

Size of airports in France

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April 8, 2016

What can affect the size of an airport in France?

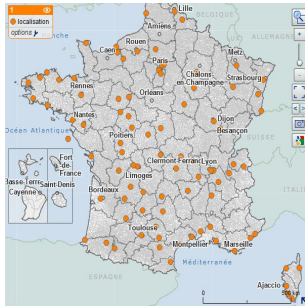


Figure: Maps of the airports in France in 2011.

Source: data.gouv.fr

There are about 160 airports in France, with different sizes such as Charles de Gaulle or Carcassonne.

What can affect the size of airports in France ?

How did we gather our data ?

Sources

Laboratoire d'Economie de l'ENAC, Albatros' database, Insee's database and data.gouv.fr

Dependent variables

Total passenger traffic, domestic passenger traffic, international passenger traffic, Low-Cost passenger traffic, total cargo traffic

Explanatory variables

City's size, city's inhabitants, distance to a major city, region's GDP, number of tourist attractions, high-speed rail link (TGV), airport operator

Database

Year	code airport IATA	name	lat	lon	city	country	DAG	mt	pas_dom	pas_int	pas_tot	fuel	mg	monume	monume	st_ano	rufts	SDV	vite	class	n_vite
2000	AIA	AERODIOMI CAMPIDELL'ORSO, CORSICA, FRANCE	14.913	1.024.880	44.549			1.089.429	7.031	3	303									0	231,1
2005	AIA	AERODIOMI CAMPIDELL'ORSO, CORSICA, FRANCE	13.096	949.588	50.205			875.793	6.423	3	303									0	231,1
2010	AIA	AERODIOMI CAMPIDELL'ORSO, CORSICA, FRANCE	17.761	1.033.129	54.938			1.110.987	5.969	3	303	118	3							0	231,1
2015	AIA	AERODIOMI CAMPIDELL'ORSO, CORSICA, FRANCE	14.530	1.041.130	332.464			1.345.094	5.988	3	303									0	231,1
2000	AIA	AERODIOMI CAMPIDELL'ORSO, CORSICA, FRANCE	14.549	1.038.931	331.991			1.333.933	6.096	3	303									0	231,1
2005	AIA	AERODIOMI CAMPIDELL'ORSO, CORSICA, FRANCE	17.760	1.024.461	331.849			1.106.306	7.098	3	303									0	231,1
2010	BIA	BASTIA-POBETTA, CORSICA, FRANCE	17.786	778.769	54.205			832.074	7.321	3	303									0	209,3
2000	BIA	BASTIA-POBETTA, CORSICA, FRANCE	17.961	743.977	75.438			820.381	8.063	3	303									0	209,3
2010	BIA	BASTIA-POBETTA, CORSICA, FRANCE	13.037	881.450	113.075			1.006.125	6.976	3	303	184	3							0	209,3
2015	BIA	BASTIA-POBETTA, CORSICA, FRANCE	13.343	891.058	113.971			1.325.029	8.568	3	303									0	209,3
2010	BIA	BASTIA-POBETTA, CORSICA, FRANCE	14.418	1.000.207	393.294			1.183.181	7.849	3	303									0	209,3
2015	BIA	BASTIA-POBETTA, CORSICA, FRANCE	14.043	1.022.874	387.709			1.190.843	7.447	3	303									0	209,3
2000	BKA	BEAUVAIS-TULLE, FRANCE	4.502	822	386.547			387.189	347	3	2966									0	89,4
2005	BKA	BEAUVAIS-TULLE, FRANCE	13.228	6	1.847.178			1.847.384	396	3	2966									0	89,4
2010	BKA	BEAUVAIS-TULLE, FRANCE	20.538	124.434	2.815.134			3.429.148	0	3	2966	837	5,5							0	89,4
2015	BKA	BEAUVAIS-TULLE, FRANCE	20.916	137.098	3.794.290			3.953.180	0	3	2966									0	89,4
2010	BKA	BEAUVAIS-TULLE, FRANCE	16.648	79.711	3.764.911			4.023.022	0	3	2966									0	89,4
2015	BKA	BEAUVAIS-TULLE, FRANCE	27.712	87.960	4.241.011			4.328.973	0	3	2966									0	89,4
2000	GGC	BERGERAC-KOLMARIERE, FRANCE	2.746	3.182	250.588			233.740	0	2	6083									0	86,4
2010	GGC	BERGERAC-KOLMARIERE, FRANCE	3.375	3.244	256.479			259.723	0	2	6083	884	8,4							0	86,4
2015	GGC	BERGERAC-KOLMARIERE, FRANCE	3.431	3.209	284.020			286.226	0	2	6083									0	86,4
2010	GGC	BERGERAC-KOLMARIERE, FRANCE	3.296	2.803	274.509			277.112	0	2	6083									0	86,4
2015	GGC	BERGERAC-KOLMARIERE, FRANCE	2.403	1.088	280.338			281.456	0	2	6083									0	86,4
2010	B2N	BIEZERS-VAS, FRANCE	3.189	395	330.179			330.174	0	3	4630	509	7,4							0	53,1
2015	B2N	BIEZERS-VAS, FRANCE	3.853	812.466	375.638			328.024	0	3	4630									0	53,1
2010	B2N	BIEZERS-VAS, FRANCE	3.839	53.191	370.589			343.180	0	3	4630									0	53,1
2015	B2N	BIEZERS-VAS, FRANCE	3.753	17.737	387.287			240.284	0	3	4630									0	53,1
2000	B4D	BIARRITZ-BAYONNE-ANGLET, FRANCE	10.001	686.675	92.140			778.812	644	3	6083									0	170
2000	B4D	BIARRITZ-BAYONNE-ANGLET, FRANCE	9.939	652.556	383.843			836.199	357	3	6083									0	170
2010	B4D	BIARRITZ-BAYONNE-ANGLET, FRANCE	9.938	785.884	203.268			989.152	0	3	6083	388	8,4							0	170
2015	B4D	BIARRITZ-BAYONNE-ANGLET, FRANCE	9.547	809.777	237.399			1.087.176	0	3	6083									0	170
2010	B4D	BIARRITZ-BAYONNE-ANGLET, FRANCE	9.984	809.309	237.802			1.083.738	0	3	6083									0	170
2015	B4D	BIARRITZ-BAYONNE-ANGLET, FRANCE	9.742	822.190	229.774			1.039.246	0	3	6083									0	170
2000	BOC	BORDEAUX-MERIGNAC, FRANCE	50.084	2.522.411	685.209			3.087.710	17.753	3	6083									3	38,8
2005	BOC	BORDEAUX-MERIGNAC, FRANCE	49.024	2.145.838	948.515			3.062.180	15.797	3	6083									3	38,8
2010	BOC	BORDEAUX-MERIGNAC, FRANCE	46.467	2.261.801	3.948.448			4.142.127	11.443	3	6083	515	8,4							3	38,8
2015	BOC	BORDEAUX-MERIGNAC, FRANCE	49.519	2.794.092	3.780.254			4.374.148	12.390	3	6083									3	38,8
2010	BOC	BORDEAUX-MERIGNAC, FRANCE	49.079	2.184.242	2.130.850			4.154.440	12.020	3	6083									3	38,8
2015	BOC	BORDEAUX-MERIGNAC, FRANCE	51.619	2.834.722	2.458.831			3.884.636	11.839	3	6083									3	38,8

Figure: Database used.

Sample size

42 airports, 6 years (2000, 2005, 2010, 2013, 2014, 2015).

Expected effects

Explanatory variable	Expected effect
City (=0 or 1)	+
City's inhabitants	+
Region's GDP	+
Distance to the city	?
TGV	-
Number of tourist attractions	+
Airport operator	?

What do our data give when focusing only on the year 2010?

We are looking into a linear problem of the following form:

$$Y = \beta_1 POP_VILLE + \beta_2 VILLE + \beta_3 PIB_REG + \beta_4 DISTANCE_VILLE \\ + \beta_5 TGV + \beta_6 (MGT = 1) + \beta_7 (MGT = 2) \\ + \beta_8 (MGT = 4) + \beta_9 (MGT = 5) + u$$

Results thanks to EViews

Equation: EQ_TOT Workfile: DATA_2010::Data_2010\

View Proc Object Print Name Freeze Estimate Forecast Stats Resids

Dependent Variable: PAX_TOT
 Method: Least Squares
 Date: 04/05/16 Time: 11:10
 Sample: 1 43
 Included observations: 42

	Coefficient	Std. Error	t-Statistic	Prob.
C	-628678.2	4945229.	-0.127128	0.8997
DISTANCE_VILLE	3554.418	18426.68	0.192895	0.8484
VILLE	3339828.	2975391.	1.122484	0.2709
MONUMENTS_REG	-63.43750	536.9048	-0.118154	0.9058
MONUMENTS_DEP	241.7575	3812.971	0.063404	0.9499
NUITS	254629.9	279174.7	0.912081	0.3692
PIB_REG	10.87157	31.25903	0.347790	0.7305
POP_REGION	-0.614026	1.324479	-0.463598	0.6464
TGV	228907.8	2317966.	0.098754	0.9220
MGT=1	22257689	18233067	1.220732	0.2320
MGT=2	227803.7	2189697.	0.104034	0.9179
MGT=4	-1056725.	2505064.	-0.421836	0.6763
MGT=5	-1294647.	3718799.	-0.348136	0.7303
R-squared	0.839127	Mean dependent var		3357686.
Adjusted R-squared	0.772559	S.D. dependent var		9674015.
S.E. of regression	4613618.	Akaike info criterion		33.77560
Sum squared resid	6.17E+14	Schwarz criterion		34.31345
Log likelihood	-696.2875	Hannan-Quinn criter.		33.97274
F-statistic	12.60551	Durbin-Watson stat		3.007259
Prob(F-statistic)	0.000000			

Conclusion

None of our variables is significant.

What do our data give when focusing only on the year 2010?

We are now looking into a linear problem with variables non linearly entered:

$$Y = \beta_1 \log(\text{POP_VILLE}) + \beta_2 \text{VILLE} + \beta_3 \log(\text{PIB_REG}) \\ + \beta_4 \log(\text{DISTANCE_VILLE}) + \beta_5 \text{TGV} + \beta_6 (\text{MGT} = 1) \\ + \beta_7 (\text{MGT} = 2) + \beta_8 (\text{MGT} = 4) + \beta_9 (\text{MGT} = 5) + u$$

Results thanks to EViews

Equation: LOG_TOT Workfile: DATA_2010::Data_2010\

View Proc Object Print Name Freeze Estimate Forecast Stats Resids

Dependent Variable: LOG(PAX_TOT)
 Method: Least Squares
 Date: 04/06/16 Time: 08:31
 Sample: 1 43
 Included observations: 42

	Coefficient	Std. Error	t-Statistic	Prob.
C	14.22075	4.432292	3.208441	0.0030
LOG(DISTANCE_VILLE)	-0.216534	0.317244	-0.682547	0.4998
VILLE	1.599334	0.858476	1.862992	0.0717
LOG(MONUMENTS_REG)	-0.301130	0.304930	-0.987537	0.3308
LOG(MONUMENTS_DEP)	-0.513651	0.401251	-1.280123	0.2097
LOG(POP_REGION)	0.358125	0.326446	1.097042	0.2808
MGT=1	2.844002	1.109702	2.562851	0.0153
MGT=2	0.019762	0.478253	0.041320	0.9673
MGT=4	-0.711651	0.546759	-1.301581	0.2024
MGT=5	-0.905113	0.753012	-1.201991	0.2382
R-squared	0.705424	Mean dependent var		13.36189
Adjusted R-squared	0.622574	S.D. dependent var		1.648424
S.E. of regression	1.012709	Akaike info criterion		3.067392
Sum squared resid	32.81854	Schwarz criterion		3.481122
Log likelihood	-54.41522	Hannan-Quinn criter.		3.219040
F-statistic	8.514518	Durbin-Watson stat		1.790454
Prob(F-statistic)	0.000002			

Conclusion

Lot of insignificant variables : only two variables are significant (VILLE and (MGT=1)).

Model taking data from each year into account

Model (variables entered non linearly)

$$\begin{aligned} Y = & \beta_1 \log(\text{POP_VILLE}) + \beta_2 \text{VILLE} + \beta_3 \log(\text{PIB_REG}) \\ & + \beta_4 \log(\text{DISTANCE_VILLE}) + \beta_5 \log(\text{MONUMENTS_REG}) \\ & + \beta_6 \text{TGV} + \beta_7 (\text{MGT} = 1) + \beta_8 (\text{MGT} = 2) \\ & + \beta_9 (\text{MGT} = 4) + \beta_{10} (\text{MGT} = 5) + u \end{aligned}$$

Results thanks to EViews

Equation: LOG_TOT Workfile: APT_FRANCE00_15_REGION-1::Apt_france00_15_re...

View Proc Object Print Name Freeze Estimate Forecast Stats Resids

Dependent Variable: LOG(PAX_TOT)
Method: Least Squares
Date: 04/06/16 Time: 09:25
Sample (adjusted): 1 251
Included observations: 118 after adjustments

	Coefficient	Std. Error	t-Statistic	Prob.
C	13.59837	1.703963	7.980435	0.0000
TGV	-0.032496	0.250791	-0.129575	0.8971
VILLE	0.812102	0.447356	1.815337	0.0723
LOG(DISTANCE_VILLE)	-0.306837	0.153585	-1.997828	0.0483
LOG(MONUMENTS_REG)	-0.588445	0.142139	-4.139931	0.0001
LOG(POP_VILLE)	0.373419	0.096129	3.884561	0.0002
LOG(PIB_REG)	0.125557	0.137278	0.914617	0.3624
MGT=1	1.746085	0.581068	3.004955	0.0033
MGT=2	-0.063406	0.241908	-0.262109	0.7937
MGT=4	-0.551257	0.260377	-2.117149	0.0366
MGT=5	-1.104908	0.422884	-2.612794	0.0103

R-squared	0.744920	Mean dependent var	13.42777
Adjusted R-squared	0.721081	S.D. dependent var	1.613946
S.E. of regression	0.852370	Akaike info criterion	2.606994
Sum squared resid	77.73927	Schwarz criterion	2.865278
Log likelihood	-142.8126	Hannan-Quinn criter.	2.711865
F-statistic	31.24759	Durbin-Watson stat	0.263705
Prob(F-statistic)	0.000000		

Figure: Screenshot of the results given by EViews.

Marginal effects

Signs

Explanatory variable	Expected effect	Observed effect
City (=0 or 1)	+	+
City's inhabitants	+	+
Region's GDP	+	+
City's distance	?	-
TGV	-	-
Number of tourist attractions	+	-
Airport operator	?	+/-

Case of Airport operator

The marginal effects are observed comparing airports operated by CCI to other forms of management.

Validity of the first model

Fitness

$$R^2 = 72,1\%$$

Wald's test

We have to test if we can reject the hypothesis :

$$H_0 : \beta_i = 0, i = 1, 2, \dots, 8$$

with a level of certainty of 90%.

Wald's tests

Variable	t-Statistic	Prob.
C	7,9804	0
TGV	-0,1295	0,8971
VILLE	1,8115	0,0723
LOG(DISTANCE_VILLE)	-1,9978	0,0483
LOG(MONUMENTS_REG)	-4,1399	0,00001
LOG(POP_VILLE)	3,8845	0,002
LOG(PIB_REG)	0,9146	0,3624
MGT=1	3,0049	0,0033
MGT=2	-0,2621	0,7937
MGT=4	-2,1171	0,0366
MGT=5	-2,96127	0,0103

We can drop the variables with the highest probability (do not reject the null hypothesis) :

- TGV
- LOG(PIB_REG)
- (MGT=2)

Figure: Results of the Wald's tests

Conclusion on the validity of the model

Possible improvements

We can drop the variables TGV and PIB_REG.

Equation: LOG_TOT_RESTRICTED Workfile: APT_FRANCE00...

View Proc Object Print Name Freeze Estimate Forecast Stats Resids

Dependent Variable: LOG(PAX_TOT)
Method: Least Squares
Date: 04/06/16 Time: 14:33
Sample (adjusted): 1 251
Included observations: 118 after adjustments

	Coefficient	Std. Error	t-Statistic	Prob.
C	14.49416	1.412155	10.26386	0.0000
VILLE	0.784899	0.440558	1.781602	0.0776
LOG(DISTANCE_VILLE)	-0.338571	0.149137	-2.270198	0.0252
LOG(MONUMENTS_REG)	-0.516634	0.116172	-4.447156	0.0000
LOG(POP_VILLE)	0.375269	0.094273	3.980658	0.0001
MGT=1	1.963181	0.487333	4.028416	0.0001
MGT=2	-0.057874	0.236894	-0.244303	0.8075
MGT=4	-0.493372	0.251068	-1.965091	0.0519
MGT=5	-1.041208	0.364292	-2.858169	0.0051

R-squared	0.742767	Mean dependent var	13.42777
Adjusted R-squared	0.723888	S.D. dependent var	1.613946
S.E. of regression	0.848070	Akaike info criterion	2.581499
Sum squared resid	78.39528	Schwarz criterion	2.792822
Log likelihood	-143.3084	Hannan-Quinn criter.	2.667302
F-statistic	39.34260	Durbin-Watson stat	0.263571
Prob(F-statistic)	0.000000		

Figure: Screenshot of the results given by EViews.

Improvements

Fitness

$$R^2 = 72.4\% \geq R^2_{firstmodel}$$

Insignificant variables

We only have (MGT=2)(operated by airports' companies) as an insignificant variable

Correlation

Correlation				
	DISTANCE_...	MONUMENT...	POP_VILLE	VILLE
DISTANCE_...	1.000000	-0.308120	-0.439889	-0.714227
MONUMENT...	-0.308120	1.000000	0.062244	0.032905
POP_VILLE	-0.439889	0.062244	1.000000	0.576017
VILLE	-0.714227	0.032905	0.576017	1.000000

Figure: Screenshot of the correlation matrix.

Problem of the variable DISTANCE_VILLE

Correlated with MONUMENTS_DEP, VILLE and POP_VILLE, but needed in order to observe the effect of the proximity of a major city.

Classical assumptions (1/4)

Linearity

The model fits the data (fitted-R squared) and the model is meaningful (global F-test).

Classical assumptions (2/4)

Strict exogeneity

$$E(u) = 0$$

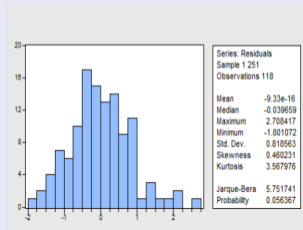


Figure: Histogram of residuals given by EViews.

We have :

$$E(u) \approx 0$$

The classical assumption is held.

Classical assumptions(3/4)

Homoscedasticity

Heteroskedasticity Test: White

F-statistic	1.465926	Prob. F(29,88)	0.0891
Obs*R-squared	38.43635	Prob. Chi-Square(29)	0.1129
Scaled explained SS	42.11096	Prob. Chi-Square(29)	0.0549

Test Equation:

Dependent Variable: RESID^2
Method: Least Squares
Date: 04/06/16 Time: 14:36
Sample: 1 251
Included observations: 118
Collinear test regressors dropped from specification

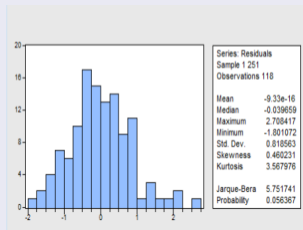
We do not reject the null hypothesis (no heteroscedasticity) at a level of significance of 95 %.

Homoscedasticity of the model.

Figure: White's test with EViews

Classical assumptions(4/4)

Normality of the residuals (Jarque-Bera test)



$$Prob = 5.6\%$$

We do not reject the null hypothesis of normality at a level of significance of 95%.

Figure: Histogram of residuals given by EViews.

Strengths and weaknesses

Strengths

- Good model which fits the data while respecting classical assumptions
- Almost all marginal effects correspond to intuition

Weaknesses

- Some variables collected do not seem relevant (TGV and PIB_REG)
- Problem of the sign of the marginal effect MONUMENTS_REG

Explanations

- Different politics of airports' companies
- For the number of tourist attractions, categorization not precise enough
- A lot of other factors which are difficult to estimate have to be taken into account

Comparison between the first and the logarithm model

Variable	t-Statistic	Prob.
C	-0,3458	0,7302
TGV	-0,5264	0,5997
VILLE	0,8349	0,4056
DISTANCE_VILLE	0,3554	0,723
MONUMENTS_REG	-0,5033	0,6158
POP_VILLE	2,075	0,0404
PIB_REG	1,558	0,1223
MGT=1	2,3382	0,0212
MGT=2	0,2967	0,7673
MGT=4	-0,9205	0,3594
MGT=5	-0,1827	0,8554

Figure: Results for the first model

Variable	t-Statistic	Prob.
C	7,9804	0
TGV	-0,1295	0,8971
VILLE	1,8115	0,0723
LOG(DISTANCE_VILLE)	-1,9978	0,0483
LOG(MONUMENTS_REG)	-4,1399	0,00001
LOG(POP_VILLE)	3,8845	0,002
LOG(PIB_REG)	0,9146	0,3624
MGT=1	3,0049	0,0033
MGT=2	-0,2621	0,7937
MGT=4	-2,1171	0,0366
MGT=5	-2,96127	0,0103

Figure: Results for the log model

Conclusion

With the exception of TGV, MONUMENTS_REG and MGT=2, we have the same signs for the marginal effect.

The logarithm model presents fewer insignificant variables.