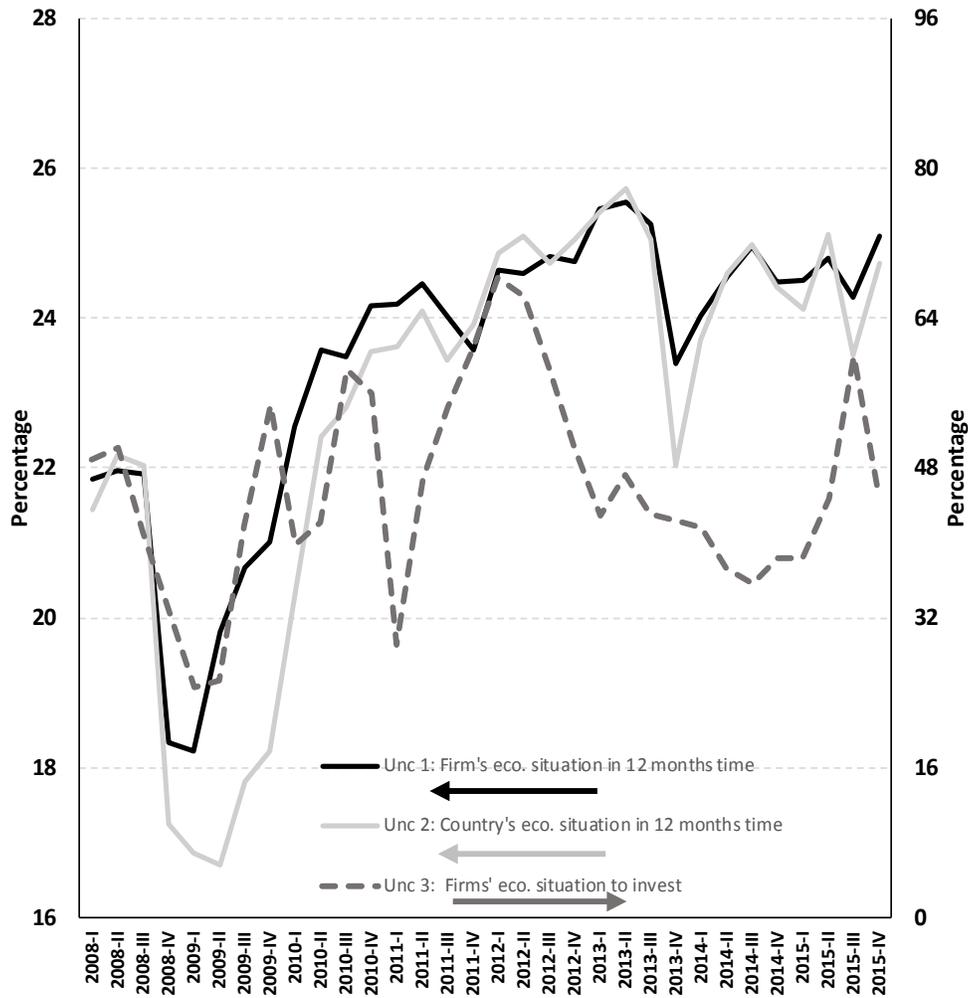
options mentioned previously (“much better”, “better”, “the same”, “worst” or “much worst”). Hence, uncertainty increases when there is more disagreement about the economic situation in a given manufacturing sector. These variables are available from 01/2008 to 04/2015.

From *Banco de México*'s Survey, we focus on the following question to derive our third measure of uncertainty:

- 1) How would you evaluate the current economic climate for firm investment decisions?

The answers to this question can only be “good moment”; “bad moment” or “I am not sure”, so the uncertainty measure, which varies across time but not across subsectors, is built with the percentage of forecasters that answered “I am not sure”. For this variable we focus on the period 01/2007 to 04/2015 and we will refer to it from now on as *unc3: Firms' eco. situation to invest*. Figure 2 shows the evolution of these three uncertainty measures.

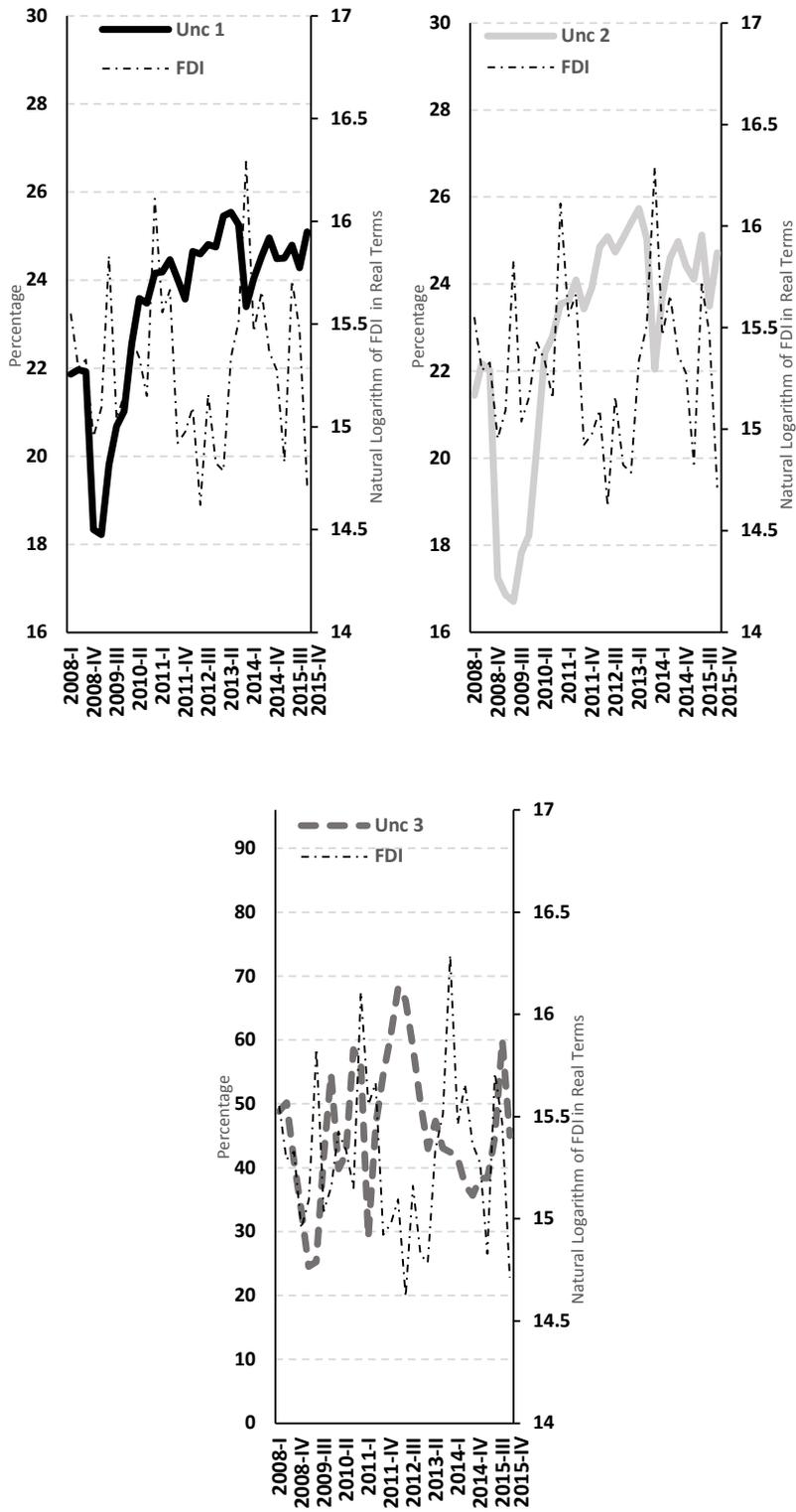
Figure 2. Evolution of the Idiosyncratic Uncertainty Measures



As it can be seen, the three idiosyncratic uncertainty measures we built are correlated among them for most of the period under analysis and regardless of the Survey being employed.

Figure 3 shows the three idiosyncratic uncertainty measures and the FDI variable together. As it can be seen, the uncertainty proxies and the FDI variable present an opposite trend during most of the analyzed period.

Figure 3. Idiosyncratic Uncertainty Measures vs. FDI



III.2 Econometric Specification

Following Ghosal and Loungani (1996), we estimate an econometric specification where FDI is the dependent variable and, its lag, a proxy for uncertainty and, manufacturing cash flow are the main independent variables. However, we also include some domestic and external factors in our model, since we consider they might encourage or hamper FDI flows into Mexico. These additional independent variables are the following: an insecurity index for Mexico, an interest rate on Mexico's inflation indexed bonds, the Peso - US dollar real exchange rate, a Mexican exports to GDP ratio, the US industrial production index and, the US Federal Funds Rate.

The estimated specification can be written as follows:

$$\begin{aligned} FDI_{i,t} = & \beta_0 + \beta_1 FDI_{i,t-1} + \beta_2 Uncertainty_{i,t} + \beta_3 Cash Flow_{i,t-3} + \\ & \beta_4 Insecurity_t + \beta_5 Interest Rate_t + \beta_6 Exchange Rate_t + \\ & \beta_7 Exports/GDP_t + \beta_8 US Industrial Production_t + \\ & \beta_9 US Federal Funds Rate_t + \mu_i + \varepsilon_{i,t} \dots (1) \end{aligned}$$

Where:

i, t – These are sub-indexes for manufacturing subsectors and time, respectively.

$FDI_{i,t-1}$ – The lagged dependent variable indicates we are estimating a dynamic model. The data on FDI flows are obtained from *Secretaría de Economía* (2017).

$Uncertainty_{i,t}$ – This variable stands for the uncertainty measures we built based on *INEGI's* (2017a) and/or *Banco de México's* (2017a) surveys, in other words:¹¹

- *unc 1: Firm's eco. situation in 12 months time,*
- *unc 2: Country's eco. situation in 12 months time; and*
- *unc 3: Firms' eco. situation to invest*

$Cash Flow_{i,t-3}$ – It is measured as manufacturing sales minus wages, at a 3 digit manufacturing level. The data on this variable are obtained from *INEGI* (2017b). The reason for introducing cash flow into the specification is the following. Investment expenditure depends on Tobin's q , the ratio of the market value of an additional unit of capital to its

¹¹ The uncertainty measure in equation (1) has sub-indexes i and t . Nonetheless, we should clarify that the two uncertainty measures derived from *INEGI's* (2017a) survey do vary across subsectors and time; while that derived from *Banco de México's* (2017a) survey only varies across time.

replacement cost. Investment increases if Tobin's q is greater than 1 and decreases if it is less than 1 (Ferderer (1993)). Since Tobin's q or marginal q is not observable, researchers approximate it using average q , which is the ratio of the market value of the firm to the replacement cost of its assets. We do not introduce Tobin's q in the estimated specification for two main reasons: 1) our data is at an industry level so we cannot build this variable and, 2) following Ghosal and Loungani (1996), Tobin's q does not "out-perform simpler measures such as sales or cash flow". Hence, we include cash flow as an independent variable in our specification.¹² This variable is generally used as a proxy for future demand growth or profitability and we expect it to have a positive effect on FDI flows into Mexico.

Insecurity_t – We proxy insecurity, at a national level, as the sum of robberies, homicides and kidnappings in Mexico with data from *Secretariado Ejecutivo del Sistema Nacional de Seguridad Pública* (2016) (*Secretariadoejecutivo.gob.mx* (2016)). Criminal activity is generally perceived as a risk to an individual's safety, so we expect that higher insecurity levels in Mexico will reduce FDI flows into the manufacturing sector.

We also introduce the *Interest Rate_t* and the *Exchange Rate_t* in our specification since they are prices economic agents consider when they are taking investment decisions.

Exports/GDP_t – We use the ratio of exports to GDP as a proxy for trade openness in Mexico since we expect that higher trade openness might induce more FDI flows into Mexico.

The data source for the domestic real interest rate and the real exchange rate is *Banco de México* (2017b and 2017c, respectively), while that for the exports to GDP ratio is *INEGI* (2017c).

US Industrial Production_t and *US Federal Funds Rate_t* – We incorporated these "external" factors into our specification since we consider that an economic boom

¹² We include the third lag of this variable for the following reasons: 1) we consider it might take some time for cash flow to have an effect on foreign investors' decisions; 2) we use quarterly data; and 3) we use the general to specific approach to decide which lags of the cash flow variable to include in the estimated specification. This approach consists on first estimating a "general" model that, for this specific document, includes a maximum number of lags of the variable cash flow. That "general" model is then reduced by eliminating (again for this specific document) those lags of the cash flow variable that are not statistically significant. In this sense, by following this approach, we just included the third lag of the cash flow variable in the estimated specification since it had the appropriate sign and it was statistically significant.

and/or a tightening of the monetary policy in the US might deter FDI flows into Mexico.¹³ These variables were obtained from the US Federal Reserve System (2016) and from the Federal Reserve Bank of St. Louis (2016), respectively.

μ_i – This term controls for unobserved time-invariant industry specific effects.

$\varepsilon_{i,t}$ – It represents the specification’s error term.

All the variables in the model are expressed in logarithms, except for the uncertainty variables, the interest rate on Mexico’s inflation indexed bonds, the exports to GDP ratio, and the US Federal Funds Rate, which are expressed in percentages.¹⁴ Table 1 presents some summary statistics on the variables described before.

Table 1. Summary Statistics

<i>Variable</i>	<i>Mean</i>	<i>Std. Dev.</i>	<i>Min</i>	<i>Max</i>
Unc 1: Firm's eco. situation in 12 months time (percentage points)	23.5	2.2	15.3	28.4
Unc 2: Country's eco. situation in 12 months time (percentage points)	23.0	2.9	13.7	27.5
Unc 3: Firms' eco. situation to invest (percentage points)	46.0	10.2	24.5	68.2
FDI (2010 millions of pesos)	194	670	0	14,213
Cash flow (2010 millions of pesos)	150	201	4	1,090
Insecurity ¹ (number of crimes)	172,563	21,771	133,769	206,788
Interest rate (%)	1.9	1.0	0.5	3.8
Exports/GDP (%)	2.5	0.3	1.9	3.1
Real Exchange Rate (peso-US dollar)	11.8	0.9	10.5	14.0
US Industrial Production (index)	95.6	4.9	83.9	101.4
US Federal Funds Rate (%)	0.9	1.6	0.1	5.3

1/ Note: Calculated as the sum of robberies, homicides and kidnappings in Mexico.

¹³ See Calvo et al. (1996) and Ying and Kim (2001) for more details on the domestic and external factors that affect capital and investment flows.

¹⁴ The data we used on FDI contain positive and negative values, as well as zeroes. 15.34 percent of the observations are negative values, while 0.13 percent are zeroes. We consider these percentages of non-positive FDI values to be relatively small, so we decided to present the results without making any transformation to the data (i.e. setting negative values and zeroes equal to 0.1 before taking logarithms (Blonigen and Wang (2004) and Eichengreen and Tong (2007); adding a constant to all the observations before taking logarithms or, replacing zeroes with the minimum of the log of positive values in the sample, among others). In other words, we expressed the FDI variable in logarithms and, therefore, only positive values are considered in the analysis. Nonetheless, for robustness checks, we did perform some conventional practices to deal with these inconveniences in the data, and the results remain qualitatively similar.

Equation (1) presents two main problems: 1) the lagged dependent variable and the lagged value of cash flow are correlated with the error term due to the presence of unobserved time-invariant industry specific effects, and 2) some regressors might be a function of FDI, rather than a determinant of it. The estimation of equation (1) by traditional panel data techniques that do not control for these two problems (e.g. Ordinary Least Squares) generates inconsistent and biased estimates of the unknown parameters. Therefore, we estimate equation (1) by System GMM in order to restore consistency of the parameters' estimates and to control for possible cases of endogeneity. In addition, the GMM estimator "turns out to be efficient within the class of instrumental variable estimators" (p. 115, Nucci and Possolo (2010)). Following Arellano and Bover (1995) and Blundell and Bond (1998), equation (1) is estimated by System GMM using STATA's `xtabond2` command written by Roodman (2006).

The Hansen test for over-identifying restrictions, which is a test for the exogeneity of the set of instruments included; the Arellano-Bond tests for first and second order autocorrelation in the first differenced residuals, as well as robust standard errors, to account for general forms of heteroskedasticity and serial correlation in the error term, are computed and registered together with the results.

III.3 Estimation Results

Columns (1), (2) and (3) of Table (2) show the results from estimating equation (1) without considering the proxy for trade openness and the US economy related variables, while columns (4), (5) and (6) of the same Table present the findings with this additional variables. We mainly find that the coefficient of the uncertainty variables is negative and statistically significant (except for column (1) where the coefficient is not statistically significant), which suggest that an uncertain economic environment discourages FDI flows into the Mexican manufacturing sector. These results support the existing theoretical and empirical literature on irreversible investment presented in Section II.

Even though specification (1) is not designed to analyze the causal effect between other controls and the FDI flows, we also report the interpretation of the coefficients regarding these extra variables. In particular, we find that cash flow has a positive and a statistically significant effect on FDI flows (except for column (3) where the coefficient is

not statistically significant), which suggests that foreign investors do consider variables such as future demand growth and/or profitability when deciding whether to invest or not in the Mexican manufacturing sector.

The results also show that the Mexican manufacturing sector becomes less attractive to foreign investors if insecurity levels increase. Nonetheless, these findings are not robust since this variable's coefficient is not statistically significant in the last three columns of Table 2.

The rest of the domestic and external variables included in the model do not have a statistically significant effect on FDI flows.

Table 2. Estimation Results: Idiosyncratic Uncertainty Variables

DEPENDENT VARIABLE: FDI	1	2	3	4	5	6
Unc 1: Firm's eco. situation in 12 months time	-0.0624 (0.040)			-0.1531* (0.077)		
unc 2: Country's eco. situation in 12 months time		-0.1151** (0.050)			-0.1616** (0.071)	
unc 3: Firms' eco. situation to invest			-0.0261*** (0.009)			-0.0361* (0.020)
FDI (lagged)	0.1070* (0.058)	0.0929 (0.062)	0.1492*** (0.037)	0.1406*** (0.045)	0.1004 (0.065)	0.1268 (0.088)
Cash flow (lagged)	0.8074* (0.431)	1.0061** (0.443)	0.6382 (0.460)	0.9415*** (0.322)	1.3525* (0.739)	1.1359* (0.650)
Insecurity	-5.2535** (2.112)	-4.8447* (2.559)	-6.1543*** (2.155)	-3.1482 (2.837)	-1.4933 (5.718)	-4.0114 (3.980)
Interest rate on Inflation Indexed Bonds	-0.6289** (0.240)	-0.6002** (0.243)	-0.7393*** (0.245)	-0.9591*** (0.293)	-0.6199 (0.490)	-0.2457 (0.301)
Real Exchange Rate	1.5324 (1.135)	0.4274 (1.211)	0.5536 (1.022)	0.1813 (2.444)	0.6845 (2.768)	5.4890 (3.417)
Exports/GDP				-0.0002 (0.000)	0.0001 (0.000)	0.0008 (0.001)
US Industrial Production				1.0353 (3.482)	0.5528 (4.393)	5.0833 (4.397)
US Federal Funds Rate				0.3628 (0.249)	0.2844 (0.377)	-0.1997 (0.284)
Constant	61.8253** (28.596)	57.4904* (29.594)	77.1543** (28.113)	35.2334 (38.720)	8.0808 (74.415)	4.6969 (53.505)
Observations	487	487	503	487	487	503
Number of industries	21	21	21	21	21	21
AR(1) in first differences:	-2.735	-2.802	-2.824	-2.823	-2.834	-3.016
AR(1) in first differences p-value:	0.00624	0.00508	0.00474	0.00475	0.00459	0.00256
AR(2) in first differences:	-0.524	-0.385	-1.144	-1.450	-0.478	-0.284
AR(2) in first differences p-value:	0.600	0.700	0.253	0.147	0.632	0.776
Hansen test of overid. restrictions:	8.958	8.732	9.795	6.334	10.26	4.607
Hansen test of overid. restrictions p-value:	0.707	0.726	0.938	0.957	0.743	0.708

Robust standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

III.4 Robustness Tests

We re-estimate equation (1) by incorporating additional uncertainty measures which we consider are more related to the economic and political situation of Mexico and/or of developed countries. The purpose of this second exercise is two-fold: 1) we assess which uncertainty proxies are more important in explaining FDI flows into the Mexican manufacturing sector, either the idiosyncratic uncertainty measures used before or, the new uncertainty proxies which we consider may affect all manufacturing subsectors in general, and 2) we verify for the robustness of our results. In this sense, the new specification takes the form:

$$\begin{aligned} FDI_{i,t} = & \beta_0 + \beta_1 FDI_{i,t-1} + \beta_2 \text{Idiosyncratic Uncertainty}_{i,t} + \\ & \beta_3 \text{Aggregate Uncertainty}_t + \beta_4 \text{Cash Flow}_{i,t-3} + \beta_5 \text{Insecurity}_t + \\ & \beta_6 \text{Interest Rate}_t + \beta_7 \text{Exchange Rate}_t + \beta_8 \text{Exports/GDP}_t + \\ & \beta_9 \text{US Industrial Production}_t + \beta_{10} \text{US Federal Funds Rate}_t + \mu_i + \\ & \varepsilon_{i,t} \dots (2) \end{aligned}$$

Where:

Idiosyncratic Uncertainty_{i,t} – This term stands for the uncertainty measures we used in the previous Section and, that we consider may affect individual manufacturing subsectors (i.e. *unc 1: Firm’s economic situation in 12 months time; unc 2: Country’s economic situation in 12 months time; and unc3: Firms’ eco. situation to invest*).

Aggregate Uncertainty_t – This term stands for the new uncertainty proxies, which we consider may affect all manufacturing subsectors in general. These uncertainty measures are the following:

- ***Economic and Political Uncertainty Index for Mexico:*** This uncertainty proxy is built by *Banco de México* as the ratio of news found in Bloomberg containing the words “Mexico”, “uncertainty”, “economy” and “politics”, or “*México*”, “*incertidumbre*”, “*economía*” and “*política*” to the news found in this same source related to other topics.

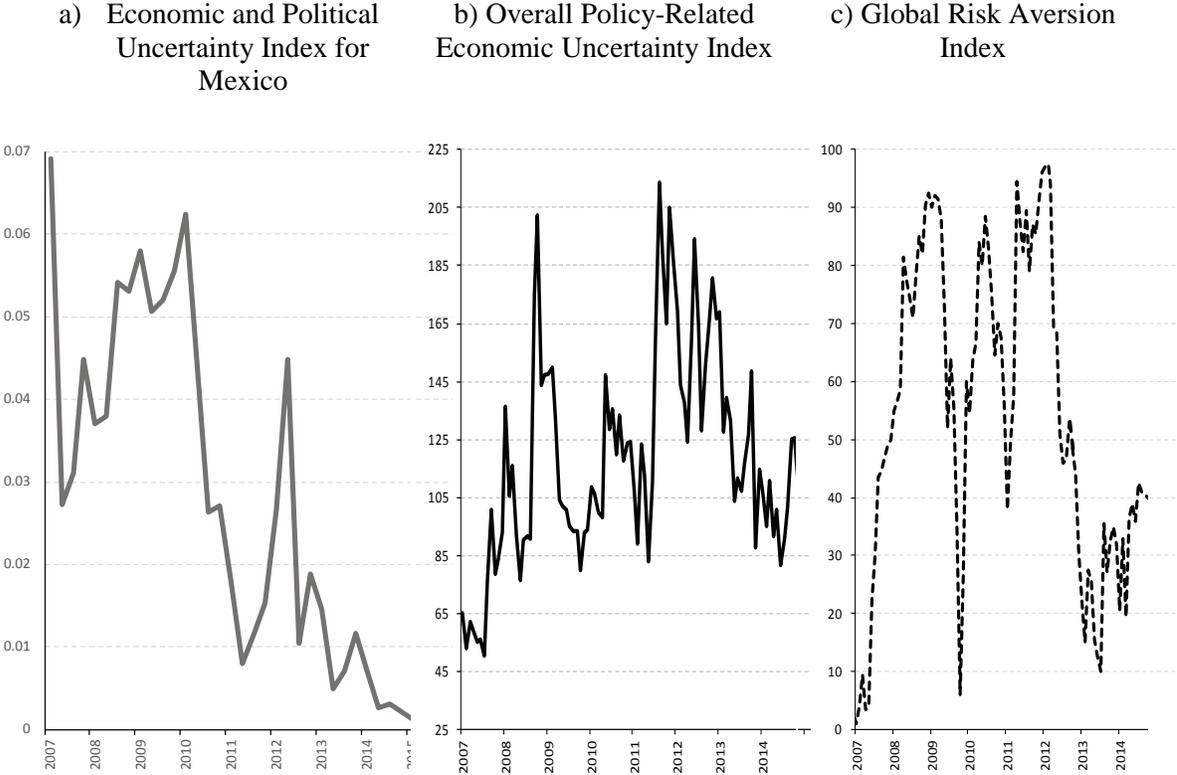
- ***Overall Policy-Related Economic Uncertainty Index:*** This index is built by Nicholas Bloom, Scott R. Baker, and Steven J. Davis from Economic Policy Uncertainty using three types of components¹⁵:
 - 1) The news coverage about policy-related economic uncertainty consists on searching for articles in newspapers that contain the terms “uncertainty” or “uncertain”; “economic” and “economy”; “congress”, “legislation”, “white house”, “regulation”, “Federal Reserve” or “deficit”, which belong to the uncertainty, the economy and policy categories.
 - 2) The tax code expiration data reflects “the number of federal tax code provisions set to expire in future years”.
 - 3) The economic forecaster disagreement is based on the Federal Reserve Bank of Philadelphia’s Survey of Professional Forecasters, particularly on the forecast of three variables which are directly influenced by the monetary and fiscal policies in the US: the consumer price index; the purchases of goods and services by state and local governments; and the purchases of goods and services by the federal government. “The dispersion in the forecasts of these variables is treated as proxies for uncertainty about monetary policy and about government purchases of goods and services at the federal level”.
- ***Global Risk Aversion Index:*** UBS publishes this index, which represents the historic percentile of a risk aversion coefficient γ of an investor’s utility function, who chooses how much of his/her resources to invest and/or to consume today. This γ is estimated by considering monthly returns on MSCI Developed Markets Index and changes on the composite Developed Markets Purchasing Managers’ Index to proxy for demand swings.

Figure 4 shows the evolution of these three uncertainty measures which we call aggregate. As we can see, the three series present a different pattern among them, and they behave different from the three idiosyncratic uncertainty measures presented in Figure 1. On

¹⁵ For more details see the following website: <http://www.policyuncertainty.com/index.html>.

this regard, it seems interesting to know how the idiosyncratic and aggregate uncertainty measures affect the dependent variable.

Figure 4. Evolution of the Aggregate Uncertainty Measures



The rest of the variables in equation (2) remain as in equation (1). Columns (1) to (9) from Table 3 show the results from this second estimation. We mainly find that the coefficient of both the idiosyncratic and the aggregate uncertainty measures in all the specifications presented are negative. However, in columns (1) to (6) the coefficient of the idiosyncratic uncertainty measures is statistically significant, while that of the aggregate uncertainty measures is not. Columns (7) to (9) show a different story since the coefficient of the global risk aversion index is statistically significant, while that of the idiosyncratic uncertainty measures is not.

Table 3. Estimation Results: Idiosyncratic and Aggregate Uncertainty Variables

DEPENDENT VARIABLE: FDI		1	2	3	4	5	6	7	8	9
Economic and Political Uncertainty Index for Mexico	Unc 4: Unc_Mex_ecopol	-18.0238 (19.264)	-16.2204 (14.162)	-2.2488 (12.435)						
	Unc 1: Firm's eco. situation in 12 months time	-0.2871*** (0.133)								
	unc 2: Country's eco. situation in 12 months time		-0.1511* (0.085)							
	unc 3:Firms' eco. situation to invest			-0.0328* (0.017)						
Overall Policy Related Economic Uncertainty for the US	Unc 5: Unc_Us				-0.3494 (1.019)	-0.0263 (0.646)	-0.7048 (0.698)			
	Unc 1: Firm's eco. situation in 12 months time				-0.3048* (0.174)					
	unc 2: Country's eco. situation in 12 months time					-0.2446* (0.138)				
	unc 3:Firms' eco. situation to invest						-0.0327* (0.017)			
Global Risk Aversion Index (Developed Economies)	Unc 6: Risk_Aversion							-0.8822** (0.398)	-0.5366 (0.602)	-0.7049** (0.298)
	Unc 1: Firm's eco. situation in 12 months time							-0.1359 (0.215)		
	unc 2: Country's eco. situation in 12 months time								-0.1053 (0.159)	
	unc 3:Firms' eco. situation to invest									-0.0167 (0.018)
Controls ^{1/}	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
Observations	487	487	503	487	487	503	487	487	487	503
Number of industries	21	21	21	21	21	21	21	21	21	21
AR(1) in first differences:	-2.706	-2.826	-2.453	-2.668	-2.860	-2.170	-2.703	-2.791	-2.851	
AR(1) in first differences p-value:	0.0068	0.0047	0.0142	0.0076	0.0042	0.0300	0.0069	0.0053	0.0044	
AR(2) in first differences:	-1.634	-0.421	-0.548	-1.506	-1.608	-0.297	-1.233	-0.567	-0.624	
AR(2) in first differences p-value:	0.102	0.674	0.583	0.132	0.108	0.766	0.217	0.571	0.532	
Hansen test of overid. restrictions:	5.521	11.960	7.045	13.910	8.753	12.720	10.970	10.210	8.005	
Hansen test of overid. restrictions p-value:	1	1	1	0.998	1	1	1	0.994	0.999	

Robust standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

Note: 1/Controls include the following variables : FDI (lagged), Cash flow (lagged), Insecurity, Interest rate on Inflation Indexed Bonds, Real Exchange Rate, Exports/GDP, US Industrial Production and US Federal Funds Rate.

Overall, the results suggest that the idiosyncratic uncertainty measures seem to be more important in explaining FDI flows into the Mexican manufacturing sector than aggregate uncertainty measures, which affect all manufacturing subsectors in general, with the exception of the global risk aversion index.

IV. Quantifying the Results

The previous exercises show evidence of how uncertainty may discourage FDI flows into the Mexican manufacturing sector. In order to assess the size of this negative effect, we conduct some simulations using specifications 4, 5 and 6 of Table 2 (one simulation per idiosyncratic uncertainty measure), under the assumption that uncertainty for each and every quarter during the period 01/2010 – 04/2015 is equal to its registered level minus 1 percentage point.¹⁶ In other words, we build a counterfactual scenario for the FDI flows going into the Mexican manufacturing sector during the period 01/2010 – 04/2015 by assuming that uncertainty, for each and every quarter during the period 01/2010 – 04/2015, is equal to its registered level in that period minus 1 percentage point. We then compare the counterfactual scenario with a base scenario in which the uncertainty variable takes its registered values and, the difference between both scenarios is finally attributed to the change in uncertainty.

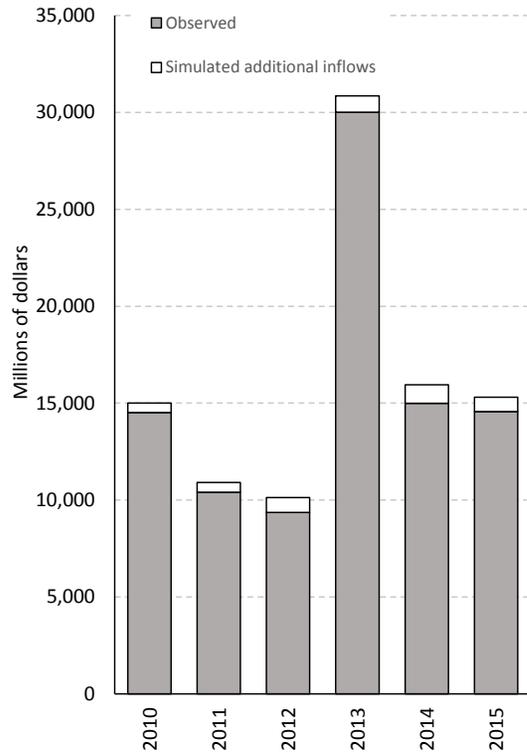
Figure 5 plots the response of FDI flows to an increase in uncertainty, while Table 4 presents the simulated average (period 2010 – 2015) FDI flows in millions of dollars and as a percentage of the actual average (period 2010 – 2015) FDI flows. The findings show that if uncertainty (proxied by unc 1: Firm's eco. situation in 12 months time, for example) for each and every quarter during the period 2010 – 2015 had been lower by 1 percentage point, the Mexican manufacturing sector would have received, on average, an additional 714 million dollars (or an additional 4.6 percent of actual average FDI inflows) per year in the period 2010 – 2015.¹⁷ See Table 4. A similar interpretation could be associated to unc 2: Country's eco. situation in 12 months time or unc3: Firms' eco situation to invest.

¹⁶ We consider specifications 4, 5 and 6 from Table 2 to perform the simulations since they include both domestic and external factors as independent variables and given the uncertainty measure considered in each specification is negative and statistically significant.

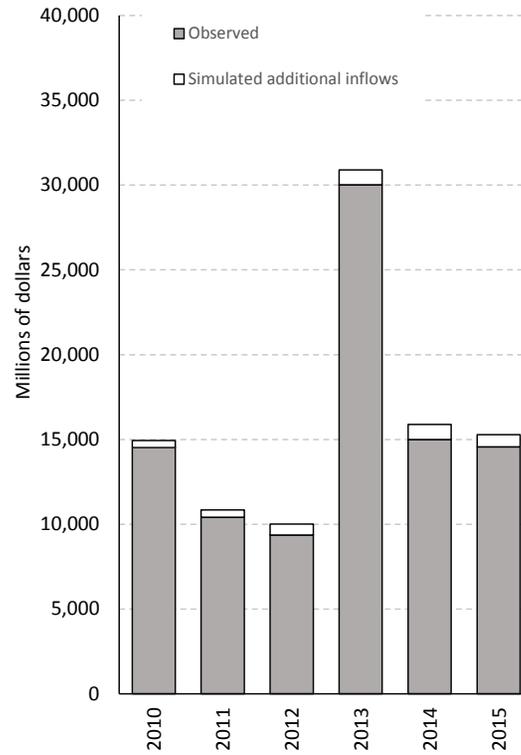
¹⁷ The counterfactual scenarios presented are conditional on the covariates used in each specification.

Figure 5. Simulations

a) Scenario: unc 1: Firms' eco.
situation in 12 months time
Specification (4) from Table 2



b) Scenario: unc 2: Country's eco.
situation in 12 months time
Specification (5) from Table 2



c) Scenario: unc 3: Firms' eco.
situation to invest
Specification (6) from Table 2

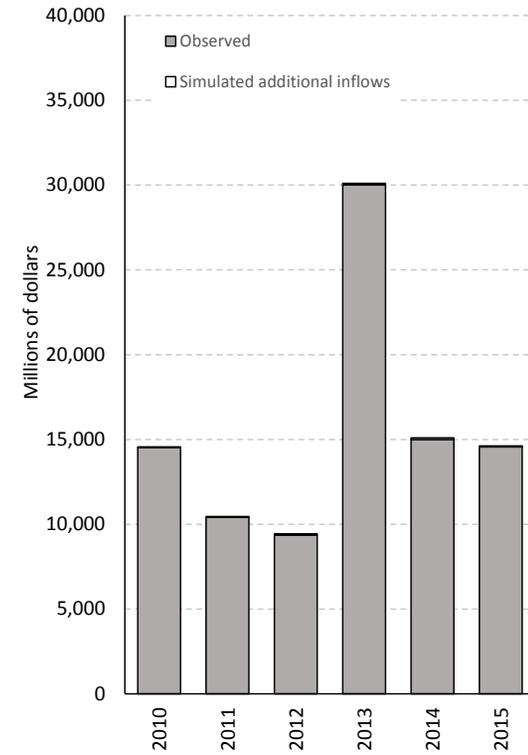


Table 4. Simulation Results

	<i>Actual Average FDI Inflows (2010 - 2015), million dollars</i>	<i>Simulated Average Additional FDI Inflows (2010 - 2015), million dollars</i>	<i>Simulated Average Additional FDI Inflows as a Percentage of Actual Average FDI Inflows</i>
Scenario: Unc 1: Firm's eco. situation in 12 months time	15,644	714	4.6
Scenario: Unc 2: Country's eco. situation in 12 months time	15,644	665	4.3
Scenario: Unc 3: Firms' eco. situation to invest	15,644	70	0.4

V. Conclusion

In this paper, we investigate the effect of uncertainty on FDI flows into the Mexican manufacturing sector over the period 2007 – 2015. To perform the analysis, we built uncertainty measures based on entrepreneurs'/forecasters' expectations about individual manufacturing subsectors' and/or Mexico's economic situation (which we call idiosyncratic uncertainty measures). Nonetheless, to verify the robustness of our results, we also use uncertainty measures that affect all manufacturing subsectors in general and that are more related to the economic and political situation of Mexico and/or of developed countries (which we call aggregate uncertainty measures). In addition to uncertainty measures, our estimated model also includes both domestic and external factors that we consider may encourage or hamper FDI. Our main results show that an increase in uncertainty discourages FDI flows into the Mexican manufacturing sector. These effects are in line with the theoretical and empirical literature on irreversible investment. We also find that the idiosyncratic uncertainty measures are more important in explaining FDI flows into the Mexican manufacturing sector than the aggregate uncertainty measures, which affect all manufacturing subsectors in general, with the exception of the global risk aversion index.

Finally, in order to assess the size of the negative effect of uncertainty on FDI flows into the Mexican manufacturing sector, we perform some simulations under the assumption that uncertainty for each and every quarter during the period 2010 – 2015 is equal to its registered level minus 1 percentage point. The results indicate that if uncertainty (proxied, as an example, by `unc1`: Firm's eco situation in 12 months time) for each and every quarter during the period 2010 – 2015 had been lower by 1 percentage point, the Mexican manufacturing sector would have received, on average, an additional 714 million dollars FDI flows (or an additional 4.6 percent of actual average FDI inflows) per year over 2010 - 2015.

References

- Abel, A. B. 1983. Optimal Investment under Uncertainty. *The American Economic Review*, 73 (1): 228-233.
- . 1984. The Effects of Uncertainty on Investment and the Expected Long-Run Capital Stock. *Journal of Economic Dynamics and Control*, 7 (1): 39-53.
- . 1985. A Stochastic Model of Investment, Marginal q and the Market Value of the Firm. *International Economic Review*, 26 (2): 305-322.
- Anheuser-Busch InBev. 2017. Reports & filings.
<http://www.ab-inbev.com/investors/reports-and-filings.html>
- Arellano, M. and O. Bover. 1995. Another Look at the Instrumental Variable Estimation of Error-Components Model. *Journal of Econometrics*, 68 (1): 29-51.
- Banco de México. 2017a. *Encuesta Sobre las Expectativas de los Especialistas en Economía del Sector Privado*.
<<http://www.banxico.org.mx/informacion-para-la-prensa/comunicados/resultados-de-encuestas/expectativas-de-los-especialistas/index.html>>
- . 2017b. *Estadísticas de tasas de interés*.
<www.banxico.org.mx/SieInternet/consultarDirectorioInternetAction.do?sector=18&accion=consultarCuadroAnalitico&idCuadro=CA51&locale=es>
- . 2017c. *Estadísticas sobre el mercado cambiario*.
<<http://www.banxico.org.mx/tipcamb/tipCamMIAction.do?idioma=sp>>
- Blonigen, B. A., and M. Wang. 2004. Inappropriate Pooling of Wealthy and Poor Countries in Empirical FDI Studies. NBER Working Paper Series. WP 10378.
- Bloom, N. 2013. The Macroeconomics of Time-Varying Uncertainty. IMF Lectures, slides, <<http://pages.stern.nyu.edu/~dbackus/BFZ/Literature/Bloom-slides-Jan-13.pdf>>. Bloom, N., S. Bond, and J. Van Reenen, 2007. Uncertainty and Investment Dynamics. *Review of Economic Studies*, 74 (2): 391-415.
- Blundell, R. and S. Bond. 1998. Initial Conditions and Moment Restrictions in Dynamic Panel Data Models. *Journal of Econometrics*, 87 (1): 115-143.
- Bond, S., R. Moessner, H. Mumtaz, and M. Syed. 2005. Microeconomic Evidence on Uncertainty and Investment. Institute for Fiscal Studies.

- <<https://www.ifs.org.uk/docs/wpinvunc.pdf>>, unpublished.
- Caballero, R. J. and R. S. Pindyck. 1992. Uncertainty, Investment, and Industry Evolution. NBER Working Paper Series, No. 4160.
- Calvo, G. A., L. Leiderman, and C. M. Reinhart. 1996. Inflows of Capital to Developing Countries in the 1990s. *The Journal of Economic Perspectives*, 10 (2): 123-139.
- Carruth, A., A. Dickerson, and A. Henley. 2000. What Do We Know About Investment Under Uncertainty? *Journal of Economic Surveys*, 14 (2): 119-154.
- CBI Press Team. 2016. CBI Predicts Slower Growth as Dark Cloud of Uncertainty Looms. Confederation of British Industries, <<http://www.cbi.org.uk/news/cbi-predicts-slower-growth-as-dark-cloud-of-uncertainty-looms/>>.
- Denis, S. and P. Kannan. 2013. The Impact of Uncertainty Shocks on the UK Economy. IMF Working Paper WP/13/66.
- Dixit, A., and R. S. Pindyck. 1994. Investment under Uncertainty. Princeton University Press. Princeton. NJ.
- Economic Policy Uncertainty. <www.policyuncertainty.com/index.htm>.
- Eichengreen, B. and H. Tong. 2007. Is China's FDI Coming at the Expense of Other Countries. *Journal of the Japanese and International Economies*, 21 (2): 153-172.
- Ellyatt, H. and G. Cutmore. 2016. European Banking Sector Still in Very Difficult Place: UBS' Axel Weber, CNBC, <<https://www.cnbc.com/2016/05/24/european-banking-sector-still-in-very-difficult-place-ubs-axel-weber.html>>.
- Episcopos, A. 1995. Evidence on the Relationship between Uncertainty and Irreversible Investment. *The Quarterly Review of Economics and Finance*, 35 (1): 41-52.
- Federal Reserve System. 2016. Industrial Production and Capacity Utilization. <<https://www.federalreserve.gov/releases/g17/>>.
- Federal Reserve Bank of St. Louis. 2016. Federal Funds Rate. <<https://fred.stlouisfed.org/series/FEDFUNDS>>.
- FEMSA. 2017. *Resultados trimestrales*.

<<http://ir.femsa.com/mx/results.cfm>>.

- Ferderer, J. P. 1993. The Impact of Uncertainty on Aggregate Investment Spending: An Empirical Analysis. *Journal of Money, Credit and Banking*, 25 (1): 30-48.
- Fuss, C. and P. Vermeulen. 2004. Firms' Investment Decision in Response to Demand and Price Uncertainty. European Central Bank Working Paper Series, No. 347.
- Ghosal, V. and P. Loungani. 1996. Product Market Competition and the Impact of Price Uncertainty on Investment: Some Evidence from US Manufacturing Industries. *The Journal of Industrial Economics*, 44 (2): 217-228.
- . 2000. The Differential Impact of Uncertainty on Investment in Small and Large Businesses. *The Review of Economics and Statistics*, 82 (2): 338-343.
- Guiso, L. and G. Parigi. 1999. Investment and Demand Uncertainty. *The Quarterly Journal of Economics*, 114 (1): 185-227.
- Hartman, R. 1972. The Effects of Price and Cost Uncertainty on Investment. *Journal of Economic Theory*, 5 (2): 258-266.
- . 1976. Factor Demand with Output Price Uncertainty. *The American Economic Review*, 66 (4): 675-681.
- Instituto Nacional de Estadística y Geografía*. 2017a. *Encuesta Mensual de Opinión Empresarial*.
<http://www.inegi.org.mx/est/contenidos/proyectos/encuestas/establecimientos/otras/emoe/>.
- . 2017b. *Encuesta Mensual de la Industria Manufacturera (EMIM)*.
<http://www.inegi.org.mx/est/contenidos/proyectos/encuestas/establecimientos/secundario/emim/>.
- . 2017c. *Banco de Información Económica (BIE)*. <http://www.inegi.org.mx/sistemas/bie/>.
- Jordaan, J. A. 2012. Agglomeration and the Location Choice of Foreign Direct Investment: New Evidence from Manufacturing FDI in Mexico. *Estudios Económicos*, 27 (1): 61-97.

- Kalckreuth, U. 2000. Exploring the Role of Uncertainty for Corporate Investment Decisions in Germany. Economic Research Group of the Deutsche Bundesbank. Discussion Paper 5/00.
- Leahy, J. V. and T. M. Whited. 1996. The Effect of Uncertainty on Investment: some Stylized Facts. *Journal of Money, Credit and Banking*, 28 (1): 64-83.
- Lemi, A. and S. Asefa. 2001. Foreign Direct Investment and Uncertainty: Empirical Evidence from Africa. Center for economic Research on Africa, School of Business, Montclair State University.
<<https://msuweb.montclair.edu/~leahy/p/CERAFRM071Lemi2002.pdf>>.
- Nucci, F. and A. F. Pozzolo. 2010. The Exchange Rate, Employment, and Hours: What Firm-Level Data Say. *Journal of International Economics*, 82 (2): 112-123.
- Oi, W. Y. 1961. The Desirability of Price Instability under Perfect Competition. *Econometrica*, 29 (1): 58-64.
- Patillo, C. 1998. Investment, Uncertainty and Irreversibility in Ghana. IMF Staff Papers, 45 (3): 522-553.
- Pindyck, R. S. 1988. Irreversible Investment, Capacity Choice, and the Value of the Firm. *The American Economic Review*, 78 (5): 969-985.
- . 1991. Irreversibility, Uncertainty, and Investment. *Journal of Economic Literature*, 29 (3): 1110-1148.
- Price, S. 1996. Aggregate Uncertainty, Investment and Asymmetric Adjustment in the UK Manufacturing Sector. *Applied Economics*, 28 (11): 1369-1379.
- Roodman, D. 2006. How to do xtabond2: An introduction to difference and system GMM in Stata. *The Stata Journal*, 9(1): 86-136.
- Secretaría de Economía. 2017. Competitividad y Normatividad / Inversión Extranjera Directa.
<<https://www.gob.mx/se/acciones-y-programas/competitividad-y-normatividad-inversion-extranjera-directa?state=published>>.
- Secretariado Ejecutivo del Sistema Nacional de Seguridad Pública. 2016. Incidencia Delictiva del Fuero Común.

<<http://secretariadoejecutivo.gob.mx/incidencia-delictiva/incidencia-delictiva-fuero-comun.php>>.

Sharifi-Renani, H. and M. Mirfatah. 2012. The Impact of Exchange Rate Volatility on Foreign Direct Investment in Iran. *Procedia Economics and Finance*, 1: 365-373.

UNCTAD 2017a. Global foreign direct investment fell 13% in 2016, but modest recovery expected in 2017 - new figures,

< http://unctad.org/en/pages/new_sdetails.aspx?OriginalVersionID=1437>.

----- 2017b. FDI Statistics.

<<http://unctad.org/en/Pages/DIAE/FDI%20Statistics/FDI-Statistics.aspx>>.

Valencia-Herrera, H. and E. Gándara-Martínez. 2009. Relación entre Incertidumbre e Inversión en México, Enfoque de Opciones Reales. *Revista de Administración, Finanzas y Economía*, 3 (2): 74-90.

Varela-Mollick, A., R. Ramos-Duran, and E. Silva-Ochoa. 2006. Infrastructure and FDI Inflows into Mexico: A Panel Data Approach. *Global Economy Journal*, 6 (1): 1-25.

Ying, Y.-H. and Y. Kim. 2001. An Empirical Analysis on Capital Flows: The Case of Korea and Mexico. *Southern Economic Journal*, 67 (4):954-968.