

How Does Monetary Policy Affect Income and Wealth Inequality? Evidence from Quantitative Easing in the Euro Area Online Appendix

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Abstract

This paper evaluates the impact of quantitative easing on income and wealth of individual euro area households. We first estimate the aggregate effects of a QE shock, identified by means of external instruments, in a multi-country VAR model with unemployment, wages, interest rates, house prices and stock prices. We then distribute the aggregate effects across households using a reduced-form simulation on micro data, which captures the portfolio composition, the income composition and the earnings heterogeneity channels of transmission. The earnings heterogeneity channel is important: QE compresses the income distribution since many households with lower incomes become employed. In contrast, monetary policy has only negligible effects on the Gini coefficient for wealth: while high-wealth households benefit from higher stock prices, middle-wealth households benefit from higher house prices.

Keywords Monetary Policy, Household Heterogeneity, Inequality, Income, Wealth, Quantitative Easing, Great Recession

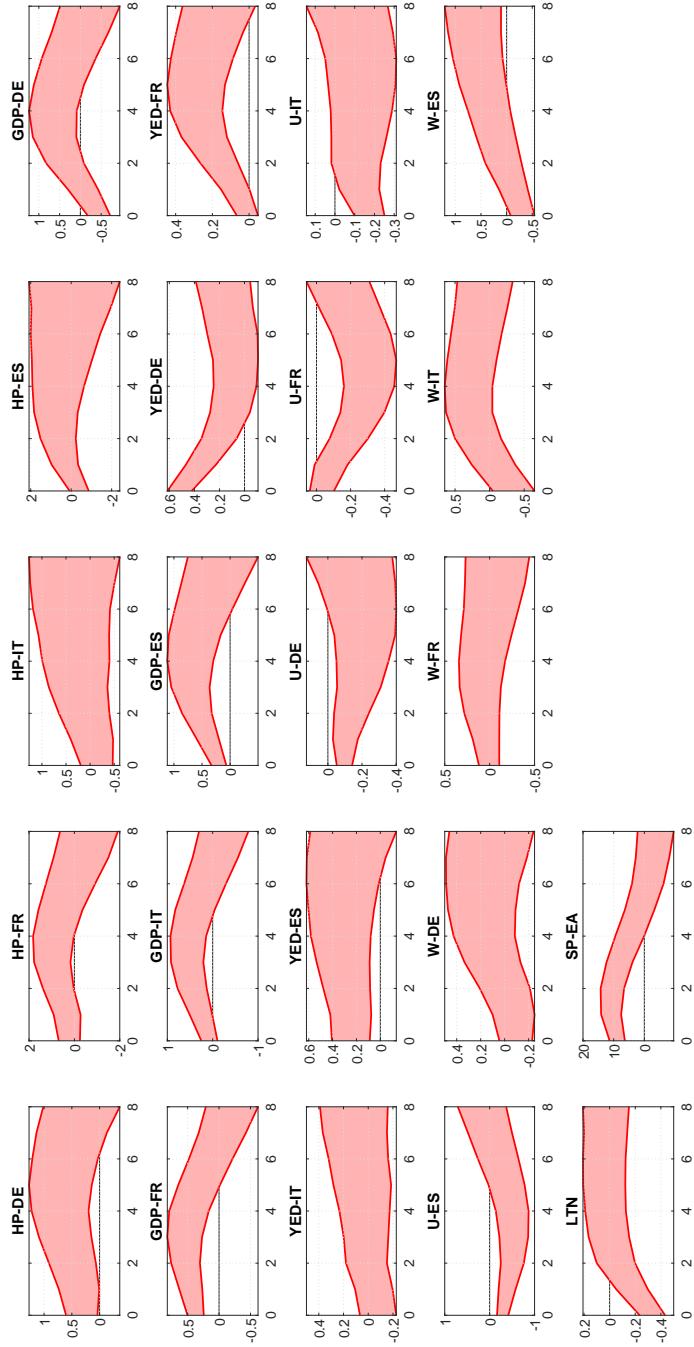
JEL codes D14, D31, E44, E52, E58

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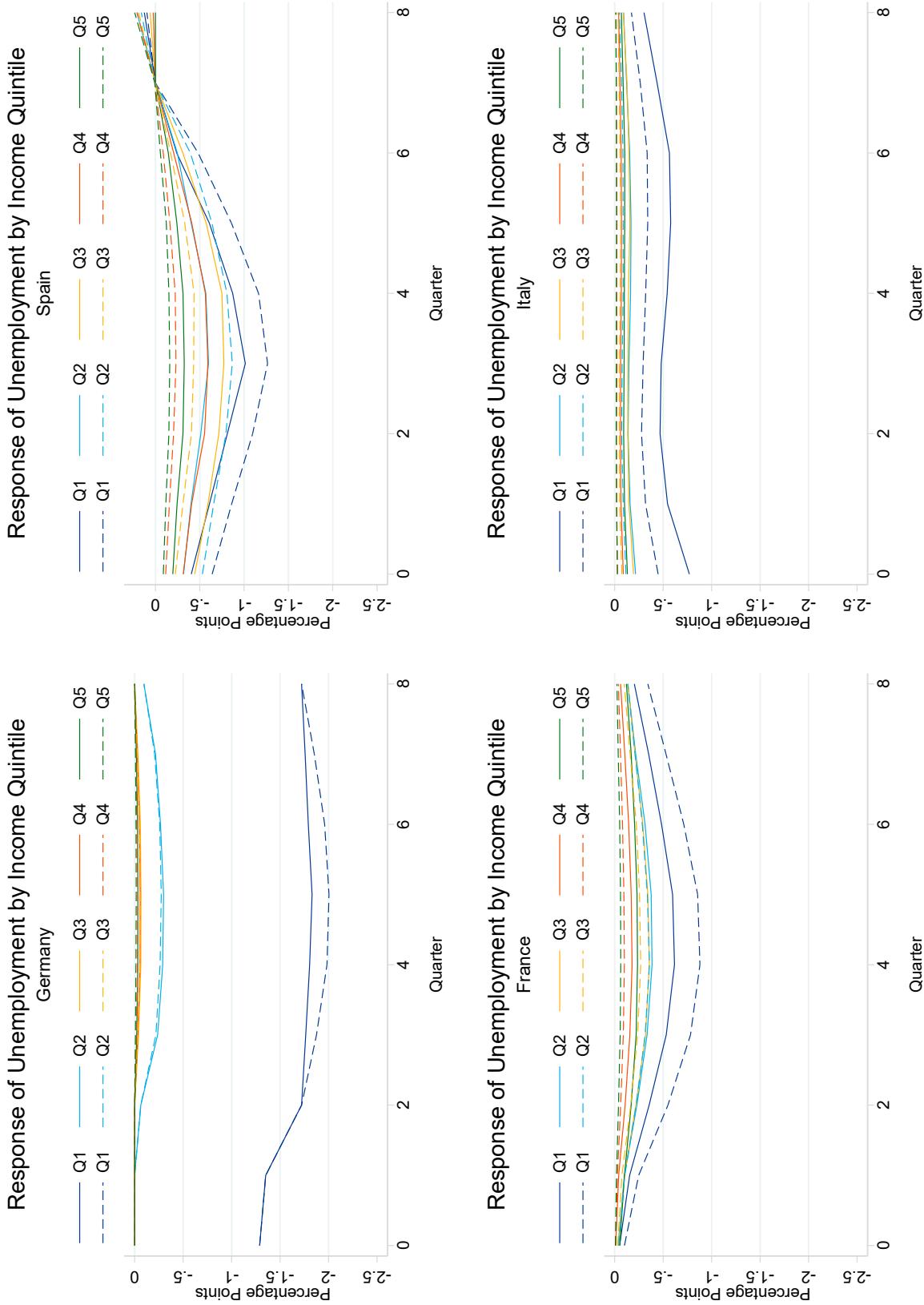
Appendix C: Additional Figures and Tables—For Online Publication

Figure C.1 Impulse Responses to QE Shock



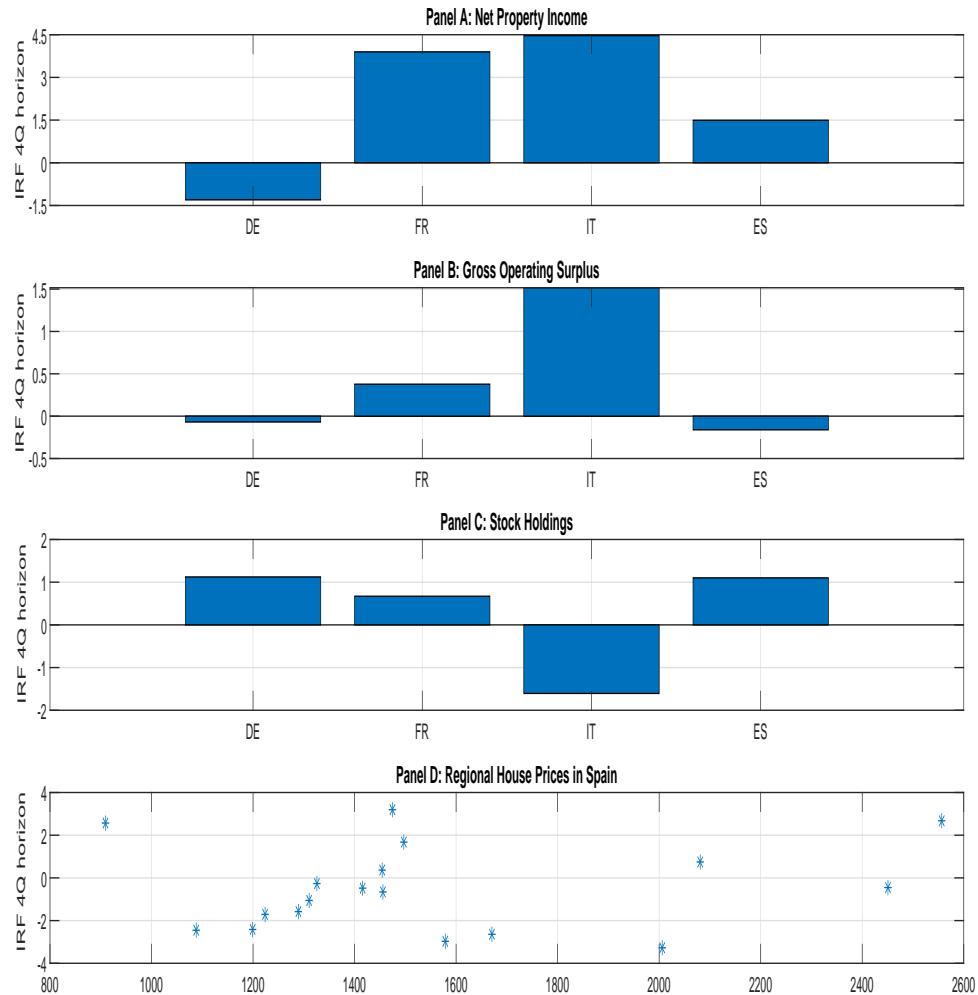
Note: The figure shows the impulse response of all the variables in the model to the QE shock (30 bp drop in the term spread). The red shaded area reflects the 16th–84th percentile range. HP: house prices; GDP: real gross domestic product; YED: GDP deflator; U: unemployment rate; W: compensation per employee, wage; LTN: nominal long-term interest rate; SP: stock prices. EA: euro area; DE: Germany; FR: France; IT: Italy; ES: Spain.

Figure C.2 Impulse Responses of Unemployment—Baseline IRFs (Solid) vs IRFs Generated Under Uniform Probability of Getting Employed (Dashed)



Source: Household Finance and Consumption Survey
Note: The charts show impulse responses of unemployment by income quintile.

Figure C.3 Robustness exercises



Note: The figure reports the median of the impulse responses (four quarters after the shock) of the additional variables used in the robustness checks. The responses are estimated by means of the local linear projection method of Jordà (2005) applied to the median QE shock estimated in the VAR. Units: percentage point deviation from pre-shock levels. Panel D shows a scatter plot of impulse responses of regional house prices across provinces in Spain across house prices per square meter (in EUR). DE: Germany; FR: France; IT: Italy; ES: Spain.

Probit and Heckman estimates

Table 1 Probit Estimation Results—Germany

Variable	Coefficient	(Std. Err.)
gender	-0.047	(0.060)
college	0.962**	(0.100)
highschool	0.539**	(0.087)
age2	-0.052	(0.098)
age3	0.034	(0.088)
age456	-0.170 ^y	(0.091)
single	-0.348**	(0.072)
children	0.104	(0.076)
Intercept	1.136**	(0.111)

N	4868
Log-likelihood	-1020.965
$\chi^2_{(8)}$	172.886

Significance levels : \circ : 10% * : 5% ** : 1%

Table 2 Probit Estimation Results—Spain

Variable	Coefficient	(Std. Err.)
gender	0.178**	(0.035)
college	0.695**	(0.040)
highschool	0.335**	(0.048)
age2	0.264**	(0.055)
age3	0.259**	(0.053)
age456	0.415**	(0.055)
single	-0.400**	(0.045)
children	0.068	(0.045)
Intercept	0.211**	(0.061)

N	6815
Log-likelihood	-3320.85
$\chi^2_{(8)}$	619.444

Significance levels : \circ : 10% * : 5% ** : 1%

Table 3 Probit Estimation Results—France

Variable	Coefficient	(Std. Err.)
gender	0.006	(0.030)
college	0.809**	(0.041)
highschool	0.401**	(0.039)
age2	0.394**	(0.043)
age3	0.552**	(0.042)
age456	0.504**	(0.047)
single	-0.368**	(0.034)
children	0.110**	(0.035)
Intercept	0.509**	(0.053)

N	13408
Log-likelihood	-4291.914
$\chi^2_{(8)}$	1026.339

Significance levels : ý : 10% * : 5% ** : 1%

Table 4 Probit Estimation Results—Italy

Variable	Coefficient	(Std. Err.)
gender	0.106**	(0.035)
college	0.805**	(0.054)
highschool	0.501**	(0.038)
age2	0.591**	(0.049)
age3	0.857**	(0.050)
age456	0.931**	(0.056)
single	-0.395**	(0.046)
children	0.145**	(0.046)
Intercept	-0.002	(0.061)

N	7979
Log-likelihood	-3552.257
$\chi^2_{(8)}$	1210.823

Significance levels : ý : 10% * : 5% ** : 1%

Table 5 Heckman Estimation Results—Germany

Variable	Coefficient	(Std. Err.)
Equation 1 : log_laborincome		
gender	0.591**	(0.028)
college	1.021**	(0.125)
highschool	0.579**	(0.099)
age2	0.428**	(0.046)
age3	0.500**	(0.044)
age456	0.459**	(0.041)
Intercept	8.912**	(0.155)
Equation 2 : job		
gender	-0.031	(0.061)
college	1.016**	(0.102)
highschool	0.588**	(0.089)
age2	-0.030	(0.099)
age3	0.064	(0.089)
age456	-0.150	(0.092)
single	-0.348**	(0.072)
children	0.099	(0.077)
Intercept	1.038**	(0.115)
Equation 3 : /mills		
lambda	-0.778*	(0.377)
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N	4650	
Log-likelihood	.	
$\chi^2_{(6)}$	760.647	
Significance levels : ý : 10% * : 5% ** : 1%		

Table 6 Heckman Estimation Results—Spain

Variable	Coefficient	(Std. Err.)
Equation 1 : log_laborincome		
gender	0.323**	(0.028)
college	0.604**	(0.052)
highschool	0.233**	(0.044)
age2	0.190**	(0.057)
age3	0.426**	(0.057)
age456	0.568**	(0.062)
Intercept	9.300**	(0.127)

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... table 6 continued

Variable	Coefficient	(Std. Err.)
Equation 2 : job		
gender	0.187**	(0.036)
college	0.712**	(0.041)
highschool	0.341**	(0.049)
age2	0.360**	(0.056)
age3	0.364**	(0.055)
age456	0.518**	(0.057)
single	-0.428**	(0.046)
children	0.086 ^y	(0.046)
Intercept	0.060	(0.063)
Equation 3 : /mills		
lambda	-0.859**	(0.138)
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N	6365	
Log-likelihood	.	
$\chi^2_{(6)}$	278.775	
Significance levels : ^y : 10% * : 5% ** : 1%		

Table 7 Heckman Estimation Results—France

Variable	Coefficient	(Std. Err.)
Equation 1 : log_laborincome		
gender	0.414**	(0.017)
college	0.593**	(0.043)
highschool	0.094**	(0.033)
age2	0.248**	(0.035)
age3	0.361**	(0.038)
age456	0.465**	(0.037)
Intercept	9.454**	(0.076)
Equation 2 : job		
gender	-0.001	(0.031)
college	0.866**	(0.042)
highschool	0.442**	(0.040)
age2	0.447**	(0.044)
age3	0.613**	(0.042)
age456	0.563**	(0.048)
single	-0.380**	(0.034)
children	0.142**	(0.036)
Intercept	0.385**	(0.055)

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Variable	Coefficient	(Std. Err.)
Equation 3 : /mills		
lambda	-0.808**	(0.119)
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N 12753		
Log-likelihood .		
$\chi^2_{(6)}$ 1124.399		
Significance levels : γ : 10% * : 5% ** : 1%		

Table 8 Heckman Estimation Results—Italy

Variable	Coefficient	(Std. Err.)
Equation 1 : log_laborincome		
gender	0.376**	(0.017)
college	0.580**	(0.036)
highschool	0.265**	(0.026)
age2	0.263**	(0.045)
age3	0.446**	(0.052)
age456	0.468**	(0.054)
Intercept	9.179**	(0.086)
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Equation 2 : job		
gender	0.105**	(0.035)
college	0.806**	(0.054)
highschool	0.502**	(0.038)
age2	0.591**	(0.049)
age3	0.856**	(0.050)
age456	0.931**	(0.057)
single	-0.394**	(0.046)
children	0.146**	(0.046)
Intercept	-0.004	(0.061)
<hr/>		
Equation 3 : /mills		
lambda	-0.273**	(0.095)
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N 7964		
Log-likelihood .		
$\chi^2_{(6)}$ 634.546		
Significance levels : γ : 10% * : 5% ** : 1%		

Table C.1 Estimates of the Effects of Nonstandard Monetary Policy Using Event Studies

Authors	Country	Type of Event	Typical Impact on 10-Year Rate (p.p.)	Notes
Altavilla et al. (2016)	DE, ES, FR, IT	OMT	0.2 to 1	
Altavilla et al. (2015)	EA, DE, ES, FR, IT	APP	0.3 to 0.5	
Andrade et al. (2016)	EA	APP	0.45	
Joyce and Tong (2012)	UK	APF1	1	
Christensen and Rudebusch (2012)	UK, US	APF1	0.43 to 0.89	
Lam (2011)	JP	CME+	0.24 to 0.27	
Fukunaga et al. (2015)	JP	QQE	0.33 to 0.47	
Gagnon et al. (2011)	US	LSAP1	0.55 to 1.05	
Krishnamurthy and Vissing-Jorgensen (2013)	US	LSAP1, LSAP2, MEP	0.07 to 1.07	
Bauer and Rudebusch (2014)	US	LSAP1	0.89	
Krishnamurthy and Vissing-Jorgensen (2011)	US	LSAP1, LSAP2	0.3 to 1.07	
Cahill et al. (2013)	US	LSAP1, LSAP2, MEP	0.089 to 0.131	for \$100bn purchases

Notes: See also Andrade et al. (2016), Appendix B for other studies and details. Abbreviations: OMT—Outright Monetary Transactions (Announcement), APP—Asset Purchases Programmes, APF—Asset Purchase Facility, CME—Comprehensive Monetary Easing, QQE—Quantitative and Qualitative Monetary Easing, LSAP—Large Scale Asset Purchase Program, MEP—Maturity Extension Program.

Table C.2 Estimates of the Effects of Nonstandard Monetary Policy Using VARs

Authors	Method (Country)	Type of Event	Effect on Real Economy and Inflation
Altavilla et al. (2016) Baumeister and Benati (2013)	VAR (DE, ES, FR, IT) TVP VAR (US, UK)	OMT LSAP	Real GDP: 0.34%–2.01%, HICP: 0.28%–1.21% Inflation: trough of –1% to –4% GDP gr: trough –10% to –12%, UR: peak 10.6%
Kapetanios et al. (2012) Weale and Wieladek (2016)	TVP VAR (UK) Bayesian VAR (US, UK)	BoE LSAP LSAP	Real GDP: peak effect of 1.42% Real GDP: 0.25%–0.58%, CPI: 0.32%–0.62%
Gambacorta et al. (2014)	Panel VAR (EA, non-EA countries)	Various	GDP: –0.25% to 0.25%, CPI: –0.12% to 0.10%
Darracq-Paries and De Santis (2015) Babecka Kucharcukova et al. (2016)	Panel VAR (EA countries) VAR (EA, non-EA countries)	3-year LTROs Spillovers from ECB QE	GDP: peak of 0.8%, GDP Defl: peak of 0.35% IP: –0.2% to 0.2%, HICP: –0.1% to 0.06%
Bluwstein and Canova (2016)	Bayesian SVAR (EA, EU countries)	Spillovers from ECB QE	IP: –0.1% to 0%, CPI: 0%–0.5%
Hachula et al. (forthcoming)	SVAR (EA, EA countries)	LTROs	GDP: 0.1%–0.65%, CPI: 0%–0.45%
Behrendt (2017) Boeckx et al. (2017)	SVAR (EA) SVAR (EA, EA countries)	ECB QE 3Y LTRO, CBPP1	UR: –0.21%–0.07% IP –0.0032%–0.0023%, HICP –0.0006%–0.0005% GDP: –0.35%–0.6%, HICP: –0.1%–0.3%

Notes: See also Andrade et al. (2016), Appendix B for other studies and details. Abbreviations: OMT—Outright Monetary Transactions, LSAP—Large Scale Asset Purchase Program, LTROs—long-term refinancing operations, CBPP1—Covered Bond Purchases Program.

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