

BANK OF ENGLAND

Appendix to Staff Working Paper No. 856 High water, no marks? Biased lending after extreme weather

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Table 1: Setting up the dataset

Panel A. Housing market dataset

Step	Observations
Constructing the stock of properties transacted in the 3-year pre-event window Population of all property transactions in the affected areas since 1999 Dropping duplicate transactions	4,954,734
Dropping properties that are transacted during the flood event period Dropping properties that are house prices outliers (above GBP 10 million)	
Dropping properties that are new built after the flood event period	
Keeping only properties which were transacted in the three year window	
After keeping only one transaction per pre- vs. post-event window	$623,\!328$
Constructing the house transaction panel	
 After constructing the panel of transactions (assign "0" to non-transacted)	$1,\!085,\!672$
Panel B. Mortgage market dataset	
Step	Observations
Constructing the stock of mortgages transacted in the 3-year pre-event window Population of the flow of mortgage transactions from PSD 001 since 2005 Adding the "internal" refinance from the stock of mortgages in 2017 (PSD 007) Dropping mortgages that are transacted during the flood event period	3,743,110
Dropping duplicate transactions	
Dropping all mortages by first time buyers after the $2013/14$ winter flood	
Keeping only mortgages which were transacted before the event Dropping mortgages which were refinancing transactions before the event	
Keeping only one transaction per pre- vs. post-event window	
After dropping movers after the $2013/14$ winter flood	416,917
Constructing the mortgage transaction panel	
After constructing the panel of transactions (assign "0" to non-transacted)	$736,\!924$
Panel C. Combined dataset	
Step	Observations
Constructing the panel of property and mortgage transactions	
\dots Appending the panel of property transactions and mortgage transactions	$1,\!822,\!596$
Constructing the repeated transaction dataset	
Keeping only repeated transactions	110.000
After keeping only the transactions in the second event window	119,239

This table describes the individual steps of setting up the final dataset. Panel A shows individual steps of setting up the panel of property transactions. Panel B shows individual steps of setting up the panel of mortgage transactions. Panel C describes individual steps of setting up the combined dataset.

Dependent variable	Flood	Risk up
Column	(1)	(2)
		()
L.Property value (Ln)	0.003***	-0.004
	(0.001)	(0.004)
Distance to water (Ln)	-0.001**	0.001
	(0.000)	(0.002)
Distance to water $(<100m)$	-0.000	0.011^{***}
	(0.001)	(0.004)
Distance to water $(<200m)$	0.002^{***}	0.007^{**}
	(0.001)	(0.003)
Distance to water $(<500m)$	0.003^{***}	0.002
	(0.001)	(0.003)
Risk (EA) (2013)	0.005^{***}	0.034^{***}
	(0.001)	(0.004)
Matrix Scores (JBA) (2013)	0.000^{***}	0.002^{***}
	(0.000)	(0.000)
Observations	$74,\!921$	77,234
Pseudo R2	0.339	0.262
Mean of dependent variable	0.012	0.042

Table 2: Logit estimates: housing market

Logit regressions estimated using ML. Marginal effects are reported at the mean. Standard errors clustered at the postcode district and reported in brackets. Standard errors clustered at the postcode district and reported in brackets. Stars denote statistical significance at the 0.01 & 0.05 & 0.10-level respectively.

Dependent variable	Flood	Risk up
Column	(1)	(2)
L.Property value (Ln)	0.003***	-0.002
	(0.001)	(0.003)
Distance to water (Ln)	-0.001	0.000
	(0.000)	(0.002)
Distance to water $(<100m)$	0.000	0.014^{***}
	(0.001)	(0.004)
Distance to water $(<200m)$	0.002^{***}	0.006^{*}
	(0.001)	(0.004)
Distance to water $(<500m)$	0.002	0.001
	(0.002)	(0.003)
Risk (EA) (2013)	0.005^{***}	0.030***
	(0.001)	(0.003)
Matrix Scores (JBA) (2013)	0.000***	0.001***
	(0.000)	(0.000)
Observations	$36,\!563$	$37,\!475$
Pseudo R2	0.344	0.268
Mean of dependent variable	0.013	0.037

Table 3: Logit estimates: mortgage market

Logit regressions estimated using ML. Marginal effects are reported at the mean. Standard errors clustered at the postcode district and reported in brackets. Standard errors clustered at the postcode district and reported in brackets. Stars denote statistical significance at the 0.01 & 0.05 & 0.10-level respectively.

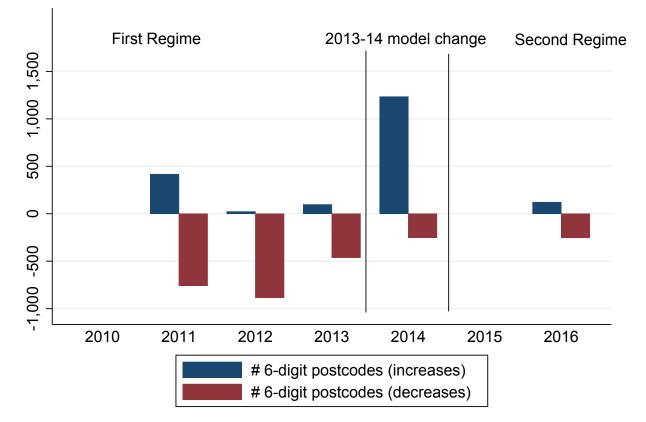


Figure 1: Number of reclassified 6-digit postcodes

Large increase or decrease in flood risk is defined as at least .8 percentage point change.

These graph shows the number of reclassified 6-digit postcodes over time.

