



Europe Economics

TPD2 and standardised tobacco packaging — What impacts have they had so far?

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Summary

This report was commissioned from Europe Economics by JTI (Japan Tobacco International). In it, we have considered the impacts of TPD2 and plain packs requirements, introduced in the UK and France between 2016 and 2017, upon tobacco consumption and prevalence, in three types of models: simple linear trend models, time series models and simultaneous equations models.

- A simple trend model considers whether the prior trend in tobacco consumption or prevalence was changed at the time TPD2 and plain packs requirements were introduced.
- A time series model makes the simple trend model more sophisticated by considering the possibility that the prior evolution was more complex than simply a trend, possibly reflecting lags, seasonality and moving averages, and also taking account of prices.
- A simultaneous equations model allows for the possibility that TPD2 and plain packs requirements might have their effects on prevalence or consumption either directly or via having an impact on prices, which in turn had an impact on consumption or prevalence.

We have found no statistically significant impact, specifically:

- No statistically significant impacts on prevalence in the UK¹.
- No statistically significant impacts on consumption in the UK or in France.

We also note that at the time of the impact assessment accompanying TPD2 it was anticipated that there would be an impact by this point. In the UK government's TPD2 impact assessment of 2015 it provides estimates of projected prevalence levels and projected reductions associated with TPD2.² These were for a reduction of 1.9 per cent in both smoking prevalence and smoking consumption over 5 years, applied linearly implying a reduction of 0.38 per cent in the 1st year, equivalent to 0.08 percentage points reduction in prevalence in the first year (2016) if prevalence would have been 19.7 per cent.

No such impact is yet identifiable in the data.

¹ We have publically available data with sufficient frequency to model prevalence impacts for the UK, but not for France

² See paragraphs 73-74 of Tobacco Products Directive (TPD) IA No: 3131, 29/06/2015, <https://www.bma.org.uk/-/media/files/pdfs/working%20for%20change/policy%20and%20lobbying/uk%20consultations/po-tobacco-products-directive-impact-assesment-2015-09-01.pdf>

1 Introduction

In February 2014 the European Union agreed a revised Tobacco Products Directive³, often referred to as “TPD2”. As well as introducing various other measures, such as restrictions on the advertising of electronic cigarettes and other vaping devices, TPD2 introduced a series of additional restrictions on the packaging of tobacco products, such as:

- making 20 the minimum number of cigarettes per cigarette pack, and 30 grams the minimum weight for roll-your-own tobacco packs;
- updating health warnings and requiring that combined (picture and text) health warnings cover 65 per cent of the front and back of cigarette and roll-your-own tobacco packages; and
- banning certain descriptors on packaging of tobacco products (such as “natural” and “organic”).

At around the same time France and the UK have adopted additional measures imposing standardised packaging of tobacco products (“plain packs” which we refer to as “PP” requirements). Such measures involve precise restrictions with regards to:

- the banning of all brand elements with the exception of the name which has to appear in a standardized font and size;
- the material, size, shape and opening mechanism of packaging;
- the colour of packaging and cigarettes; and
- the font, colour, size, case and alignment of text on packs.

Europe Economics was commissioned by Japan Tobacco International to assess any impacts yet discernible from TPD2 and plain packs upon tobacco consumption in France and the UK, and smoking prevalence in the UK.

Our analysis indicates that, based on data up to and including February 2018, there has been no statistically significant relationship, in the sorts of models we have used, between TPD2 and plain packs (hereafter frequently referred to as “TPD2+PP”) requirements and tobacco consumption in France or the UK. Similarly, again in the models we have used, those data indicate no statistically significant relationship between the presence of TPD2+PP requirements and smoking prevalence in the UK.

This report is structured as follows.

- Section 2 describes the data used for the analysis and intuitively explains our modelling approach.
- Section 3 sets out the analysis on tobacco consumption in France and the UK.
- Section 4 presents the analysis on the prevalence of smoking in the UK.
- A technical appendix is included at the end of this report, setting out various mathematical points in more detail, for reference.

³ Directive 2014/40/EU ec.europa.eu/health/sites/health/files/tobacco/docs/dir_201440_en.pdf

2 Data and modelling approach

2.1 Introduction

In this section we describe the data sources the analysis relies upon and provide an intuitive description of the modelling approach adopted.

2.2 Primary data sources

The raw data underpinning the analysis has been provided by JTI (and sourced from Nielsen and IRI) and consist of monthly retail prices and numbers of sticks for each tobacco product sold in France and the UK. The volume of sticks data covers the period January 2008 — February 2018, whilst price data is available only from January 2012.

The data is available at a high level of disaggregation — information on the number of sticks sold and retail prices is provided at the product level (i.e. for each separate sub-brand and package size). In this respect, for the purpose of the analysis, the data has been aggregated so as to obtain:

- The number of sticks sold for the whole tobacco market (consisting of cigarettes, roll-your-own products, and make-your-own products).⁴
- The average price of tobacco products which has been calculated as the weighted average price across all products, whereby the weights are proportional to the number of sticks sold. These average prices have then been expressed in a “20 sticks” equivalent form.

For the UK, our analysis also incorporates information on smoking and vaping prevalence. The underlying hypothesis is that the emergence of electronic cigarettes may have affected the dynamics of the tobacco market in the UK and should therefore be incorporated in the analysis. Data on smoking and vaping prevalence in the UK were obtained from the Smoking Toolkit Study (STS).⁵ The STS includes up-to-date data tracking national smoking and vaping patterns, as well as cessation-related behaviours.

2.3 Measuring TPD2 and plain-packaging requirements

In France and the UK, the TPD2 and plain-packaging requirements were adopted in May 2016. Following that transposition there was a transition period at the end of which all tobacco products sold needed to be compliant with the new regulation. The deadlines after which all products were obliged to comply with TPD2+PP requirements were: January 2017 for France; and May 2017 for the UK. This means that, during the implementation period, TPD2+PP products were sold next to products with the “old” branded packaging format. Therefore, for the purpose of the analysis, the degree of implementation of TPD2+PP requirements can be interpreted in terms of the penetration rate of TPD2+PP compliant products in the market.

For France and the UK we have fairly detailed information on the evolution of the actual penetration rate of TPD2+PP products within each tobacco brand and, therefore we have accurate information on the penetration of TPD2-PP products.

⁴ For RYO/MYO products the raw data provides also a sticks-equivalent conversion.

⁵ See <http://www.smokinginengland.info/latest-statistics/>

2.4 Modelling approach

The class of models we rely upon in this report are so-called “time series” models, in which we first attempt to infer the underlying evolution through time of the variables we are interested in (e.g. prevalence, consumption,) and then consider whether (and if so to what extent) the introduction of TPD2 and plain packs disturbed that underlying evolution path. Such models answer the question “What would have happened had TPD2 and/or plain packs not been introduced?” roughly as “The variables we are interested in would have continued to evolve through time in the ways they had done prior to the introduction of TPD2 and/or plain packs.”

More detail on time series models

The most well-known “time series” relationship is probably a trend. Suppose that in country X the consumption of cigarettes fell steadily by 0.5 percentage points each year for 20 years before TPD2 and plain packs were introduced. Then (assuming, naïvely, for the purposes of illustration, that no other factors were found to be relevant) it might be reasonable to assume that if TPD2 and plain packs had not been introduced, the consumption of cigarettes in country X would have continued to drop by 0.5 percentage points each year. If in fact the consumption of cigarettes fell consistently by 0.75 percentage points each year after TPD2 and plain packs were introduced, we might (again assuming naïvely that no other factors were found to be relevant) infer that the introduction of TPD2 and plain packs had been correlated with an acceleration of 0.25 percentage points each year in the decline in cigarette consumption.

Trends are only one form of time series relationship. Others include lags (the value of a variable in any one year is some multiple of its value in the year before) and moving averages (which can take a number of forms — e.g. a moving average, over three periods, of the variation of a variable from its trend value). In a time series model we use our data to describe, as closely as possible, the evolution through time of the variable(s) we are interested in via such time series relationships. The impact of a measure such as TPD2 and plain packs is, then, the way in which it leads to changes in the evolution of the variable(s) of interest relative to these time series relationships.

One important point to note here is that in this sort of time series model it is not possible to disentangle the impacts of two measures introduced at the same time. For example, in France and the UK TPD2 and plain packs are introduced together (as of May 2016⁶). All that the model can do is to say whether the time series relationships were disturbed from May 2016 onwards. It cannot say why or what proportion of any impacts should be attributed to different things that happened at that same date.

A further point to note here is that the use of time series relationships automatically accounts for seasonal effects in our data (where they exist). For example if the value of a variable in a particular month of the year is always, say, elevated, then it will have some relationship to its value one year before, which the time series tests will automatically incorporate.

Pure time series versus simultaneous equations models

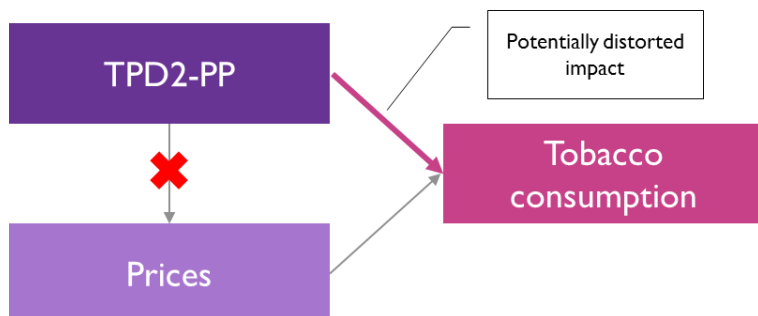
Our time series models infer time series relationships whilst controlling for prices. For example, consider a model such as the impact of TPD2 and plain packs upon consumption of tobacco products. It is natural to suppose that the average price of tobacco products might affect their consumption. So it is natural that such relative prices feature in the model.

But now suppose that as well as having a direct impact on tobacco consumption, measures such as TPD2 and plain packs also affected tobacco prices. Then there would be two routes by which impacts would occur: the direct one and the indirect one. If the model controls for price changes it will miss part of the

⁶ May 2016 corresponds to the manufacturing deadline

impact — the impact that arises indirectly by causing the prices themselves to change. Such a distortion is illustrated below:

Figure 2.1: Distorted impact found in a model where prices are controlled for but the measure has an impact on prices themselves

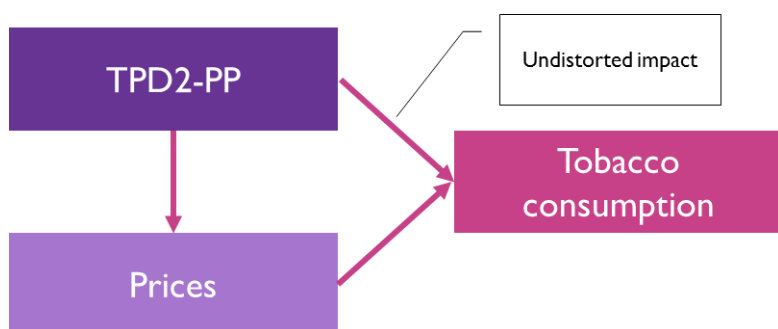


We can correct for this by what are commonly referred to as econometric “simultaneous equations” techniques. In our simultaneous equations models we estimate the impact of the measure (here, TPD2 and plain packs) upon prices and consumption (or other relevant variables in other models, such as prevalence) at the same time as measuring the direct impact of the measure upon consumption (or other variables). Doing so allows us to capture both impacts:

- The direct impact of TPD2+PP on consumption; and
- The indirect impact of TPD2+PP on consumption, which is enabled by the direct impact of TPD2+PP on prices.

These are further illustrated in the figure below:

Figure 2.2: Undistorted impact found in a model where impacts on prices and via prices are estimated simultaneously with direct impacts



3 Impacts on Consumption

3.1 Introduction

In this section we first provide an overview of smoking consumption in the UK and France. We then present a series of statistic models: to make ideas concrete we start by illustrating ideas using simple (indeed, in some senses naïve) models and then we move to more sophisticated ones.

3.2 Tobacco consumption in the UK and France

The following charts indicate the total number of sticks (these include both cigarette and RYO/MYO products) sold to consumers in the UK and France.

Figure 3.1: Number of sticks (millions) sold in the UK

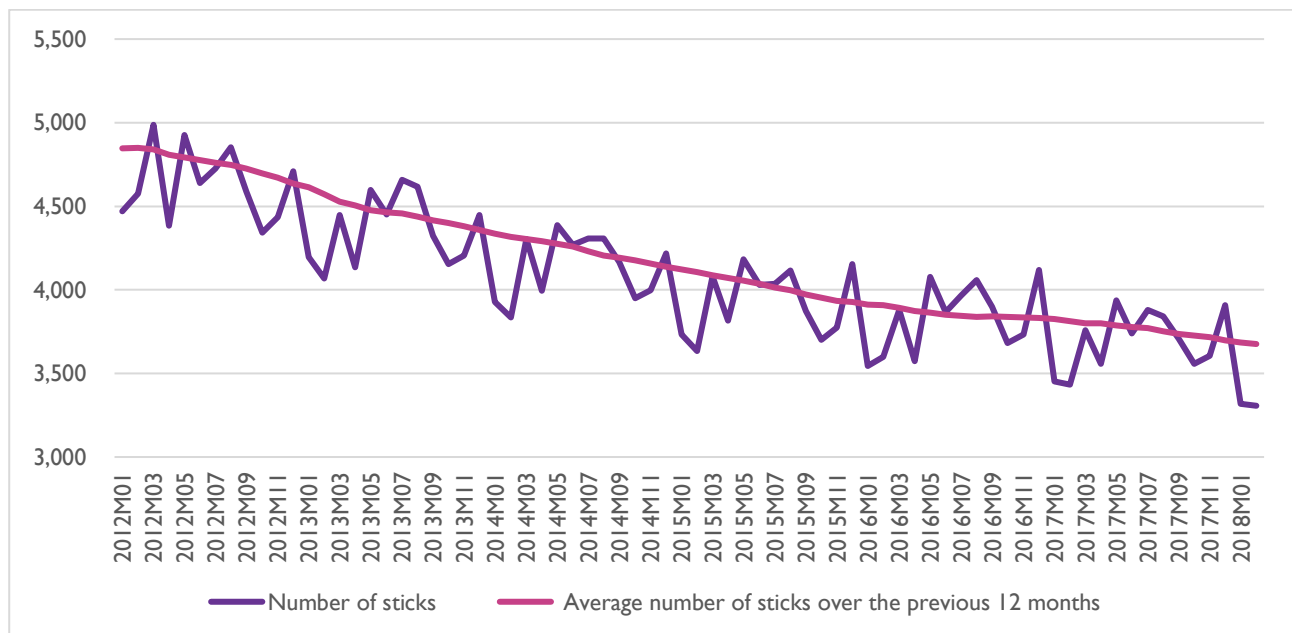
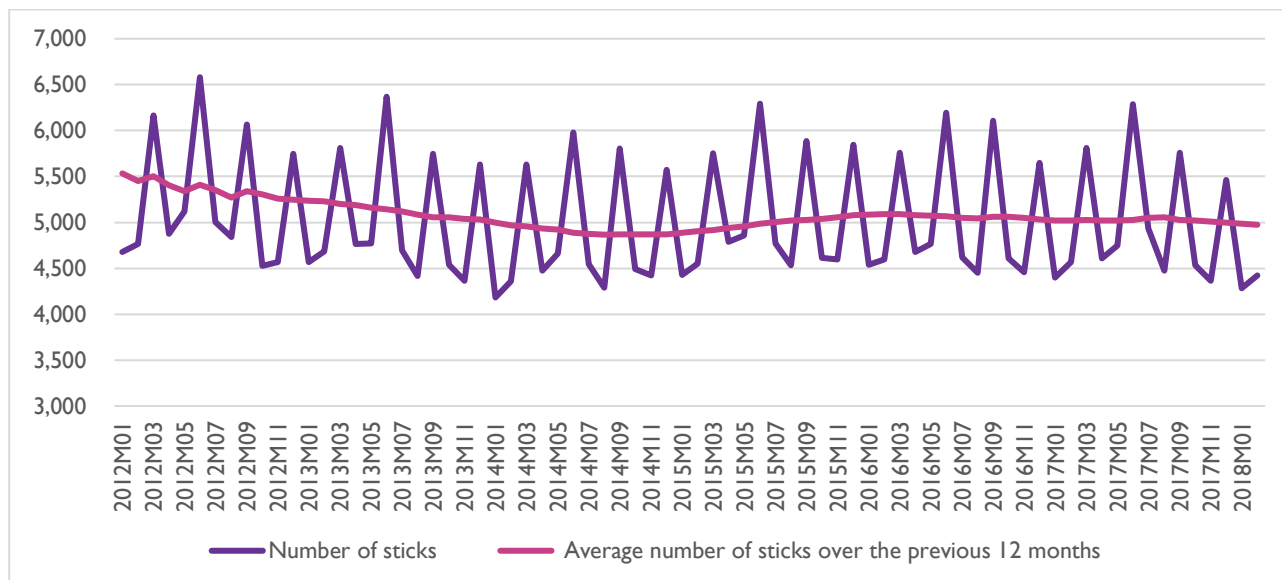


Figure 3.2: Number of sticks (millions) sold in France

We can notice from Figure 3.1 that in the UK there has been a declining trend in the number of RMC sticks sold since 2012, with perhaps some suggestion of a flattening in the rate of fall since mid-2016. In France the sales of sticks decreased up to 2014 (albeit on a declining trend less pronounced than in the UK), but somewhat stabilised afterwards.

3.3 Simple trend model

We begin with a very simple “naïve trend analysis”. In such a test we try to explain the evolution of tobacco consumption with merely a simple linear trend, and test whether there was a break in the series after the transposition of the TPD2+PP regulation. More specifically we have tested for the following types of breaks:

- A break in absolute consumption levels;
- A break in consumption trend;
- A break in, both absolute consumption levels and consumption trend.

Such a model would not itself demonstrate whether the results of later, more advanced models were correct or not, but they would help us to understand to what extent results for later, more sophisticated models with extra control variables were a matter of those extra control variables

- validating (i.e. producing similar results to);
- reinforcing (i.e. producing results with the same sign but with larger coefficients than);
- removing (i.e. eliminating results that were there before in)
- reversing (i.e. producing results in the opposite direction to); or
- adding to (i.e. producing results that were not initially there for)

For France we ran a model with a single linear trend but in such model the trend was not statistically significant. This suggests that tobacco consumption in France is better described as fluctuating around a constant mean rather than following a single linear trend. Furthermore none of the tests described above indicates the presence of a statistically significant break in a naïve trend analysis of the consumption series for France.

In contrast, the analysis for the UK suggests that tobacco consumption can be described — in statistically significant terms — as following a declining trend. The break tests we conducted suggest that the

consumption series for the UK can be described either as having a break in its level (resulting in a higher level thereafter — i.e. more tobacco consumption, not less) at around May 2016 or as having a break in trend (resulting in a lower rate of decrease in tobacco consumption — i.e. more tobacco consumption, not less) after May 2016.

Table 3.1: Results of a naïve trend model with a break in the level of consumption (UK)

	Coefficient	Std. Error	Prob.
Constant	5,610,000,000	137,000,000	0.0000
Trend	-19,250,399	1,847,839	0.0000
Dummy (<i>break in levels</i>)	240,000,000	86,354,359	0.0070

Table 3.2: Results of a naïve trend model with a break in consumption trend (UK)

	Coefficient	Std. Error	Prob.
Constant	5,620,000,000	140,000,000	0.0000
Trend	-19,321,918	1,888,478	0.0000
Trend * dummy (<i>break in trend</i>)	2,197,146	804,050	0.0079

We also checked whether there might be a break affecting both consumption levels and consumption trend simultaneously. We do not find such a break to be statistically significant. So there is either a rise in tobacco consumption levels, or a fall in the rate of decrease, but not both.

Table 3.3: Results of a naïve trend model with a break in consumption level and trend (UK)

	Coefficient	Std. Error	Prob.
Constant	5,610,000,000	143,000,000	0.0000
Trend	-19,147,176	1,929,494	0.0000
Dummy (<i>break in levels</i>)	397,000,000	783,000,000	0.6139
Trend * dummy (<i>break in trend</i>)	-1,468,621	7,277,940	0.8407

More specifically, the results of Table 3.1 indicate that the introduction of TPD2+PP is associated (in this simple model) with a permanent increase in consumption of around 240 million sticks. Similarly, the results of Table 3.2 indicate that, whilst before the introduction of TPD2+PP, the monthly reduction in number of sticks sold in the UK was, on average, of the order of 19 million (the coefficient of the trend variable in Table 3.2), since May 2016 the average monthly reduction is of the order of approximately 17 million (the coefficient of the coefficient of the trend variable in Table 3.2 plus that of the trend-times-dummy variable, i.e. $-19.3\text{m} + 2.2\text{m} = -17.1\text{m}$).

Thus **in this simple model, the introduction of TPD2+PP results in higher tobacco consumption.** We shall now explore to what extent this basic result from the data is validated, reinforced, removed, reversed or added to in more sophisticated models. We shall see that for our preferred class of models the result is removed (i.e. there is no impact), but in some of our cross-check

models for the UK it is validated (i.e. the result that TPD2+PP is associated with an increase in consumption in the UK is repeated).⁷

3.4 More sophisticated tests

As noted, the naïve trend model does not take into account certain statistical properties of the data. For example, a statistical inspection of the data indicates that tobacco consumption has a strong seasonal component (i.e. tobacco consumption higher in certain months of the year) and the simple trend model does not account for this. Moreover, the introduction of TPD2+PP regulation is modelled with a crude dummy variable which does not reflect the fact that the penetration of TPD2+PP compliant products in the market has increased gradually over the implementation period. Finally, the model does not account for other economic variables (such as prices or household income) that may also affect consumption.

We have therefore implemented more sophisticated tests to model the evolution of tobacco consumption. The first class of such tests includes pure-time series models, the second class includes simultaneous equation models tests. These are presented in turns below.

3.4.1 Time-series models

The time-series models we use aim to explain monthly percentage changes in the number of sticks sold by time series components, monthly percentage changes in *prices* and the *penetration of TPD2+PP products* in the market. The models also include *monthly dummy variables* to account for seasonal patterns in the data.

If we were conducting tests over a large number of years, as well as adjusting for prices it would be important to adjust, also, for household income. Our data is, however, relatively high frequency (monthly) but over only a few years (five). That means both that adjusting for household income is less crucial (it is less likely to be statistically significant over a short time period), less available (GDP data is available quarterly, not monthly, not available for the last two data points in our series and GDP per capita data is less reliable than GDP data at very short time periods because population estimates tend not to be reliable at very high frequency), and less straightforwardly interpreted (it is not clear that monthly fluctuations in GDP per capita would, even in theory, be expected to drive fluctuations in consumption insofar as such fluctuations reflected annual income stream volatility (e.g. self-employment revenue streams, bonuses, etc). Nonetheless, we have cross-checked the results that follow using models that allowed for the presence of GDP per capita, also.⁸ None of the results below changed materially.

In all the models we have tested there is a strong seasonal pattern to consumption. More formally, monthly dummies are strongly significant and account for a sizable portion of the variation in the data, or, in other words, information on the specific calendar month is a very strong predictor of the tobacco consumption taking place in that month.

Finally, in addition to seasonal patterns, changes in prices and, changes in the penetration of TPD2+PP compliant products, there might be other factors that are important in explaining the evolution of consumption behaviour. When data evolve through time, it is common to model them using a class of what are referred to as “autoregressive–integrated–moving–average” (ARIMA) processes. Such a process attempts to describe the behaviour of variables by exploiting any systematic relationship between a variable’s current value and its past values.

⁷ Specifically, we find an increase of around 100m in the level of number of sticks sold in the UK. That compares with the increase of 200m we found in Table 3.1. We regard these as broadly similar impacts, in this context.

⁸ GDP per capita is statistically significant in the preferred consumption model and prevalence model for the UK, but leaves the results intact. In the preferred consumption model for France GDP per capita is not statistically significant.

Two key components in ARIMA processes are the “autoregressive” (AR) term and the moving average (MA) term.⁹ The “autoregressive” term describes how the present value of the variable depends on its previous value at some point in the past (say the previous month, or three months ago, or twelve months ago). The moving average term describes how the noisy fluctuations around the current values of the variable depend on the noisy fluctuations observed in the past.¹⁰ ARIMA models are particularly useful because they can provide an accurate description of a time-series variable by using only the information contained in the variables itself, i.e. without the need for additional control variables. However, in our setting the purpose of including an ARIMA process is that of capturing residual patterns in consumption data that cannot be explained by the other explanatory variables included in the model (namely seasonal dummies, prices and TPD-2 penetration). Among the available range of ARIMA models, the “best” ARIMA model (i.e. the one with the correct orders for the AR and MA terms) can be selected algorithmically based on standard statistical tests.¹¹

We present below the results for the time series model that our algorithm indicated should be our preferred time series model for both the UK and France (for presentational purposes the results for the monthly seasonal dummies are omitted from the tables).

Table 3.4: Pure time-series model of tobacco consumption for the UK

	Coeff.	Std. Error	P-value
% Change in the average price of tobacco products	-0.323022	0.288759	0.2685
Penetration of TPD+PP products in the market	0.000414	0.001263	0.7443
AR(1)	-0.317916	0.189936	0.1003
AR(2)	0.156164	0.156641	0.3235
AR(3)	0.176782	0.101741	0.0883
MA(1)	-0.385867	0.214269	0.0776
MA(2)	-0.355758	0.148286	0.0201

Table 3.5: Pure time-series model of tobacco consumption for France

	Coeff.	Std. Error	P-value
% Change in the average price of tobacco products	0.069220	0.091970	0.4556
Penetration of TPD2-PP products in the market	-0.002631	0.004442	0.5566
AR(1)	-0.131169	0.102698	0.2081
AR(12)	-0.435833	0.197074	0.0321

⁹ The other element, the “I” in ARIMA, which stands for “Integrated”, in this context means the model is calculated in first differences (i.e. in changes in values, rather than in levels).

¹⁰ Within the broad class of ARIMA models, a specific model is characterised by an order for the autoregressive components (p), and an order for the moving average component (q). The order simply indicates the lag of the relationship linking current values to past values, so, for example, an autoregressive term of order two AR(2) indicates that the current value of a variable depends on the variables’ value observed two periods earlier.

¹¹ For example, the Akaike Information Criterion (AIC) or the Schwarz Bayesian Information Criterion (BIC). For the purpose of selecting the best ARIMA process we used the BIC statistic, applied iteratively across possible ARIMA models to an order of up to 3 so as to identify those that perform best. We then inspect the correlogram of the residuals of the preferred model in order to decide whether the inclusions of additional AR components (such as, e.g. an AR(12) term to capture residual seasonal patterns) is appropriate.

The results presented in Table 3.4 and Table 3.5 indicate that the introduction of TPD2+PP requirements do not appear to have any statistical association with the number of sticks sold.

3.4.2 Simultaneous equation model

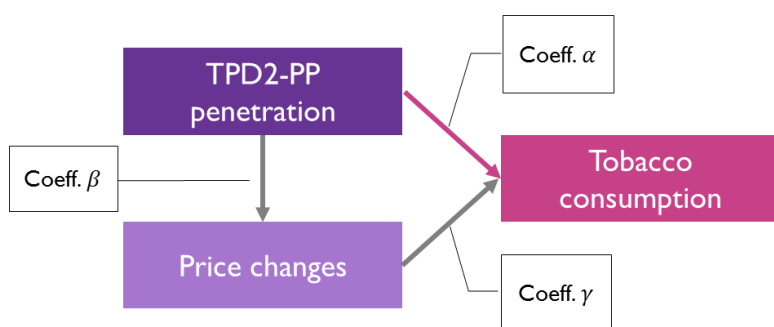
A simultaneous equation model is made up of two different equations that are estimated simultaneously. The first is a consumption equation similar to that presented in Section 2.4, where changes in the number of sticks sold are explained by *seasonal monthly* dummies, *percentage changes* in price, *penetration of TPD2+PP products* in the market and an appropriate time series process. The second equation is a price equation where changes in prices are modelled by a time-series process and the *penetration of TPD2+PP products*. Therefore, in a simultaneous equation model, the potential impact of TPD2+PP on consumption can be broken down into two different components:

- A direct impact of TPD2+PP on consumption (through the consumption equation)
- An indirect impact of TPD2+PP on consumptions that feeds through the price channel, i.e. TPD2+PP has an impact on prices (through the price equation), and prices in turns affect consumptions (through the consumption equation).

Since a moving average (MA) cannot be estimated within a simultaneous equation framework, the selection of an appropriate ARMA process for the two equations has been restricted to include only autoregressive components.

A stylised representation of our modelling approach is presented in the figure below.

Figure 3.3: Simultaneous equations approach to time series estimation of impacts on consumption



The coefficients α , β and γ are, respectively, the coefficient for the direct impact of TPD2+PP on consumption, the coefficient for the impact of TPD2+PP on prices, and the coefficient for the impact of prices on consumption. The following table, sets out the relevant conditions and calculation steps followed in order to calculate the aggregate TPD2+PP effect and to identify its constituent elements (i.e. direct and indirect, indirect only, direct only).

Table 3.6: Identifying and calculating the aggregate TPD2/PP penetration effect

Coefficient condition	Identification of aggregate effect of TPD2+PP penetration	Calculation of aggregate effect of TPD2+PP penetration
α , β , and γ are all statistically significant	Both direct and indirect effect	$\alpha + \beta \times \gamma$
β and γ are statistically significant	Only indirect effect	$\beta \times \gamma$

α is statistically significant	Only direct effect	α
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Source: Europe Economics

The results of the simultaneous equation model are reported below.

Table 3.7: Results of simultaneous equations time series model of impact of TPD2/PP on tobacco consumption in the UK and France

	Direct impact of TPD2+PP (α)	Indirect impact of TPD2+PP (β)	% Price change impact (γ)	Overall impact of TPD2/PP
UK	Statistically insignificant	Statistically insignificant	Statistically insignificant	No impact
FR	Statistically insignificant	Statistically insignificant	-0.06**	No impact

Note: * = "Significant at 90% confidence level"; ** = "Significant at 95% confidence level"; *** = "Significant at 99% confidence level".

As we can see from Table 3.7, the model estimates suggest that **there is no statistically significant impact of TPD2 and PP on tobacco consumption either directly or indirectly via an impact on prices**. The only statistically significant relationship is the negative association between price of tobacco products and the smoking consumption in France (i.e. γ is statistically significant). But since β is not statistically significant (i.e. is not statistically distinguishable from zero), that means $\beta \times \gamma$ is not statistically distinguishable from zero.

4 Time-series smoking prevalence model for the UK

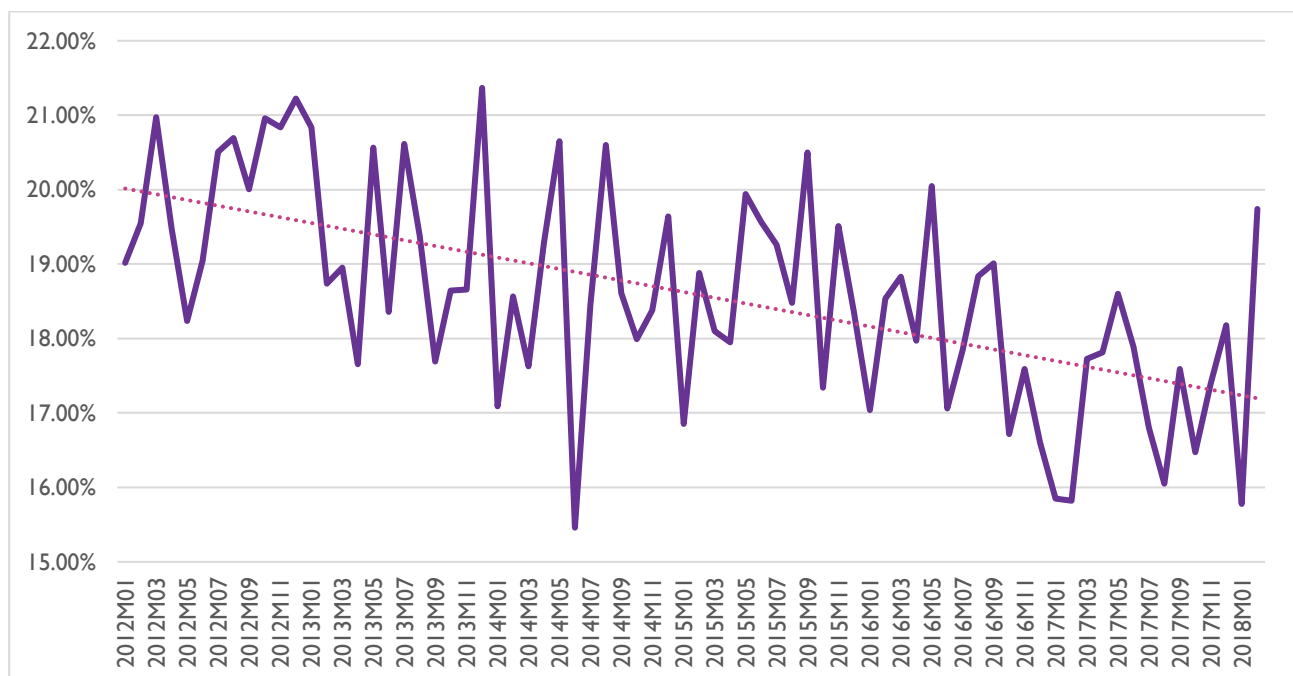
4.1 Introduction

We have publically available data with sufficient frequency to model prevalence impacts for the UK, but not for France. The data on smoking prevalence for England is reported on a monthly basis by the Smoking Toolkit Study (STS).¹² We use such data as a proxy for smoking prevalence in the UK as a whole. In this section we first provide an overview of monthly smoking prevalence in England, and we then present the results of our statistical analysis. Like for the analysis on tobacco consumption, we first conduct a simple trend analysis and we then employ more sophisticated models.

4.2 Smoking prevalence England

As we can see from Figure 4.1 smoking prevalence in England has decreased steadily since 2012. Moreover, the rate of decline in prevalence appears to follow a linear trend.

Figure 4.1: Smoking prevalence in England



4.3 Simple trend model

A simple trend model confirms the visual patterns observed in Figure 4.1, i.e. the evolution of smoking prevalence can be described – in statistically significant terms – as following a declining linear trend. We

¹² See <http://www.smokinginengland.info/latest-statistics/>

have then conducted number of tests to determine whether the series has a break (in levels, in trend, or both) in May 2016, but none of the tests suggests the presence of a statistically significant break.

4.4 More sophisticated tests

We present below the results of the pure-time series analysis and those of a simultaneous equation model

4.4.1 Time-series models

In our time-series analysis we explain model monthly changes in prevalence with monthly percentage changes in *prices* and the *penetration of TPD2+PP products* in the market. Seasonal patterns in smoking prevalence are much less marked than those observed for smoking consumption. As a result, the time-series model that our algorithm indicates should be preferred does not include monthly dummies.¹³

The results of the pure-time series analysis are reported below and indicate that there is no statistically significant association between prevalence and the introduction of TPD2 and PP requirements.

Table 4.1: Pure time-series model of smoking prevalence in the UK

	Coeff.	Std. Error	P-value
% Change in the average price of tobacco products	0.523068	0.240905	0.0343
Penetration of TPD-PP products in the market	0.002223	0.001502	0.1447
C	-0.00233	0.00095	0.0174
AR(1)	-0.840788	0.122972	0.0000
AR(2)	-0.625422	0.15369	0.0002
AR(3)	-0.30389	0.113089	0.0096
AR(12)	0.267523	0.116042	0.0250

4.4.2 Simultaneous equation model

The simultaneous equation model we present here is similar to that presented in Section 3.4.2, and attempts to identify two separate impacts:

- A direct impact of TPD2+PP on smoking prevalence;
- An indirect impact of TPD2+PP on smoking prevalence.

The results are reported below and suggest that there is no statistically significant association (either directly or indirectly) between TPD2+PP requirements and smoking prevalence in the UK.

Table 4.2: Results of simultaneous equations time series model of impact of TPD2 andPP on smoking prevalence in the UK

	Direct impact of TPD2+PP	Indirect impact of TPD2+PP	% Price change impact	Overall impact of TPD2/PP
UK	Statistically insignificant	Statistically insignificant	Statistically insignificant	No impact

Note: * = "Significant at 90% confidence level"; ** = "Significant at 95% confidence level"; *** = "Significant at 99% confidence level".

¹³ We have also run models with monthly dummy variables but such models underperformed (in terms of BIC statistics output) the model without seasonal dummies that we present here. We note that our algorithm also considered the possibility that vaping might affect prevalence, but this was not statistically significant in any of our preferred models.

5 Conclusion

In this report we have considered the impacts of TPD2 and plain packs requirements, introduced in the UK and France between 2016 and 2017, upon tobacco consumption and prevalence, in three types of models: simple linear trend models, time series models and simultaneous equations models. We have found no statistically significant impact with the models we have used:

- No statistically significant impacts on prevalence in the UK.
- No statistically significant impacts on consumption in the UK or in France.

We note that the analysis is limited in time and that the results are for the combined impact of TPD2 and plain packs rather than their individual impacts. It might be possible to attempt to disentangle the impacts of plain packs and TPD2 by using models that deployed other countries that introduced TPD2 but not plain pack requirements as controls, to explore the possibility that there was a statistically significant impact in one direction from TPD2 (e.g. a reduction in consumption) but an offsetting statistically significant impact in the opposite direction from plain packs requirements.

It is also possible that with additional time, the dynamic in the market might change and become more pronounced. We note, however, that at the time of the impact assessment accompanying TPD2 it was anticipated that there would be an impact by this point. In the UK government's TPD2 impact assessment of 2015 provides estimates of projected prevalence levels and projected reductions associated with TPD2. These were for a reduction of 1.9 per cent in both smoking prevalence and smoking consumption over 5 years, applied linearly implying a reduction of 0.38 per cent in the 1st year, equivalent to 0.08 percentage points reduction in prevalence in the first year (2016) if prevalence would have been 19.7 per cent.¹⁴

No such impact is yet identifiable in the data.

¹⁴ See paragraphs 73-74 of Tobacco Products Directive (TPD) IA No: 3131, 29/06/2015, <https://www.bma.org.uk/-/media/files/pdfs/working%20for%20change/policy%20and%20lobbying/uk%20consultations/po-tobacco-products-directive-impact-assesment-2015-09-01.pdf>

6 Appendix

This section provides a more formal details on the statistical models employed. We also provide the results for a number of alternative models we have used as a cross-check.

6.1 Simple trend models for tobacco consumption (FR and UK) and smoking prevalence (UK).

6.1.1 Consumption

The estimation results of a simple trend model without break for tobacco consumption in France and the UK are provided respectively in Table 6.1 and Table 6.2. The result indicate that tobacco consumption in France does not appear to follow a trend, whilst in the UK is follows a negative trend.

Table 6.1: Trend model of tobacco consumption for France

Dependent Variable: Number of sticks (FR)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.28E+09	3.10E+08	17.02932	0.0000
Trend	-3063549.	3596359.	-0.851848	0.3971
R-squared	0.009978	Mean dependent var		5.02E+09
Adjusted R-squared	-0.003772	S.D. dependent var		6.60E+08

Table 6.2: Trend model of tobacco consumption for the UK

Dependent Variable: Number of sticks (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.35E+09	1.02E+08	52.55156	0.0000
Trend	-15188290	1180193.	-12.86933	0.0000
R-squared	0.696995	Mean dependent var		4.08E+09
Adjusted R-squared	0.692786	S.D. dependent var		3.91E+08

We report below the results of the trend models in which three types of breaks are introduced (namely a break in the level, a break in the trend, and a break in the level and the trend). The only break tests to be statistically significant (at the 99 or 95 per cent confidence level) are the break and in the level and the break in the trend for the UK.

Table 6.3: Trend model of tobacco consumption for the UK (break in level)

Dependent Variable: Number of sticks (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.61E+09	1.37E+08	40.96751	0.0000
Trend	-19250399	1847839.	-10.41779	0.0000
Dummy	2.40E+08	86354359	2.776553	0.0070
R-squared	0.726673	Mean dependent var		4.08E+09
Adjusted R-squared	0.718974	S.D. dependent var		3.91E+08

Table 6.4: Trend model of tobacco consumption for the UK (break in trend)

Dependent Variable: Number of sticks (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.62E+09	1.40E+08	40.21754	0.0000
Trend	-19321918	1888478.	-10.23148	0.0000
Dummy*Trend	2197146.	804049.9	2.732599	0.0079
R-squared	0.725829	Mean dependent var		4.08E+09
Adjusted R-squared	0.718106	S.D. dependent var		3.91E+08

Table 6.5: Trend model of tobacco consumption for the UK (break in level and trend)

Dependent Variable: Number of sticks (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.61E+09	1.43E+08	39.24231	0.0000
Trend	-19147176	1929494.	-9.923420	0.0000
Dummy	3.97E+08	7.83E+08	0.506817	0.6139
Dummy*Trend	-1468621.	7277940.	-0.201791	0.8407
R-squared	0.726832	Mean dependent var		4.08E+09
Adjusted R-squared	0.715125	S.D. dependent var		3.91E+08

Table 6.6: Trend model of tobacco consumption for France (break in level)

Dependent Variable: Number of sticks (FR)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.42E+09	4.39E+08	12.35588	0.0000
Trend	-5262601.	5919509.	-0.889027	0.3770
Dummy	1.30E+08	2.77E+08	0.469211	0.6404
R-squared	0.013038	Mean dependent var		5.02E+09
Adjusted R-squared	-0.014764	S.D. dependent var		6.60E+08

Table 6.7: Trend model of tobacco consumption for France (break in trend)

Dependent Variable: Number of sticks (FR)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.40E+09	4.47E+08	12.08809	0.0000
Trend	-4968468.	6043151.	-0.822165	0.4137
Dummy*Trend	1012521.	2572969.	0.393522	0.6951
R-squared	0.012132	Mean dependent var		5.02E+09
Adjusted R-squared	-0.015695	S.D. dependent var		6.60E+08

Table 6.8: Trend model of tobacco consumption for France (break in level and trend)

Dependent Variable: Number of sticks (FR)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 74				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5.35E+09	4.56E+08	11.71470	0.0000
Trend	-4198566.	6164248.	-0.681116	0.4980
Dummy	1.75E+09	2.50E+09	0.698961	0.4869
Dummy*Trend	-15138618	23251192	-0.651090	0.5171
R-squared	0.018979	Mean dependent var		5.02E+09
Adjusted R-squared	-0.023064	S.D. dependent var		6.60E+08

6.1.2 Prevalence

Finally we provide here the same trend model and break test for UK smoking prevalence. The models show that smoking prevalence can be described as following declining trend without any statistically significant breaks.

Table 6.9: Trend model of smoking prevalence (UK)

Dependent Variable: Smoking prevalence (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 73				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	21.93942	0.580512	37.79323	0.0000
Trend	-0.039745	0.006703	-5.929306	0.0000
R-squared	0.331177	Mean dependent var		18.60082
Adjusted R-squared	0.321757	S.D. dependent var		1.465348

Table 6.10: Trend model of smoking prevalence (UK, break in level)

Dependent Variable: Smoking prevalence (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 73				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	21.63579	0.828175	26.12464	0.0000
Trend	-0.035185	0.011104	-3.168805	0.0023
Dummy	-0.263443	0.509899	-0.516658	0.6070
R-squared	0.333718	Mean dependent var		18.60082
Adjusted R-squared	0.314681	S.D. dependent var		1.465348

Table 6.11: Trend model of smoking prevalence (UK, break in trend)

Dependent Variable: Smoking prevalence (UK)				
Method: Least Squares				
Sample: 2012M01 2018M02				
Included observations: 73				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	21.62464	0.843931	25.62370	0.0000
Trend	-0.035035	0.011341	-3.089327	0.0029
Dummy*Trend	-0.002450	0.004745	-0.516303	0.6073
R-squared	0.333714	Mean dependent var		18.60082
Adjusted R-squared	0.314678	S.D. dependent var		1.465348

Table 6.12: Trend model of smoking prevalence (UK, break in level and trend)

Dependent Variable: Smoking prevalence (UK)				
Method: Least Squares				
Sample: 2012M01 2017M08				
Included observations: 68				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	21.65254	0.820577	26.38698	0.0000
Trend	-0.034858	0.011083	-3.145115	0.0025
Dummy	4.812869	6.982961	0.689230	0.4932
Dummy*Trend	-0.048039	0.065990	-0.727971	0.4693
R-squared	0.321712	Mean dependent var		18.77510
Adjusted R-squared	0.289917	S.D. dependent var		1.423472

6.2 Pure time-series models for tobacco consumption (FR and UK) and smoking prevalence (UK).

In Sections 6.2.1 and 6.2.2 we present the detailed results of our preferred time-series models for tobacco consumption and smoking prevalence. In Section 6.2.3 we present the result we obtain with an alternative modelling approach.

6.2.1 Benchmark models for Consumption

Table 6.13: Pure time series-model of tobacco consumption (UK)

Dependent Variable: %Change in number of sticks (UK)				
Method: Least Squares				
Sample (adjusted): 2012M05 2018M02				
Included observations: 70 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
Jan dummy	-12.94179	0.904994	-14.30042	0.0000
Feb dummy	-1.441506	0.731265	-1.971249	0.0541
Mar dummy	10.48388	0.859105	12.20327	0.0000
Apr dummy	-6.732241	0.579186	-11.62363	0.0000
May dummy	11.16269	0.786484	14.19316	0.0000
Jun dummy	-4.057646	0.482466	-8.410220	0.0000
Jul dummy	2.626959	0.820617	3.201201	0.0024
Aug dummy	0.843733	0.740402	1.139561	0.2598
Sep dummy	-4.612328	0.668338	-6.901193	0.0000
Oct dummy	-4.856267	0.393180	-12.35124	0.0000
Nov dummy	1.668171	0.214795	7.766329	0.0000
Dec dummy	7.727737	1.020683	7.571146	0.0000
% change in price	-0.323022	0.288759	-1.118654	0.2685
TPD2+PP penetration	0.000414	0.001263	0.327905	0.7443
AR(1)	-0.317916	0.189936	-1.673807	0.1003
AR(2)	0.156164	0.156641	0.996950	0.3235
AR(3)	0.176782	0.101741	1.737576	0.0883
MA(1)	-0.385867	0.214269	-1.800849	0.0776
MA(2)	-0.355758	0.148286	-2.399140	0.0201
R-squared	0.972721	Mean dependent var		-0.148157

Adjusted R-squared	0.963093	S.D. dependent var	7.156147
S.E. of regression	1.374779	Akaike info criterion	3.700651
Sum squared resid	96.39088	Schwarz criterion	4.310956
Log likelihood	-110.5228	Hannan-Quinn criter.	3.943072
Durbin-Watson stat	1.974353	Wald F-statistic	256.3743
Prob(Wald F-statistic)	0.000000		

Table 6.14: Pure time series-model of tobacco consumption (FR)

Dependent Variable: %Change in number of sticks (FR)

Method: Least Squares

Sample (adjusted): 2013M02 2018M02

Included observations: 61 after adjustments

HAC standard errors & covariance

Variable	Coefficient	Std. Error	t-Statistic	Prob.
Jan dummy	-22.26818	0.530278	-41.99337	0.0000
Feb dummy	2.854153	0.384915	7.415016	0.0000
Mar dummy	26.59104	0.446895	59.50182	0.0000
Apr dummy	-18.95243	0.409745	-46.25425	0.0000
May dummy	2.301129	0.253571	9.074898	0.0000
Jun dummy	30.46293	0.585046	52.06928	0.0000
Jul dummy	-24.35284	0.508183	-47.92137	0.0000
Aug dummy	-5.444594	0.543186	-10.02344	0.0000
Sep dummy	32.01232	1.144972	27.95905	0.0000
Oct dummy	-22.41711	0.322695	-69.46835	0.0000
Nov dummy	-2.176911	0.433479	-5.021948	0.0000
Dec dummy	26.78965	0.322916	82.96177	0.0000
% Change in price	0.069220	0.091970	0.752639	0.4556
TPD2+PP penetration	-0.002631	0.004442	-0.592395	0.5566
AR(1)	-0.131169	0.102698	-1.277230	0.2081
AR(12)	-0.435833	0.197074	-2.211526	0.0321
R-squared	0.995197	Mean dependent var	2.093888	
Adjusted R-squared	0.993596	S.D. dependent var	21.18220	
S.E. of regression	1.695069	Akaike info criterion	4.113703	
Sum squared resid	129.2967	Schwarz criterion	4.667375	
Log likelihood	-109.4679	Hannan-Quinn criter.	4.330692	
Durbin-Watson stat	1.940834	Wald F-statistic	2665.022	
Prob(Wald F-statistic)	0.000000			

6.2.2 Benchmark models for Prevalence

Table 6.15: Pure time series-model of smoking prevalence (UK)

Dependent Variable: Changes in prevalence (UK)				
Method: Least Squares				
Sample (adjusted): 2013M02 2018M02				
Included observations: 61 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price	0.523068	0.240905	2.171260	0.0343
TPD2+PP penetration	0.002223	0.001502	1.479819	0.1447
C	-0.233014	0.094961	-2.453796	0.0174
AR(1)	-0.840788	0.122972	-6.837244	0.0000
AR(2)	-0.625422	0.153690	-4.069385	0.0002
AR(3)	-0.303890	0.113089	-2.687189	0.0096
AR(12)	0.267523	0.116042	2.305396	0.0250
R-squared	0.502488	Mean dependent var		-0.018033
Adjusted R-squared	0.447208	S.D. dependent var		1.859661
S.E. of regression	1.382657	Akaike info criterion		3.593510
Sum squared resid	103.2340	Schwarz criterion		3.835741
Log likelihood	-102.6020	Hannan-Quinn criter.		3.688442
F-statistic	9.090002	Durbin-Watson stat		2.092938
Prob(F-statistic)	0.000001	Wald F-statistic		3.120805
Prob(Wald F-statistic)	0.052170			

6.2.3 Alternative modelling approaches

The time-series modelling approach used to produce the results set out in Section 6.2.1 6.2.2 relies on expressing the dependent variable (tobacco consumption and prevalence) in terms of percentage change. This transformation ensures that dependent variables are stationary and can therefore be modelled meaningfully. We then control for seasonal patterns in the data with the use of dummy variables. It is however important to stress that models in which the dependent variable are expressed in differences (or, in our case, in percentage differences) are less likely to find a statistically significant relationship between variables than models expressed in levels. In other words, if we find a statistically significant relationship in a model in difference we can be quite certain that such a relationship exists. However, if we fail to find such relationship to be statistically significant in differences, we might still be able to find it to be significant in a model expressed in levels (even though the risk of the statistical relationship being spurious is higher).

As a cross check we have therefore analysed a number of alternative models where the dependent variables are expressed in levels as opposed to percentage changes. Such modelling approach would still require some transformation of the dependent variable to ensure stationarity. In many cases we find that the dependent variables are trend-stationary, i.e. the variables themselves are non-stationary because they follow a linear trend, but their fluctuations around the trend are stationary. Furthermore, since deviations from the trend have a seasonal component, we can de-seasonalise them with the use dummy variables. Therefore after being de-trended and de-seasonalised, the variables can be analysed in a meaningful way.

One advantage of this approach is that it reduce the risk of what is called “over-differencing”. Most time series models (and in particular the form of time series models we use here) work best when they have been rendered “stationary”, i.e. normalised such that the mean, variance, autocorrelation, etc. of the normalised series are all constant over time. In a number of time series setting, differencing (i.e. considering changes in rather than the levels of variables) is a useful and common technique to render an otherwise

“non-stationary” series stationary. However, there is some risk that in the process of differencing we erase the correlation between variables that the time series model is seeking to identify (i.e. we “over-difference”). The alternative approach we set out here, where instead of differencing we use levels adjusted for trends and seasonal patterns, checks for the sorts of correlations that over-differencing might have erased.

A disadvantage of this approach — and the key reason we regard the approach we set out in the main body of this report as preferred — is that this approach is more complex and requires more judgement to execute (e.g. in determining which approach to take to detrending or seasonalising). Furthermore, our approach in the main body is slightly more demanding of the data than the alternative set out here, and thus is appropriate in testing the robustness of a perhaps-counterintuitive or at least unintended result such as that in our simple trend model — namely that TPD2+PP is associated with higher consumption. If, on the other hand, our simple trend model had identified that TPD2+PP was associated with a reduction in consumption, there might have been more of a case for considering our alternative approach as an intermediate step — i.e. we would shift from simple model to normalised levels model to differenced model and see how robust that correlation was to more and more demanding tests.

More specifically, we have carried out the following variable transformations:

- We have de-trended and de-seasonalised the number of sticks in the UK and smoking prevalence in the UK (the variables have been de-trended because the models presented above suggest that these variables follow a linear trend)
- We have de-seasonalised the number of sticks in France (the variable has not been de-trended because the analysis presented above suggests that there is no statistically significant trend)

The dependent variables after such adjustments are reported in the figures below.

Figure 6.1: Number of sticks (France and UK)

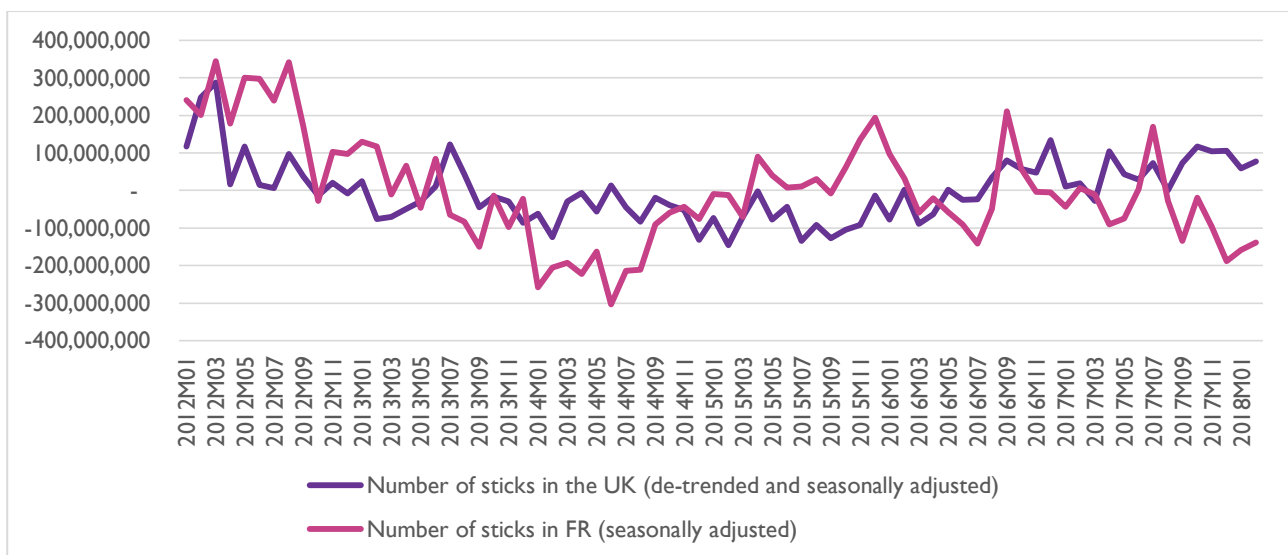
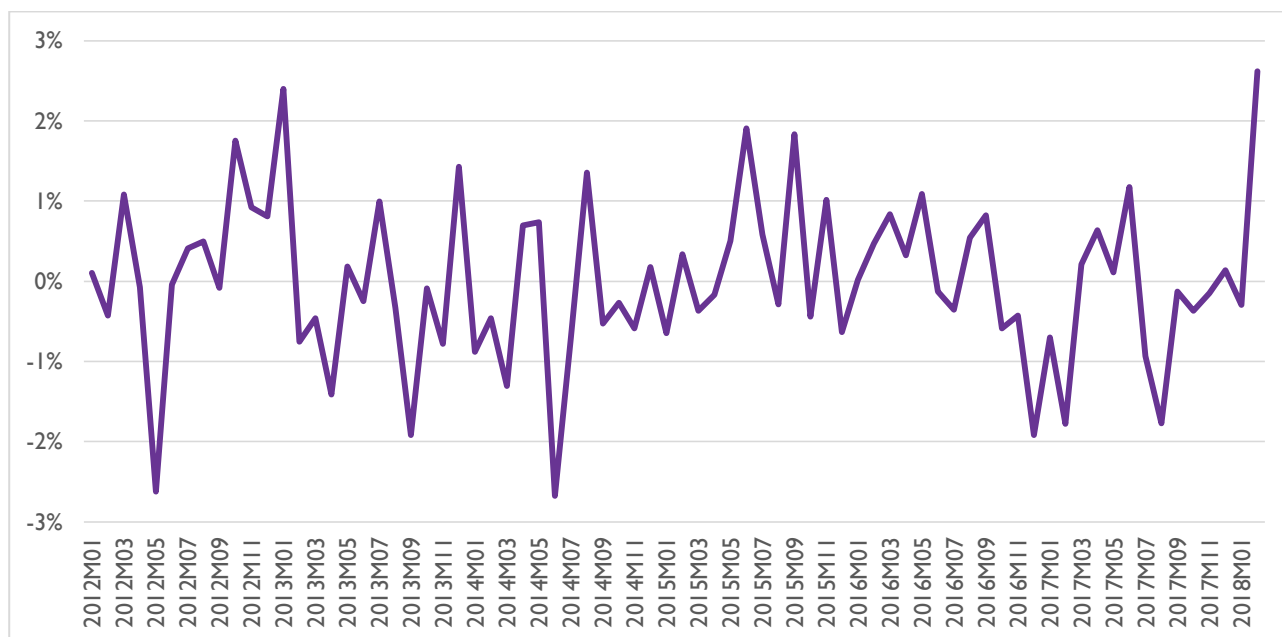


Figure 6.2: Smoking prevalence (UK)

The stationarity tests we have performed confirm that the adjusted variables are stationary and therefore can be analysed using the statistical techniques deployed throughout this section.

We have then modelled the above adjusted variables with the following explanatory variables: percentage changes in the average price of cigarettes, penetration of TPD2+PP products, and an optimally selected ARMA process. The results of such models are reported below.

Table 6.16: Alternative time-series model of tobacco consumption (UK)

Dependent Variable: Number of sticks (UK) – de-trended and de-seasonalised				
Method: Least Squares				
Sample (adjusted): 2012M05 2018M02				
Included observations: 70 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price	5.07E+08	9.65E+08	0.525614	0.6010
TPD2+PP penetration	98219533	31638094	3.104471	0.0028
C	-28591942	20191761	-1.416020	0.1616
AR(1)	0.313583	0.102408	3.062108	0.0032
AR(2)	0.257363	0.087098	2.954872	0.0044
AR(3)	0.048925	0.079841	0.612779	0.5422
R-squared	0.498425	Mean dependent var		-9555653.
Adjusted R-squared	0.459239	S.D. dependent var		70049028
S.E. of regression	51511554	Akaike info criterion		38.43433
Sum squared resid	1.70E+17	Schwarz criterion		38.62706
Log likelihood	-1339.201	Hannan-Quinn criter.		38.51088
F-statistic	12.71961	Durbin-Watson stat		1.912223
Prob(F-statistic)	0.000000	Wald F-statistic		4.881700
Prob(Wald F-statistic)	0.010638			

Table 6.17: Alternative time-series model of tobacco consumption (FR)

Dependent Variable: Number of sticks (FR) –de-seasonalised				
Method: Least Squares				
Sample (adjusted): 2013M01 2018M02				
Included observations: 62 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price	-1.03E+08	3.71E+08	-0.278810	0.7815
TPD2+PP penetration	-15973893	73914030	-0.216114	0.8297
C	-33094202	14397892	-2.298545	0.0255
AR(1)	0.474736	0.089660	5.294846	0.0000
AR(2)	0.828118	0.064204	12.89824	0.0000
AR(3)	-0.363941	0.087045	-4.181080	0.0001
AR(12)	-0.122706	0.031694	-3.871541	0.0003
MA(1)	-0.008433	0.052494	-0.160642	0.8730
MA(2)	-0.991288	0.066737	-14.85360	0.0000
R-squared	0.624097	Mean dependent var		-40125830
Adjusted R-squared	0.567357	S.D. dependent var		1.10E+08
S.E. of regression	72315733	Akaike info criterion		39.16446
Sum squared resid	2.77E+17	Schwarz criterion		39.47324
Log likelihood	-1205.098	Hannan-Quinn criter.		39.28570
F-statistic	10.99922	Durbin-Watson stat		1.901815
Prob(F-statistic)	0.000000	Wald F-statistic		0.066996
Prob(Wald F-statistic)	0.935278			

Table 6.18: Alternative time-series model of smoking prevalence (UK)

Dependent Variable: Smoking prevalence (UK) – de-trended and de-seasonalised				
Method: Least Squares				
Sample (adjusted): 2013M02 2018M02				
Included observations: 61 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price	-0.108527	0.180595	-0.600944	0.5503
TPD2+PP penetration	0.000418	0.003246	0.128886	0.8979
C	-0.000481	0.001439	-0.334536	0.7392
AR(12)	-0.171570	0.114974	-1.492248	0.1411
R-squared	0.035632	Mean dependent var		-0.000776
Adjusted R-squared	-0.015124	S.D. dependent var		0.009836
S.E. of regression	0.009910	Akaike info criterion		-6.327183
Sum squared resid	0.005598	Schwarz criterion		-6.188765
Log likelihood	196.9791	Hannan-Quinn criter.		-6.272936
F-statistic	0.702028	Durbin-Watson stat		1.905159
Prob(F-statistic)	0.554738	Wald F-statistic		0.183404
Prob(Wald F-statistic)	0.832921			

We see from Table 6.16 that under this alternative modelling approach the introduction of TPD2+PP in the UK and is statistically associated (at the 95 per cent confidence level) with an increase in the number of sticks sold. In the other models we have considered the introduction of TPD2+PP remains statistically insignificant.

6.3 Simultaneous equation models of tobacco consumption (FR and UK) and smoking prevalence (UK)

In Sections 6.3.1 and 6.3.2 we present the detailed results of our simultaneous-equation approach. In Section 6.3.3 we present the result we obtain when we model the consumption equation and the prevalence with the alternative approach described in 6.2.3, i.e. by using de-trended and de-seasonalised series.

6.3.1 Consumption

Below we first present estimation results for the two separate equations (consumption equation and the price equation) constituting the system model for the UK. We then provide the estimation output of the system as a whole (the coefficients of the systems are coded from C(1) to C(21)).

Table 6.19: Consumption equation (UK)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
Dependent Variable: %Change in number of sticks (UK)				
Method: Least Squares				
Sample (adjusted): 2012M05 2018M02				
Included observations: 70 after adjustments				
HAC standard errors & covariance				
Jan dummy – C(1)	-0.132199	0.008410	-15.71995	0.0000
Feb dummy – C(2)	-0.013161	0.007143	-1.842456	0.0710
Mar dummy – C(3)	0.102800	0.009145	11.24089	0.0000
Apr dummy – C(4)	-0.067946	0.005979	-11.36349	0.0000
May dummy – C(5)	0.108660	0.007646	14.21206	0.0000
Jun dummy – C(6)	-0.041396	0.005075	-8.157377	0.0000
Jul dummy – C(7)	0.024498	0.007751	3.160485	0.0026
Aug dummy – C(8)	0.007790	0.006769	1.150715	0.2550
Sep dummy – C(9)	-0.046659	0.006327	-7.374707	0.0000
Oct dummy – C(10)	-0.048137	0.003601	-13.36791	0.0000
Nov dummy – C(11)	0.015825	0.002416	6.549390	0.0000
Dec dummy – C(12)	0.077111	0.009859	7.821773	0.0000
% Change in price – C(13)	-0.110357	0.317699	-0.347363	0.7297
TPD2+PP penetration – C(14)	0.000543	0.001782	0.304618	0.7618
AR(1) – C(15)	-0.527014	0.101604	-5.186940	0.0000
AR(2) – C(16)	-0.145966	0.130173	-1.121322	0.2672
AR(3) – C(17)	0.052686	0.095041	0.554352	0.5817

Table 6.20: Price equation (UK)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
Dependent Variable: %Change in price (UK)				
Method: Least Squares				
Sample (adjusted): 2013M02 2018M02				
Included observations: 61 after adjustments				
HAC standard errors & covariance				
TPD2+PP penetration – C(18)	0.002483	0.002281	1.088902	0.2807
C – C(19)	-0.001492	0.002195	-0.679690	0.4994
AR(12) – C(20)	0.761434	0.070214	10.84450	0.0000

Table 6.21: Simultaneous equation model of consumption (UK)

System (UK)				
Estimation Method: Iterative Least Squares				
Sample: 2012M05 2018M02				
Included observations: 73				
Total system (unbalanced) observations 131				
	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	-0.132199	0.008516	-15.52343	0.0000
C(2)	-0.013161	0.007110	-1.850929	0.0668
C(3)	0.102800	0.007932	12.96015	0.0000
C(4)	-0.067946	0.007438	-9.134700	0.0000
C(5)	0.108660	0.007192	15.10875	0.0000
C(6)	-0.041396	0.006812	-6.076697	0.0000
C(7)	0.024498	0.007836	3.126460	0.0023
C(8)	0.007790	0.006729	1.157591	0.2495
C(9)	-0.046659	0.006730	-6.932456	0.0000
C(10)	-0.048137	0.006689	-7.196030	0.0000
C(11)	0.015825	0.006790	2.330706	0.0216
C(12)	0.077111	0.006674	11.55483	0.0000
C(13)	-0.110363	0.405717	-0.272020	0.7861
C(14)	0.000543	0.002957	0.183591	0.8547
C(15)	-0.527017	0.128109	-4.113826	0.0001
C(16)	-0.145969	0.145286	-1.004696	0.3172
C(17)	0.052684	0.123116	0.427924	0.6695
C(18)	0.002483	0.001548	1.603915	0.1116
C(19)	-0.001492	0.003278	-0.455146	0.6499
C(20)	0.761434	0.088752	8.579308	0.0000

For France, estimation results for the two separate system equations and for the system as a whole are reported below (the coefficients of the systems are coded from C(1) to C(18)).

Table 6.22: Consumption equation (FR)

Dependent Variable: %Change in number of sticks (FR)				
Method: Least Squares				
Sample (adjusted): 2013M02 2018M02				
Included observations: 61 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
Jan dummy – C(1)	-22.26818	0.530278	-41.99337	0.0000
Feb dummy – C(2)	2.854153	0.384915	7.415016	0.0000
Mar dummy – C(3)	26.59104	0.446895	59.50182	0.0000
Apr dummy – C(4)	-18.95243	0.409745	-46.25425	0.0000
May dummy – C(5)	2.301129	0.253571	9.074898	0.0000
Jun dummy – C(6)	30.46293	0.585046	52.06928	0.0000
Jul dummy – C(7)	-24.35284	0.508183	-47.92137	0.0000
Aug dummy – C(8)	-5.444594	0.543186	-10.02344	0.0000
Sep dummy – C(9)	32.01232	1.144972	27.95905	0.0000
Oct dummy – C(10)	-22.41711	0.322695	-69.46835	0.0000
Nov dummy – C(11)	-2.176911	0.433479	-5.021948	0.0000
Dec dummy – C(12)	26.78965	0.322916	82.96177	0.0000
% Change in price – C(13)	0.069220	0.091970	0.752639	0.4556
TPD2+PP penetration – C(14)	-0.002631	0.004442	-0.592395	0.5566
AR(1) – C(15)	-0.131169	0.102698	-1.277230	0.2081
AR(12) – C(16)	-0.435833	0.197074	-2.211526	0.0321

Table 6.23: Price equation (FR)

Dependent Variable: %Change in price (FR)

Method: Least Squares

Sample (adjusted): 2013M01 2018M02

Included observations: 61 after adjustments

HAC standard errors & covariance

Variable	Coefficient	Std. Error	t-Statistic	Prob.
TPD2+PP penetration – C(17)	0.010228	0.006421	1.592829	0.1165
C – C(18)	0.001308	0.002209	0.592220	0.5560
AR(12) – C(19)	0.409888	0.365466	1.121549	0.2666

System (FR)

Estimation Method: Iterative Least Squares

Sample: 2013M01 2018M02

Included observations: 74

Total system (unbalanced) observations 123

	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	-22.26818	0.565349	-39.38840	0.0000
C(2)	2.854152	0.542910	5.257133	0.0000
C(3)	26.59104	0.533745	49.81973	0.0000
C(4)	-18.95243	0.533960	-35.49410	0.0000
C(5)	2.301129	0.534415	4.305889	0.0000
C(6)	30.46293	0.541047	56.30369	0.0000
C(7)	-24.35284	0.536797	-45.36694	0.0000
C(8)	-5.444594	0.539977	-10.08301	0.0000
C(9)	32.01232	0.537099	59.60225	0.0000
C(10)	-22.41711	0.538321	-41.64261	0.0000
C(11)	-2.176911	0.544026	-4.001482	0.0001
C(12)	26.78965	0.546907	48.98390	0.0000
C(13)	0.069221	0.119399	0.579744	0.5633
C(14)	-0.002631	0.004555	-0.577721	0.5647
C(15)	-0.131169	0.127430	-1.029340	0.3057
C(16)	-0.435833	0.113845	-3.828295	0.0002
C(17)	0.010228	0.006278	1.629120	0.1063
C(18)	0.001308	0.004722	0.276971	0.7824
C(19)	0.409900	0.211180	1.940993	0.0550

6.3.2 Prevalence

The estimation results of the two equations constituting the prevalence system model for the UK are reported below together with the estimation output of the model as a whole.

Table 6.24: Prevalence equation (UK)

Dependent Variable Change in prevalence (UK)				
Method: Least Squares				
Sample (adjusted): 2013M02 2018M02				
Included observations: 61 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price – C(1)	0.523068	0.240905	2.171260	0.0343
TPD2+PP penetration – C(2)	0.002223	0.001502	1.479819	0.1447
C – C(3)	-0.002330	0.000950	-2.453796	0.0174
AR(1) – C(4)	-0.840788	0.122972	-6.837244	0.0000
AR(2) – C(5)	-0.625422	0.153690	-4.069385	0.0002
AR(3) – C(6)	-0.303890	0.113089	-2.687189	0.0096
AR(12) – C(7)	0.267523	0.116042	2.305396	0.0250

Table 6.25: Price equation (UK)

Dependent Variable: %Change in price (UK)				
Method: Least Squares				
Sample (adjusted): 2013M02 2018M02				
Included observations: 61 after adjustments				
Convergence achieved after 4 iterations				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
TPD2+PP penetration – C(8)	0.002483	0.002281	1.088902	0.2807
C – C(9)	-0.001492	0.002195	-0.679690	0.4994
AR(12) – C(10)	0.761434	0.070214	10.84450	0.0000

Table 6.26: Simultaneous equation model of prevalence (UK)

System (UK)				
Estimation Method: Iterative Least Squares				
Sample: 2013M02 2018M02				
Included observations: 73				
Total system (balanced) observations 122				
	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	0.523059	0.297759	1.756651	0.0817
C(2)	0.002223	0.001774	1.253636	0.2126
C(3)	-0.002330	0.001131	-2.060607	0.0417
C(4)	-0.840788	0.132063	-6.366551	0.0000
C(5)	-0.625421	0.157376	-3.974047	0.0001
C(6)	-0.303890	0.137549	-2.209326	0.0292
C(7)	0.267521	0.110963	2.410912	0.0175
C(8)	0.002483	0.001548	1.603915	0.1115
C(9)	-0.001492	0.003278	-0.455146	0.6499
C(10)	0.761434	0.088752	8.579308	0.0000

6.3.3 Simultaneous equation under alternative approach.

The consumption equations for France and the UK under the alternative approach are reported below (notice that, for both countries, the adjusted consumption variables are best explained by the same AR structure, ie one with one AR(1) and one AR(12) term).

Table 6.27: Alternative consumption equation (UK)

Dependent Variable: Number of sticks (de-trended and de-seasonalised)				
Method: Least Squares				
Sample (adjusted): 2012M05 2018M02				
Included observations: 70 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price – C(1)	5.07E+08	9.65E+08	0.525614	0.6010
TPD2+PP penetration – C(2)	98219533	31638094	3.104471	0.0028
C – C(3)	-28591942	20191761	-1.416020	0.1616
AR(1) – C(4)	0.313583	0.102408	3.062108	0.0032
AR(2) – C(5)	0.257363	0.087098	2.954872	0.0044
AR(3) – C(6)	0.048925	0.079841	0.612779	0.5422

Table 6.28: Alternative consumption equation (FR)

Dependent Variable: Number of sticks (de-seasonalised)				
Method: Least Squares				
Sample (adjusted): 2013M01 2018M02				
Included observations: 62 after adjustments				
HAC standard errors & covariance				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
% Change in price – C(1)	-1.04E+08	5.68E+08	-0.182149	0.8561
TPD2+PP penetration – C(2)	-16602640	47461370	-0.349814	0.7278
C – C(3)	-35691380	24183156	-1.475878	0.1455
AR(1) – C(4)	0.653903	0.094461	6.922461	0.0000
AR(12) – C(5)	-0.129300	0.060904	-2.123014	0.0381

The price equations for France and the UK are the same as in Table 6.20 and Table 6.23. However to facilitate the interpretation of the simultaneous equations models the coefficients of the price equations can be renumbered as follows:

Dependent variables and coefficients codes of price equation for the UK

TPD2+PP penetration – C(7)

C – C(8)

AR(12) – C(9)

Dependent variables and coefficients codes of price equation for France

TPD2+PP penetration – C(6)

C – C(7)

AR(12) – C(8)

The estimation output of the simultaneous equating models for consumptions are below.

Table 6.29: Alternative simultaneous equation model of consumption (UK)

System (UK)				
Estimation Method: Iterative Least Squares				
Sample: 2012M05 2018M02				
Included observations: 73				
Total system (unbalanced) observations 131				
	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	4.38E+08	8.84E+08	0.496172	0.6207
C(2)	86655998	39007116	2.221543	0.0282
C(3)	-30922171	18681632	-1.655218	0.1005
C(4)	0.313298	0.119879	2.613451	0.0101
C(5)	0.257441	0.110277	2.334505	0.0212
C(6)	0.055265	0.101015	0.547093	0.5853
C(7)	0.002483	0.001548	1.603915	0.1113
C(8)	-0.001492	0.003278	-0.455146	0.6498
C(9)	0.761434	0.088752	8.579308	0.0000

Table 6.30: Alternative simultaneous equation model of consumption (FR)

System: (FR)				
Estimation Method: Iterative Least Squares				
Sample: 2013M01 2018M02				
Included observations: 74				
Total system (balanced) observations 124				
	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	8.71E+08	4.77E+08	1.826911	0.0703
C(2)	-67858729	69292152	-0.979313	0.3295
C(3)	-20699238	28070297	-0.737407	0.4624
C(4)	0.710073	0.097520	7.281338	0.0000
C(5)	-0.145807	0.072180	-2.020043	0.0457
C(6)	0.010228	0.006278	1.629147	0.1060
C(7)	0.001308	0.004722	0.276965	0.7823
C(8)	0.409904	0.211180	1.941015	0.0547

The statistical significance of the coefficient C(2) in Table 6.29 confirms results we obtained with the alternative time-series model of Table 6.16; in the UK the introduction of TPD2+PP is associated with an increase in tobacco consumption. However, the alternative simultaneous equation model for France indicates that there is no statistically significant relationship between TPD2+PP and tobacco consumption.

Finally, the prevalence equations for the UK under the alternative approach is the same provided in Table 6.18, whilst the price equation is the same as in Table 6.20. In order to facilitate the interpretation of the simultaneous equation system the coefficients of the two equations (prevalence equation and price equation) are labelled as follows:

Dependent variables and coefficients codes of the prevalence equation for the UK

% Change in price – C(1)
 TPD2+PP penetration – C(2)
 C – C(3)
 AR(12) – C(4)

Dependent variables and coefficients codes of price equation for the UK

TPD2+PP penetration – C(5)
 C – C(6)
 AR(12) – C(7)

As Table 6.31 shows there is no statistically significant relationship between prevalence and the introduction of TPD2+PP in the UK.

Table 6.31: Alternative simultaneous equation model of prevalence (UK)

	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	-0.108527	0.181238	-0.598810	0.5505
C(2)	0.000418	0.003294	0.126989	0.8992
C(3)	-0.000481	0.001343	-0.358384	0.7207
C(4)	-0.171570	0.123470	-1.389565	0.1673
C(5)	0.002483	0.001548	1.603915	0.1115
C(6)	-0.001492	0.003278	-0.455146	0.6499
C(7)	0.761434	0.088752	8.579308	0.0000