Submitted to *Econometrica*

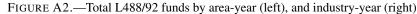
1	ONLINE APPENDIX TO	1
2	"MAKING SUBSIDIES WORK: RULES VS. DISCRETION"	2
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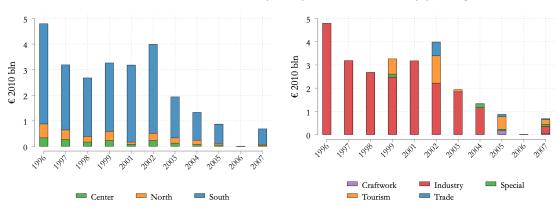
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A. ADDITIONAL FIGURES AND TABLES

FIGURE A1.—L488/92 funds and GDP per capita across regions L488 funds per capita, total 1996-2007 BAS SAR SAR ABR GDP per capita, 1995 Southern regions □ Northern-Center regions

Notes: This figure plots the total amount of L488/92 per capita received over the period 1997-2007 (vertical axis) against the GDP per capita in 1995 (horizontal axis), across Italian regions. Both variables are expressed in euros at constant 2010 prices. The size of markers is proportional to region population.





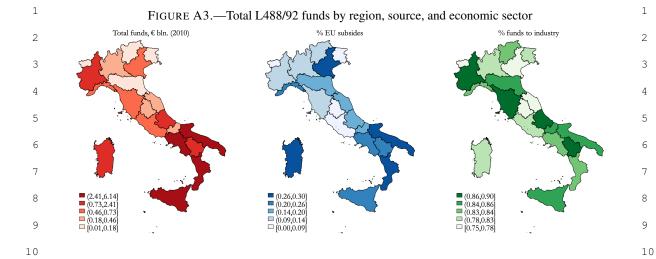
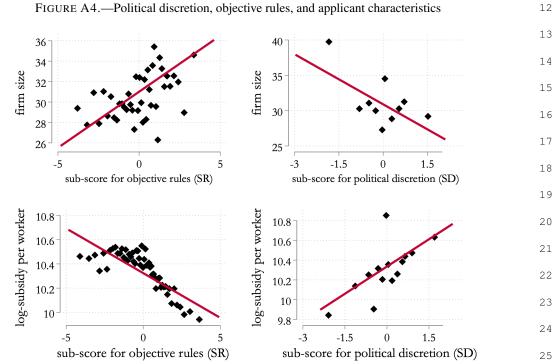
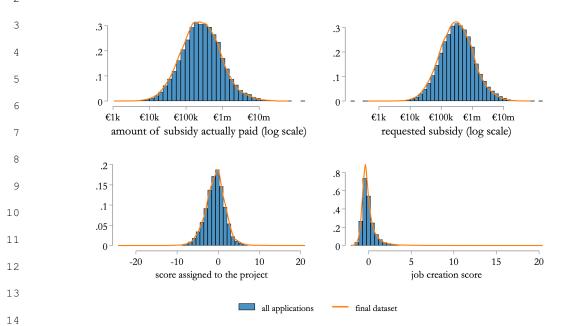


FIGURE A4.—Political discretion, objective rules, and applicant characteristics

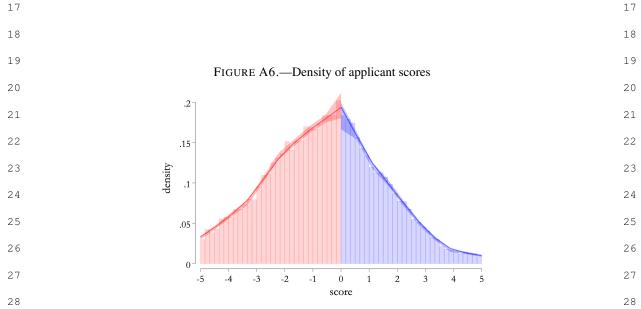


Notes: These graphs plot the sub-scores for political discretion and objective rules (on the horizontal axis), against the size of applicant firms and the amount of subsidies they applied for (on the vertical axis), controlling for cell fixed effects, across quantile-spaced bins. Covariate adjustment and the choice of the optimal number of bins are performed according to Cattaneo, Crump, Farrell & Feng (2024).

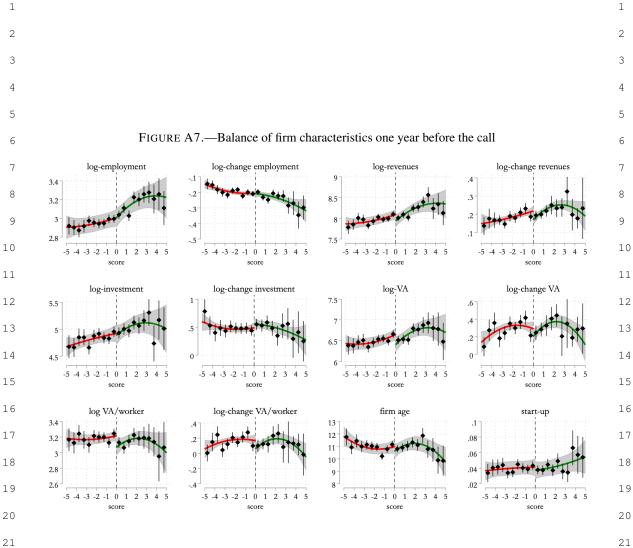
FIGURE A5.—Distribution of selected variables across all applications and within the sub-sample of matched applications



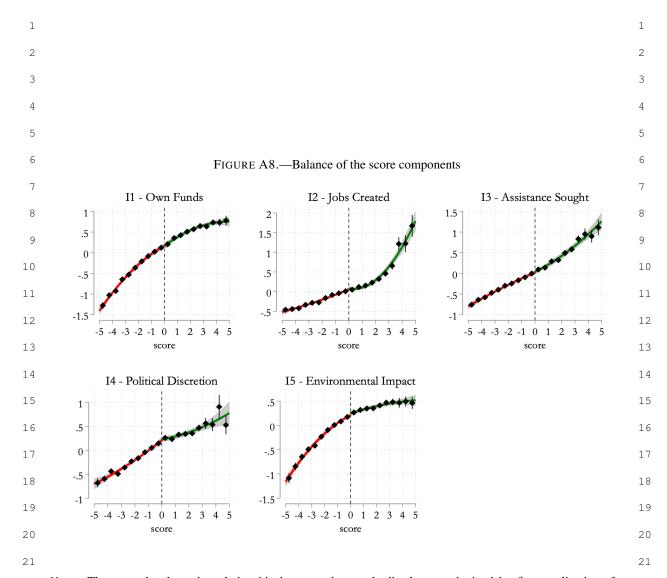
Notes: This figure shows the distribution of some variables across the entire sample of applicants and across the final sample of applicants for which we have complete information on employees and balance sheet data.



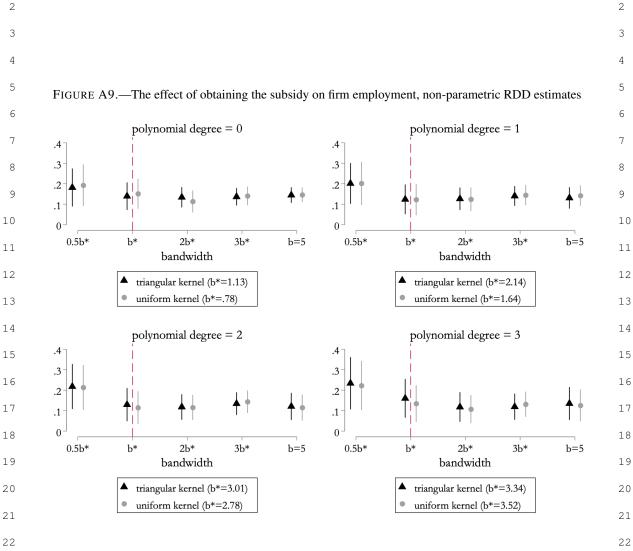
Notes: The histogram shows the distribution of applicant scores. Local polynomial density estimates (solid lines) and robust bias-corrected 95% confidence intervals (shaded areas), computed according to Cattaneo, Jansson & Ma (2020), are also reported in the figure.



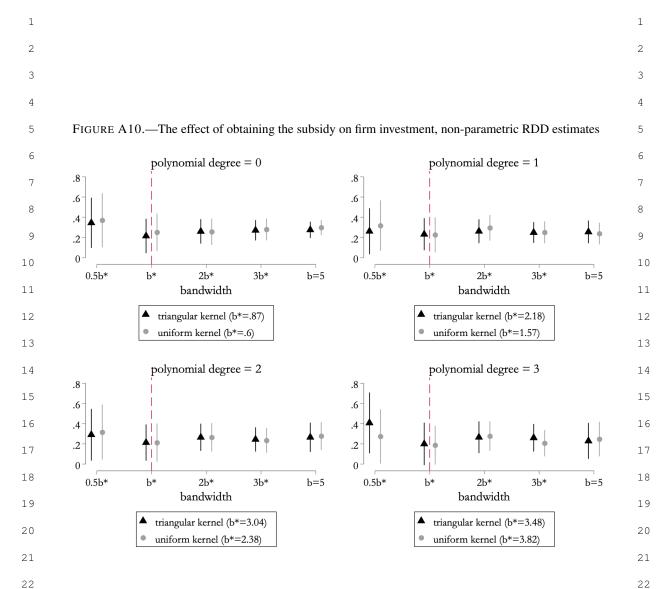
Notes: These graphs show the relationship between the standardized score obtained by firm applications for L488/92 funds, on the horizontal axis, and several firm characteristics measured one year before the call – log and yearly log-change in revenues, value added, value added per worker, investment, firm age and being a start-up. Bins represent averages over equally-spaced intervals, and confidence intervals (at the 90% significance level) are also shown by vertical lines. The predicted relationships between each variable and the score are estimated using a quadratic polynomial regression, controlling for cell-specific fixed effects. 90% confidence bands for the predicted relationship (in gray) are computed based on heteroskedasticity-robust standard errors clustered by cell.



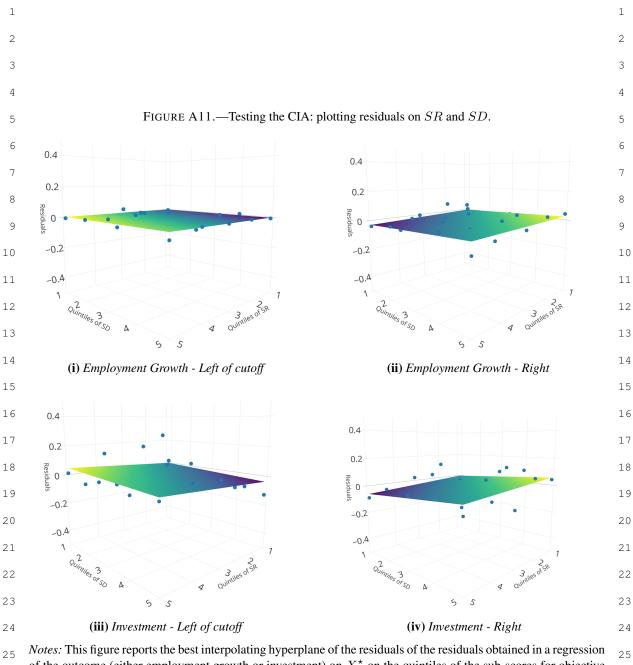
Notes: These graphs show the relationship between the standardized score obtained by firm applications for L488/92 funds, on the horizontal axis, and its five components (described in Section 2 of the main text). Bins represent averages over equally-spaced intervals, and confidence intervals (at the 90% significance level) are also shown by vertical lines. The predicted relationships between each variable and the score are estimated using a quadratic polynomial regression, controlling for cell-specific fixed effects. 90% confidence bands for the predicted relationship (in gray) are computed based on heteroskedasticity-robust standard errors clustered by cell.



Notes: This figure plots the estimated effect of being eligible for the subsidy (i.e., scoring above the cutoff) on the log-change of firm employment 6 years after the award of subsidies, for different specifications of non-parametric RDD. In particular, each graph shows point estimates and confidence intervals when using triangular and uniform kernels, for different degrees of the polynomial in the running variable (reported on top of each graph) and different bandwidths around the cutoff (on the horizontal axis). The optimal bandwidth b^* , as well as point estimates and 90% confidence intervals are computed following the approach proposed in Calonico, Cattaneo & Titiunik (2014).



Notes: This figure plots the estimated effect of being eligible for the subsidy (i.e., scoring above the cutoff) on the log of cumulated investment over the 3 years after the award of subsidies, for different specifications of non-parametric RDD. In particular, each graph shows point estimates and confidence intervals when using triangular and uniform kernels, for different degrees of the polynomial in the running variable (reported on top of each graph) and different bandwidths around the cutoff (on the horizontal axis). The optimal bandwidth b^* , as well as point estimates and 90% confidence intervals are computed following the approach proposed in Calonico, Cattaneo & Titiunik (2014).



Notes: This figure reports the best interpolating hyperplane of the residuals of the residuals obtained in a regression of the outcome (either employment growth or investment) on X^* on the quintiles of the sub-scores for objective rules (SR) and political discretion (SD).

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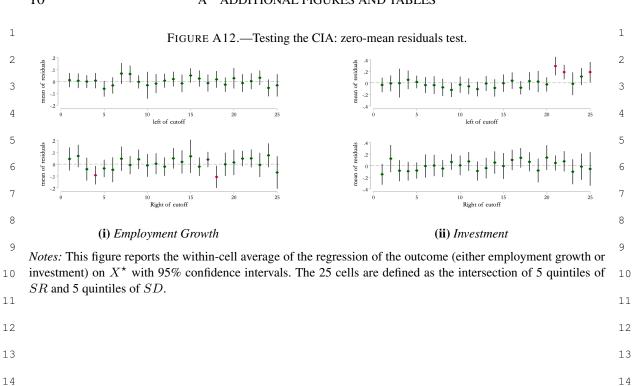
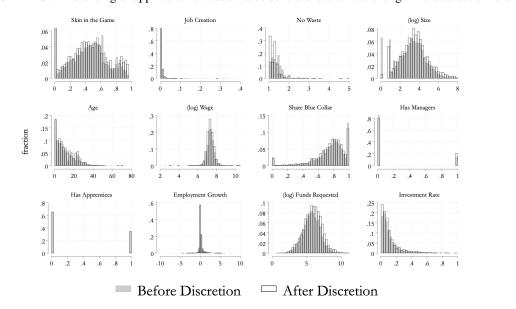
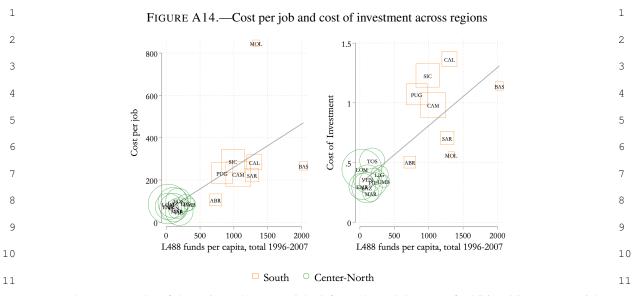


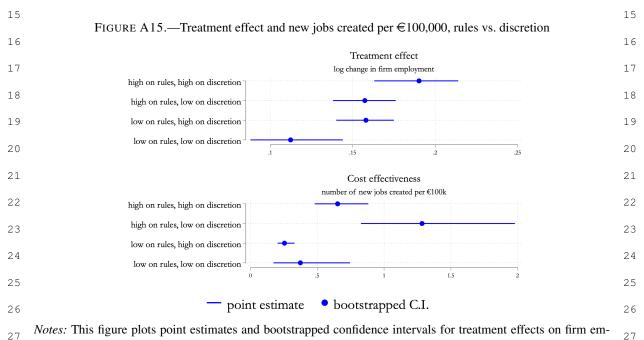
FIGURE A13.—Balancing of applicants' characteristics before and after the change in the selection rule



Notes: This figure reports the distribution of the applicants' characteristics in the auctions right before (gray bars, calls 1 and 2) and after (transparent bars, call 3) the introduction of political discretion. Only auctions concerning industry are included.

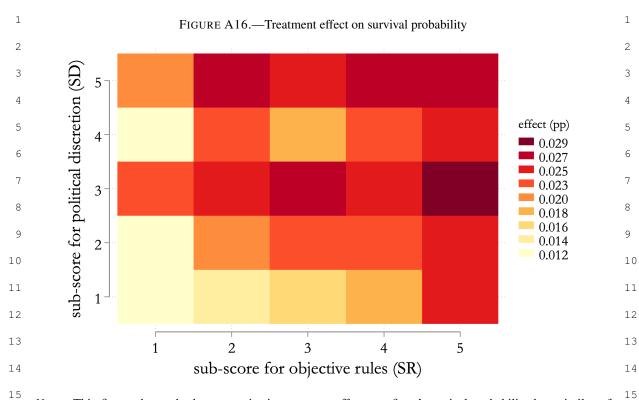


Notes: These are graphs of the estimated cost per job (left graph) and the cost of additional investment (right graph) against the total amount of L488/92 per capita across Italian regions. The size of markers is proportional to the regional population.

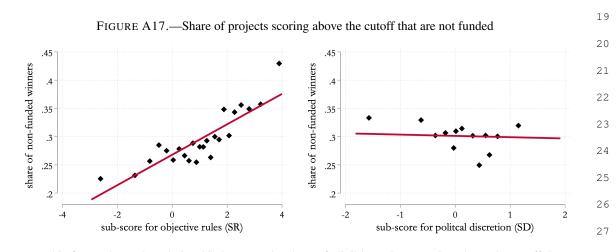


ployment growth (top graph) and the cost-effectiveness of subsidies (bottom graph), for four groups of applicants. Applicants "high on rules" ("low on rules") are those in the top (bottom) quintile of the objective sub-score SR; similarly, applicants "high on discretion" ("low on discretion") are those in the top (bottom) quintile of the discretionary sub-score SD. In practice, the four estimates in each graph refer to the four "corners" of the heatmaps in Figure 8, and 90% confidence intervals are bootstrapped as in Online Appendix Table A8.

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Notes: This figure shows the heterogeneity in treatment effects on firms' survival probability by quintiles of the sub-scores for objective rules (SR) and political discretion (SD). The treatment effect for each bin (SR = r, SD = d) is estimated as $\mathbb{E}[Y(1) - Y(0) \mid SR = r, SD = d] = (\beta_1 - \beta_0) \cdot \mathbb{E}[X^* \mid SR = r, SD = d]$. The covariates included in X^* are listed at the beginning of Section 6 of the main text.



Notes: This figure shows the relationship between the share of eligible projects scoring above the cutoff that are not funded eventually and, respectively, the sub-score for objective rules (left graph) and the sub-score for political discretion (right graph), controlling for cell fixed effects, across quantile-spaced bins. Covariate adjustment and the choice of the optimal number of bins are performed according to Cattaneo, Crump, Farrell & Feng (2024).

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TABLE A1 BALANCE OF FIRM CHARACTERISTICS ONE YEAR BEFORE THE CALL

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	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Specification:		lin	ear			quad	lratic	
Kernel:	unit	form	trian	gular	unif	orm	trian	gular
Group fixed effects	no	yes	no	yes	no	yes	no	yes
log-employment	0.044	0.002	0.027	0.006	0.000	0.017	0.020	0.026
	(0.043)	(0.034)	(0.04)	(0.034)	(0.048)	(0.04)	(0.048)	(0.04)
	[0.721]	[0.995]	[0.92]	[0.989]	[0.999]	[0.959]	[0.912]	[0.871]
log-change employment	0.016	0.013	0.009	0.011	-0.001	0.005	0.010	0.016
	(0.013)	(0.014)	(0.014)	(0.015)	(0.018)	(0.018)	(0.019)	(0.019)
	[0.721]	[0.887]	[0.92]	[0.906]	[0.998]	[0.959]	[0.912]	[0.871]
log-revenues	-0.071	-0.004	-0.102	-0.041	-0.151	-0.094	-0.120	-0.076
	(0.06)	(0.049)	(0.061)	(0.051)	(0.078)	(0.063)	(0.079)	(0.064)
	[0.721]	[0.995]	[0.486]	[0.906]	[0.342]	[0.642]	[0.63]	[0.806]
log-change revenues	-0.021	-0.030	-0.032	-0.038	-0.048	-0.051	-0.036	-0.037
0 0	(0.017)	(0.018)	(0.018)	(0.019)	(0.023)	(0.025)	(0.025)	(0.026)
	[0.721]	[0.557]	[0.457]	[0.33]	[0.282]	[0.276]	[0.642]	[0.687]
log-investment	0.022	0.049	0.001	0.022	-0.034	-0.009	-0.027	-0.009
	(0.079)	(0.071)	(0.083)	(0.077)	(0.107)	(0.098)	(0.108)	(0.098)
	[0.79]	[0.887]	[0.997]	[0.989]	[0.981]	[0.959]	[0.912]	[0.938]
log-change investment	0.124	0.088	0.102	0.065	0.066	0.045	0.109	0.088
0 0	(0.065)	(0.067)	(0.064)	(0.066)	(0.078)	(0.081)	(0.084)	(0.086)
	[0.409]	[0.733]	[0.486]	[0.906]	[0.849]	[0.959]	[0.712]	[0.871]
log-VA	-0.112	-0.088	-0.165	-0.133	-0.249	-0.208	-0.214	-0.188
U	(0.079)	(0.07)	(0.08)	(0.073)	(0.103)	(0.093)	(0.103)	(0.094)
	[0.614]	[0.733]	[0.293]	[0.41]	[0.141]	[0.215]	[0.296]	[0.342]
log-change VA	-0.065	-0.071	-0.073	-0.077	-0.084	-0.088	-0.084	-0.078
0 0	(0.05)	(0.052)	(0.055)	(0.057)	(0.073)	(0.076)	(0.075)	(0.077)
	[0.721]	[0.729]	[0.551]	[0.724]	[0.781]	[0.775]	[0.756]	[0.871]
log VA/worker	-0.081	-0.050	-0.109	-0.083	-0.153	-0.143	-0.150	-0.144
	(0.047)	(0.048)	(0.05)	(0.051)	(0.067)	(0.065)	(0.068)	(0.067)
	[0.462]	[0.844]	[0.253]	[0.542]	[0.192]	[0.239]	[0.233]	[0.249]
log-change VA/worker	-0.077	-0.089	-0.089	-0.099	-0.108	-0.116	-0.097	-0.099
	(0.05)	(0.051)	(0.057)	(0.058)	(0.077)	(0.079)	(0.078)	(0.08)
	[0.569]	[0.516]	[0.486]	[0.503]	[0.672]	[0.642]	[0.712]	[0.78]
firm age	0.261	0.177	0.029	0.029	-0.335	-0.224	-0.333	-0.249
C	(0.245)	(0.216)	(0.249)	(0.22)	(0.313)	(0.287)	(0.31)	(0.282)
	[0.721]	[0.887]	[0.991]	[0.989]	[0.786]	[0.919]	[0.756]	[0.871]
start-up	-0.006	-0.001	-0.006	-0.004	-0.007	-0.006	-0.010	-0.009
	(0.004)	(0.004)	(0.004)	(0.004)	(0.005)	(0.005)	(0.006)	(0.006)
	[0.569]	[0.977]	[0.486]	[0.906]	[0.75]	[0.775]	[0.578]	[0.591]

Notes: This table presents the results from a comparison of firm characteristics one year before the call between applicants scoring just above and just below the cutoff. All variables are described in Online Appendix Table B1. Start-up identifies firms in the age class (0-1). The numbers without brackets are the estimated coefficients from RD regressions analogous to Equation (3) in the main text in which the dependent variable is the firm characteristic indicated in each row and the main explanatory variable is a dummy equal to one for firms scoring just above the cutoff. The specification in columns (1)-(4) also includes the standardized application score and its interaction with the dummy for applicants above the cutoff, while columns (5)-(8) also include the squared application score and its interaction with the dummy; odd columns include group fixed effects for firms competing in the same ranking; and columns (3)-(4) and (7)-(8) weight observations by a triangular kernel in distance from the cutoff. Standard errors clustered by cell are reported in parenthesis. For each specification, p-values computed controlling the family-wise error rate when performing multiple hypothesis tests (Westfall & Young 1993) are reported in square brackets.

2.4

TABLE A2

BALANCE OF FIRM AND PROJECT CHARACTERISTICS IN INPS AND CERVED DATASETS.

Variable Name	INPS	INPS & CERVED	Normalized Difference	Westfall-Young p -value	
Skin in the Game	0.508	0.519	0.052	0.808	
Job Creation	0.009	0.009	-0.004	1.000	
No Waste	1.346	1.279	-0.099	0.982	
(log) Size	2.820	3.362	0.266	0.001	
Employment Growth	0.184	0.222	0.063	0.960	
(log) Funds Requested	5.659	5.846	0.150	0.960	
Age	11.222	10.705	-0.050	0.663	
South-based	0.774	0.714	-0.138	0.016	

1.0

Notes: The second and third columns of the table report averages of firms' and firms' projects' characteristics for applicants matched only in the INPS dataset and in the INPS and Cerved dataset, respectively. The fourth and fifth columns report the normalized difference (see Imbens & Rubin 2015, Section 14.2) of the second and third columns, and the Westfall & Young (1993) *p*-values computed controlling the family-wise error rate for multiple hypothesis tests. Only auctions concerning industry are included. The typical rule of thumb to detect imbalances for (the absolute value of) the normalized difference is 0.25 (see Imbens & Wooldridge 2009, p.24).

TABLE A3
CONDITIONAL INDEPENDENCE TESTS

	emplo	yment	inves	tment
Variable	left	right	left	right
Conditional on X^*				
	-			
score	0.0012	-0.0029	-0.004	-0.015
t-statistic	0.313	0.334	0.321	0.943
p-value	0.754	0.734	0.749	0.346
Unconditional				
score	0.0388	0.0145	0.047	-0.039
t-statistic	5.155	1.265	2.672	1.723
<i>p</i> -value	0.000	0.206	0.008	0.085
Obs	16,007	11,045	11,891	8,233

Notes: The table reports regression-based tests of the conditional independence assumption in Equation (4) in the main text. We regressed employment growth in the six years after the award of L488/92 subsidies on the running variable (i.e., the application score) separately for the sub-samples of applicants above and below the cutoff. The top panel shows the estimated coefficients when controlling for cell fixed effects and for the vector of covariates X^* , while the bottom panel reports the estimated coefficients when controlling only for cell fixed effects. Results are robust to including a quadratic polynomial in the running variable. The covariates included in X^* are listed at the beginning of Section 6 of the main text.

2.4

TABLE A4

APPLICANTS' CHARACTERISTICS BEFORE AND AFTER THE CHANGE IN THE SELECTION RULE

2.4

Variable Name	Before Discretion	After Discretion	Normalized Difference	Westfall-Young p-value
Skin in the Game	0.458	0.453	-0.019	0.915
Job Creation	0.006	0.007	0.098	0.915
No Waste	1.288	1.359	0.233	0.459
(log) Size	3.463	3.095	-0.247	0.414
Age	11.297	10.840	-0.048	0.999
(log) Wage	7.765	7.803	0.094	0.972
Share Blue Collar	0.749	0.751	0.005	0.993
Has Managers	0.207	0.153	-0.143	0.934
Has Apprentices	0.345	0.344	-0.004	0.873
Employment Growth	0.136	0.245	0.183	0.998
(log) Funds Requested	5.903	5.700	-0.158	0.998
Investment Rate	0.113	0.120	0.059	0.365

Notes: This table reports averages of applicants' characteristics in the auctions before (calls 1 and 2) and right after (call 3) the introduction of political discretion, their normalized difference (see Imbens & Rubin 2015, Section 14.2), and the Westfall & Young (1993) *p*-values computed controlling the family-wise error rate for multiple hypothesis tests. Only auctions concerning industry are included. The typical rule of thumb to detect imbalances for (the absolute value of) the normalized difference is 0.25 (see Imbens & Wooldridge 2009, p.24).

 $\label{thm:thm:thm:cost} TABLE\ A5$ Cost of New jobs and investment matching firms in the South

	(1)	(2)	(3)	(4)	(5)	(6)
Cost measure:	1	new job ds of €'s)		worker-year nds of €'s)		v investment of investment)
X^{\star} :	manual	data-driven	manual	data-driven	manual	data-driven
all regions	115	106	30	30	0.63	0.49
	[87; 220]	[80; 206]	[27; 38]	[27; 36]	[0.47; 1.10]	[0.36; 0.69]
south	140	122	39	35	0.78	0.63
	[107; 238]	[95; 205]	[33; 51]	[31; 44]	[0.59; 1.44]	[0.47; 0.93]
north-center	68	70	16	19	0.35	0.25
	[42; 211]	[44; 215]	[13; 20]	[16; 22]	[0.24; 0.60]	[0.18; 0.34]

Notes: This table shows the cost of new jobs and of new investment generated by the L488/92 subsidies over a six-year period. All amounts are expressed in thousand € at constant 2010 prices. Differently from Table 4 in the main text, costs are calculated on a subsample of Southern firms matched (1-to-1) to Northern firms based on a set of observables (age and industry, size and employment composition, average wage, and past employment growth). The estimates in columns labeled as "manual" employ the set of covariates listed at the beginning of Section 6 of the main text, while the estimates in columns labeled as "data-driven" employ the set of covariates selected by the algorithm described in detail in Section S3 of the Supplementary Materials. 90% confidence intervals are reported in brackets and are computed using 1,000 draws of a non-parametric cluster Efron bootstrap, where clusters are defined at the cell-level.

2.4

WY p-value

TAE	BLE A6
POLICY INVARIANCE TEST	DIFFERENCE-IN-DIFFERENCES

2.4

Variable Name	Skin in the Game	Job Creation	No Waste	$ \begin{array}{c} \textbf{(log)} \\ \textbf{Size} \ [t-1] \end{array} $	Age	(log) Wage
$POST1998 \times DISCRETION$	-0.032	-0.000	0.082	-0.108	0.138	0.052
FO31 1996 X DI3CRETION	(0.027)	(0.001)	(0.082)	(0.073)	(0.280)	(0.014)
Obs	38,367	38,367	38,367	38,367	38,367	34,747
Adj. R^2	0.109	0.100	0.685	0.121	0.045	0.092
WY p-value	0.826	0.932	0.870	0.746	0.932	0.142
	Share of	Has	Has	Employment	(log) Funds	Investment
	Blue Collar	Managers	Apprentices	Growth $[t-1]$	Requested	Rate
$POST1998 \times DISCRETION$	-0.004	-0.005	-0.065	-0.018	-0.030	0.008
	(0.011)	(0.016)	(0.012)	(0.012)	(0.062)	(0.007)
		20.265	20.265	24.010	20.267	15 104
21						
Obs	34,747	38,367	38,367	34,819	38,367	15,104
Obs Adj. R^2	0.020	0.070	0.053	0.006	0.230	0.009

0.104

0.736

0.932

0.826

Notes: This table shows the results of difference-in-differences regressions comparing project and applicant characteristics between regions attributing and not attributing discretionary points, before and after the introduction of discretion. In particular, we estimate the specification $Y_{irt} = \phi(POST1998_t \times DISCRETION_r) + FE_r + FE_t + \nu_{irt}$, where $POST1998_t = 1$ for the period after 1998 and $POST1998_t = 0$ otherwise, $DISCRETION_r = 1$ in regions attributing discretionary points and $DISCRETION_r = 0$ otherwise, and FE_r and FE_t are region and year fixed effects, respectively. Robust standard errors clustered at the region-year level and reported in brackets. The last row reports Westfall & Young (1993) p-values corrected for multiple-hypothesis tests.

0.932

0.932

TABLE A7

COST OF NEW JOBS AND INVESTMENT GENERATED BY L488 SUBSIDIES

	(1)	(2)	(3)	(4)	(5)	(6)
Cost Measure		new job ad of €'s)		worker-year and of €'s)		investment d of €'s)
X^{\star}	manual	data-driven	manual	data-driven	manual	data-driven
all firms	178	159	54	56	0.81	0.63
	[133; 299]	[118; 260]	[47; 62]	[51; 62]	[0.59; 1.25]	[0.48; 0.87]
large	78	78	24	29	0.4	0.29
	[47; 222]	[52; 174]	[19; 30]	[25; 35]	[0.27; 0.74]	[0.21; 0.41]
small	253	209	80	74	1.08	0.87
	[203; 349]	[162; 296]	[73; 90]	[68; 81]	[0.83; 1.49]	[0.68; 1.15]

Notes: This table shows the cost of new jobs and investment generated by the L488 subsidies over a six-year period. All amounts are expressed in thousand € at constant 2010 prices. The estimates in columns labeled as "manual" employ the set of covariates listed at the beginning of Section 6 of the main text, while the estimates in columns labeled as "data-driven" employ the set of covariates selected by the algorithm described in detail in Section S3 of the Supplementary Materials. 90% confidence intervals are reported in brackets and are computed using 1,000 draws of a non-parametric cluster Efron bootstrap, where clusters are defined at the cell-level.

1		1
2		2
3		3
4	TABLE A8	4
	POINT ESTIMATES AND CONFIDENCE INTERVALS OF TREATMENT EFFECT AND AVERAGE COST PER NEW	
5	JOB, RULES VS. DISCRETION	5
6	Panel A: Treatment Effect	6
7	quintiles of sub-score SR	7
8	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	8
9	る。[.14; .175] [.163; .217] [.151; .195] [.159; .201] [.163; .214] という。 0.126	9
1.0	4 [.099; .156] [.131; .179] [.141; .191] [.137; .199] [.129; .172]	1.0
10	9 0.164 0.163 0.171 0.181 0.176	10
11	5	11
12	2 [.09; .154] [.127; .171] [.122; .168] [.133; .194] [.142; .192]	12
13	1 0.112 0.107 0.132 0.135 0.157 [.088; .144] [.084; .136] [.107; .156] [.114; .155] [.138; .176]	13
14	Panel B: Cost Effectiveness	14
15	quintiles of sub-score SR 1 2 3 4 5	15
16	0.254	16
17	5	17
17	6 4 0.359 0.491 0.505 0.682 1.157 9 4 [.261; .502] [.306; .852] [.348; .754] [.466; .984] [.751; 1.739]	17
18	(1.501, 1.502) [1.500, 1.652] [1.540, 1.754] [1.400, 1.764] [1.751, 1.757] (1.751, 1.757) (1.751, 1.757)	18
19	5 3 [.388; .677] [.337; .577] [.39; .772] [.493; .991] [.611; 1.648]	19
1.7	§ 0.466 0.434 0.444 0.610 1.149	10
20	[324; .741] [.343; .589] [.32; .754] [.472; .837] [.588; 1.935] 0.373	20
21	[.172; .745] [.505; .987] [.595; .999] [.621; 1.124] [.827; 1.979]	21
22	Notes: This table reports the heterogeneity in treatment effects on firm employment growth (Panel A) and	22
23	the cost-effectiveness of subsidies (Panel B), by quintiles of the sub-scores for objective rules (SR) and political discretion (SD) . In Panel A, the treatment effect for each bin $(SR = r, SD = d)$ is estimated as	23
24	$\mathbb{E}[Y^1 - Y^0 \mid SR = r, SD = d] = (\beta_1 - \beta_0) \cdot \mathbb{E}[X^* \mid SR = r, SD = d].$ The covariates included in X^* are	24
25	listed at the beginning of Section 6 of the main text. In Panel B, cost-effectiveness is measured by the number of newly created per €100,000 of subsidies received by the firm. The number of newly created jobs in each	25
26	bin is computed by multiplying the size of each firm by the treatment effect for its respective bin, as reported in	26
	Panel A, and aggregating across all firms in that bin. 90% confidence intervals are reported in brackets and computed using 1,000 draws of a non-parametric cluster Efron bootstrap, where clusters are defined at the cell-level.	
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1	B. DATA DESCRIPTION	1
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3	The analysis leverages three main sources of microdata:	3
4	1. The administrative data on all applications for L488/92 subsidies (1996-2007),	4
5	sourced from the Italian Ministry of Economic Development, DG Firm subsidies	5
6	(Ministero dello Sviluppo Economico 2009);	6
7	2. The National Social Security Institute (INPS) firm archive (called DM10M) covering	7
8	the universe of Italian firms with at least one dependent worker, available starting from	8
9	1986 (INPS n.d.);	9
10	3. The Cerved database containing balance sheet information on Italian limited liability	10
11	companies, available starting from 1993 (Cerved Group n.d.).	11
12	The L488/92 archive contains administrative data on 74,584 applications for L488/92	12
13	subsidies, submitted by 49,082 firms. It covers nearly the universe of rankings (only some	13
14	smaller auctions are excluded) for which it contains all submitted applications.	14
15	The data contain:	15
16	(i) information on the project: a unique project identifier; the three (five) sub-indexes	16
17	measuring project quality; the score (the forcing variable) obtained aggregating the	17
18	sub-indexes standardized at the auction and region level; the position in the ranking;	18
19	an indicator for winning projects; the amount of funds requested in the application	19
20	and that of funds actually transferred - separate by each of three installments; whether	20
21	financed on EU or Italian sources.	21
22	(ii) information on the auction (number, region, and sector of destination of funds, date of	22
23	issuance, date of closure, dates of each of the three installments).	23
24	(iii) information on the firm (fiscal identifier or individual fiscal code (in case of sole pro-	24
25	prietorships), legal form, address, municipality.	25
26	Additional information on the auction, recovered from the Official Journal of the Ital-	26
27	ian Republic (Gazzetta Ufficiale della Repubblica Italiana n.d.), associates each project to	27
28	"cells" identified by the following dimensions: firm size (Large/Medium/mall), sector (In-	28
29	dustry, energy, Tourism, Trade, Services), eligibility for co-financing (Yes/No), and geo-	29
30	graphical area (Region). This additional information allowed us to allocate projects exactly	30

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to the several sub-rankings within the same call, region, and (possibly) special category of 1 applicants (see Section 1 of the Supplementary Materials). 2

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The firm archive is assembled by INPS sourcing on a master dataset collecting all social 3 security payments made every month by legal entities for any employee with open-ended, fixed-term, and apprenticeship contracts. The archive covers therefore the universe of firms with at least one employee at some point during a given calendar year. The data is available between 1986 and 2015. For each firm, it reports the fiscal code; monthly information on the number of employees; and yearly information on the number of employees and their total wage bill by qualification (manager, blue-collar; white collar; apprentices; others); date of birth, and cessation of activity; detailed geographical (municipality) and industry (3-digit) data; and an identified for firms belonging to groups.

Information on firms' balance sheets and income statements comes from a proprietary database assembled by the Cerved Group S.p.a. The Cerved Firm Registry, which is the Italian source of the Orbis database, covers the universe of limited liability firms in the private non-financial sector and is available since 1993.¹

Further data used in the paper include (i) the administrative registries of local politicians (Ministero dell'Interno n.d.a) and local elections (Ministero dell'Interno n.d.b), available from the Italian Ministry of Interior, respectively at https://elezionistorico.interno.gov.it/ eligendo/opendata.php and https://dati.interno.gov.it/; (ii) a classification of local governments' ideologies, sourced from the Local Opportunities Lab (Local Opportunities Lab n.d.) at https://www.localopportunitieslab.it/; (iii) data on various economic variables related to labor market participation, unemployment, employment rates, education, and other demographic and economic indicators at the municipality level, obtained from the Census (ISTAT (n.d.a), available at https://ottomilacensus.istat.it/); and (iv) regional data on per-capita GDP and population in 1995 (CRENOS n.d.).

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¹Nominal values were deflated sourcing on (ISTAT n.d.b), (ISTAT n.d.c), and (ISTAT n.d.d). Supplementary information on firms' industry codes was sourced from (Infocamere n.d.) and (Ministero dell'Economia e delle Finanze – Agenzia delle Entrate n.d.).

TABLE B1 1 1 DESCRIPTION AND SOURCE OF ALL THE VARIABLES USED IN THE ANALYSIS. 2 Variable **Description** Source 3 Main variables from L488/92 data 4 Info on Auctions Date, region, and result of the auction. Complementary infor-MinEcDev mation from the Official Journal includes, for each project, all 5 5 the details required to recover the rankings within each auction-6 region cell, as explained in Section S1 of the Supplementary Materials 7 7 Score Project quality obtained combining the 3(5) indicators below, MinEcDev 8 once standardized within each call-region Ratio of the applicant's own investment in the project relative to Skin in the Game MinEcDev 9 9 the amount requested 10 Job Creation Number of jobs created by the project MinEcDev 10 Proportion of funds requested in relation to an ad-hoc bench-No Waste MinEcDev 11 11 mark set by the EU Commission Score attributed on the basis of priorities indicated by the re-Political Discre-MinEcDev 12 tion gional government 13 13 Compliance with the requirements of an environmental man-Environmental MinEcDev agement system, e.g. ISO 14001 or EMAS Responsibility 14 14 Funds Requested Amount of subsidies requested in application MinEcDev 15 15 Amount of subsidies disbursed to winners, in three instalments Funds Paid MinEcDev 16 16 Main variables from INPS 17 17 Number of employees, available at monthly frequency **INPS** Size 18 18 Employment growth rate between two dates. Computed over Growth **INPS** different horizons starting and ending in the month of the auc-19 19 tion 20 20 Firm age at any given year **INPS** Age Average wage of employees. Obtained aggregating yearly data Wage **INPS** 21 21 on wage bill and employees by qualification (managers, blue collar; white collar; apprentices; others) 2.2 2.2 Share of blue col-Ratio between blue collar employees and total employees, com-**INPS** 23 2.3 puted from the same data lars Dummy for presence of (middle) managers in workforce Manager **INPS** 24 24 Apprentices Dummy for presence of apprentices in workforce **INPS** 25 25 Survival Dummy for whether firm is alive at any given future horizon **INPS** Headquarter municipality Area **INPS** 26 26 3-digits NACE Rev. 2 industry codes Industry **INPS** 27 27 28 Main variables from CERVED 28 Firm revenues (sales) (thousand €) Revenues CERVED 29 29 VA Firm value added(thousand €) CERVED 30 **Total Assets** Total assets (thousands of \in) **CERVED** Investment in tangible and intangible fixed assets (thousand €) Investment **CERVED**

Variable	Description	Source
Political proxim	nity and other predictors of the discretionary sco	ore (SD)
Political alignment	Dummy for the same party (right, centre, left, civic) ruling both the Region and the municipality the firm is located	Ministry of Interior and Local Opportu- nities Lab
Margin of victory	Dummy for the margin of victory in the last elections municipality the firm is located	Ministry of Interior
Birthplace of Regional president	Dummy for the president of the Regional government being born in the municipality the firm is located	Ministry of Interior
Birthplace of Regional counsellor	Dummy for at least one counsellor in the Regional government being born in the municipality the firm is located	Ministry of Interior
Birthplace of Regional alderman	Dummy for at least one alderman in the Regional government being born in the municipality the firm is located	Ministry of Interior
Human capital of Regional president / municipality mayor	Dummy for level of schooling of Regional president / municipality mayor (primary, lower sec., high school, university degree)	Ministry of Interior
Local unemployment	Unemployment rate at province level (ISTAT)	ISTAT
Credit constraints	Spread between loan and deposit rates in provinces	Guiso, Pistaferri & Schivardi (2013)
	Main variables from Census Data	
Participation Rate	Labor market participation (males, females), ratio of active and inactive young people	Census data
NEET	Incidence of young people aged 15-29 not studying, not working, and outside the labor market and education	Census data
Unemployment	Male, female, and youth unemployment rate	Census data
Employment	Male, female, and youth employment rate, employment turnover index, incidence of em- ployment in the agricultural, industrial, ter-	Census data
Education	tiary (excluded trade), and trade sector Early exit from the education and training system, incidence of adults with a diploma	Census data
	or higher, incidence of adults with a middle school diploma	
Socio-economic	Population density, housing usage potential in urban centers, incidence of families at risk	Census data

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C. RD ESTIMATES AT THE CUTOFF: ADDITIONAL RESULTS

Two important issues could affect the interpretation of our RD estimates in Section 5 of the main text. First, applicants in a given call may re-apply (and obtain funds) in subsequent calls. We deal with this issue in Supplementary Appendix, Section S2. Second, the effects on funded firms may spill over to other, non-funded firms.

Spillovers Employment increases by subsidized firms may affect other, non-subsidized firms. The sign of these effects is also unclear a priori. The growth of subsidized firms may benefit upstream and downstream producers in the same market, or it may erode the market share of competitors – possibly including firms in the control group. In the latter case, estimates in Section 5 would overstate the effects of the policy.

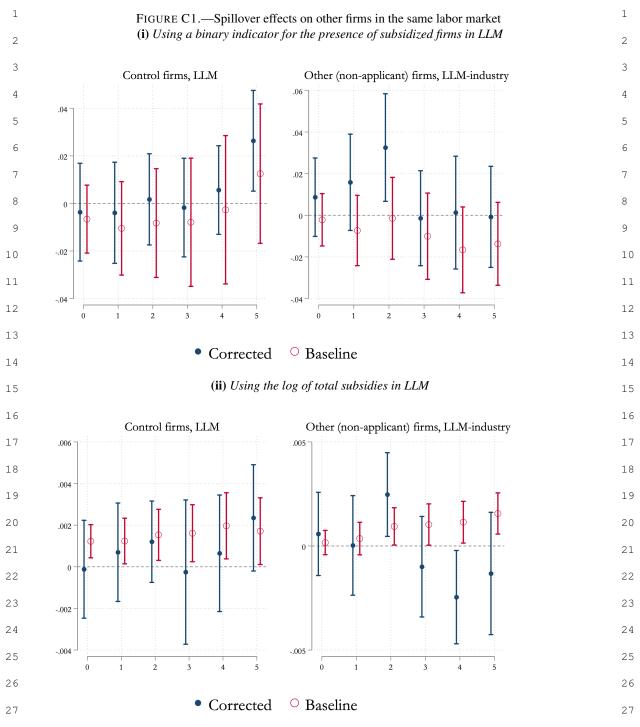
To address this possibility, we look across Italian Local Labor Markets (LLM) comparing the employment dynamics of non-subsidized firms in subsidized LLMs to those of firms in non-subsidized LLMs; spillover effects should affect more (or exclusively) employment in the former group. We focus on the following specification:

$$\ln L_{m,t+k} - \ln L_{m,t} = \theta_k D_{m,t} + \alpha \ln L_{m,t} + F E_m + F E_t + \varepsilon_{m,t}$$
(1)

where $L_{m,t+k}$ and $L_{m,t}$ are the total employment of non-subsidized firms in the m-th LLM in year t+k and t, taken from the INPS administrative data on the universe of workers in (non-agricultural) firms; $D_{m,t}$ is a dummy equal to 1 when at least one firm in LLM m received funds in year t; FE_m and FE_t are LLM- and year-specific fixed effects; and $\varepsilon_{m,t}$ is a residual summarizing the effect of other factors. The coefficient of main interest, θ_k , captures the differential employment response, after k years, of non-subsidized firms within the same LLM as a subsidized firm relative to non-subsidized firms in other LLMs. Figure C1i plots the estimated coefficients θ_k 's for two different subsets of non-subsidized firms – respectively, applicant firms not obtaining the subsidy (left graph) and non-applicant firms in the same LLM-industry cell as subsidized firms. Pother LLMs.

²Industry is defined at the 3-digit level.

present baseline difference-in-differences estimates as well as "corrected" estimates accounting for the staggered research design, using the approach suggested by de Chaisemartin & D'Haultfœuille (2020). Overall, there is no evidence of significant spillover effects; the same is true when re-placing the binary indicator $D_{m,t}$ in Equation 1 with the (log of) funds actually paid to 5 subsidized firms in each LLM or LLM-industry, see Figure C1ii. These results imply that higher employment among subsidized firms reflects a net 7 increase in aggregate employment, rather than a mere reallocation of jobs from nonsubsidized to subsidized firms. Cerqua & Pellegrini (2022) reach the same conclusion by decomposing worker flows towards subsidized firms. Using worker-level data, they show that the majority of recruits come from new entrants in the labor market, and conclude that 11 L488/92 subsidies generate few displacement effects across firms, if at all. 2.4



Notes: The graphs show the estimated spillover effects of the subsidy on local employment at different time horizons, indicated on the horizontal axis, and associated confidence intervals (at the 90% significance level). The left panel plots the aggregate employment response of control firms located in the same LLM as treated firms. The right panel focuses on non-participating firms in the LLM and (3-digit) industry as treated firms. The treatment variable is the log of funds received by treated firm in an LLM (or LLM-industry cell). "Baseline point" estimates and confidence intervals are obtained from specification 1 in the main text, clustering heteroskedasticity-robust standard errors by LMM. "Corrected" coefficients are obtained using the estimator proposed by de Chaisemartin & D'Haultfœuille (2020) to account for biases arising if group-time treatment effects are averaged with negative weights.

D. POLITICIANS' RESPONSE TO (EXPECTED) OBJECTIVE SCORES

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As explained in Section 2.2 of the main text, local politicians attribute a discretionary score SD to projects depending the municipality within the region and industrial sectors

in which they are realized, and the type of investment to be implemented. In addition, SD must be set ex-ante and communicated to the Ministry of Economic Development by

October 30th of the year before each call was issued, and it is not circulated publicly.

When allocating the discretionary points by the municipality-industry-type of each project (SD), politicians may in principle take into account the *expected* score received by projects on objective criteria, call it SR^e – the actual score will only be revealed a few months later. In particular, they should attribute more points to projects that they favor and, at the same time, they expect to score lower on objective criteria.

Letting Z denote the triple of projects' operating municipality, industry, and type of investment, and Z the set of possible values for such triple, politicians allocate SD across Z's taking into account the scores received on the objective criteria:

 $SD(Z, SR^e(Z)).$

Therefore, SD depends on Z both directly and indirectly through SR^e , and we expect that SR^e enters negatively SD. In order to estimate such effect, we need to impose assumptions on how politicians form expectations SR^e , since our data allow us to observe ex-post realizations of SR in each call t but not the ex-ante expectations. In particular, we consider two alternative hypotheses.

- (i) "Adaptive expectations": politicians form such expectations based on the average realizations of SR within each group $z \in \mathcal{Z}$ in the previous call t-1, $SR_{z,t}^e = \sum_{g(j)=z,t-1} SR_{j,t-1}/N_{t-1}^z$, where N_{t-1}^z denotes the number of projects in each group $z \in \mathcal{Z}$ in call t-1 and $g: \{1,2,\ldots,N\} \to \mathcal{Z}$ is a function mapping rankings into groups.
- (ii) "Perfect foresight": politicians are able to correctly predict the average realization of SR in group z in the call at time t ($SR_{z,t}^e = \sum_{g(j)=z,t} SR_{j,t}/N_t^z$).

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In addition, we must impose that politicians' utility is stable over time. Under these assumptions, we can identify the effect of SR^e on SD across groups defined by Z leveraging longitudinal variation over subsequent calls and controlling non-parametrically for differences in projects' characteristics across groups through a full set of fixed effects:

$$SD_{z,t} = \beta SR_{z,t}^e + FE_z + FE_t + \varepsilon_{z,t}, \tag{2}$$

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where FE_z and FE_t are fixed effects by group z and call, respectively. We weight the regression across Z's by the number of projects in each Z triple.

Estimates of Equation (2) are reported in Table D1. When assuming adaptive expectations (columns 1-4), the estimated coefficient β is essentially zero; in particular, we can reject (with 95% confidence) effects as small as a -0.03 standard deviation changes in the discretional score for a one standard deviation increase in the objective score. When assuming perfect foresight (columns 5-8), the coefficient is more precisely estimated but remains extremely small in magnitude – we can reject effects larger than a -0.04 standard deviation changes in the discretional score for a one standard deviation increase in the objective score. The results are very similar when assuming that politicians predict the median – rather than the mean – of SR (columns 3 and 7); when running on the unweighted regression across Z's (columns 2 and 6); and when considering a quadratic specification (columns 4 and 8). In the Supplementary Materials, we presents additional analysis, which allows for a flexible relationship between SD and SR^e . These additional results corroborate the evidence

that SR^e is independent of SD.

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7											7
8						TABLE	D1				8
9	SD response to SR								9		
10			(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	10

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
	expectations based on avg. SR in $t-1$				expectations based on avg. SR in t				
SR^e ,	-0.004 (0.012)	0.001 (0.013)	-0.005 (0.012)	-0.002 (0.012)	-0.024*** (0.008)	-0.023*** (0.009)	-0.025*** (0.008)	-0.026*** (0.008)	
SR^e , squared	(****)	()	-0.001 (0.002)	(****)	(******)	(3,2,2,2,7)	0.002 (0.002)	(1111)	
Obs	3,083	3,083	3,083	3,083	8,471	8,471	8,471	8,471	
N. Z-triples Weighted	1,284 •	1,284 x	1,284	1,284	3,518	3,518 x	3,518	3,518	
Statistic Adj. R^2	average 0.417	average 0.395	average 0.417	median 0.417	average 0.385	average 0.374	average 0.386	median 0.386	

Notes: This table reports the results of regressions on equation (2) across groups of projects defined by triples of municipality-industry-type of project. The dependent variable is the sub-score SD attributed by politicians to each group. The main explanatory variable are politicians' expectations on the average score SR in each group $z \in \mathcal{Z}$. In columns (1)-(4) we assume that such expectations are based on the average realizations of SR within each group in the previous call t-1 ($SR_{z,t}^e = \sum_{g(j)=z,t-1} SR_{j,t-1}/N_{t-1}^z$, where N_{t-1}^z denotes the number of projects in group $z \in \mathcal{Z}$ during the call t-1 and $g: \{1,2,\ldots,N\} \to \mathcal{Z}$ is a function mapping rankings into groups), while in columns (5)-(8) we assume that they are based on the average realizations of SR within each group in the call at time t ($SR_{z,t}^e = \sum_{g(j)=z,t} SR_{j,t}/N_t^z$). All regressions include fixed effects by group and call. Standard errors clustered by group are reported in parenthesis.

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