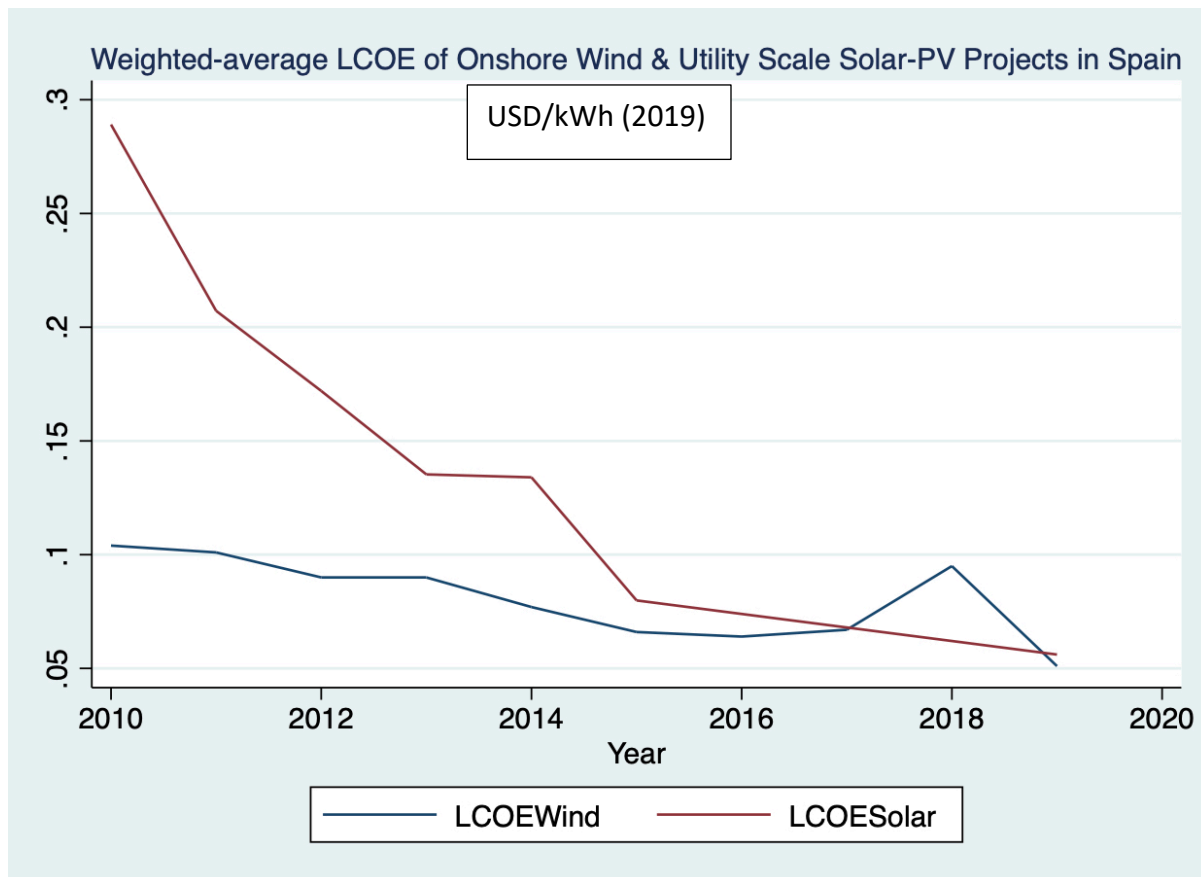


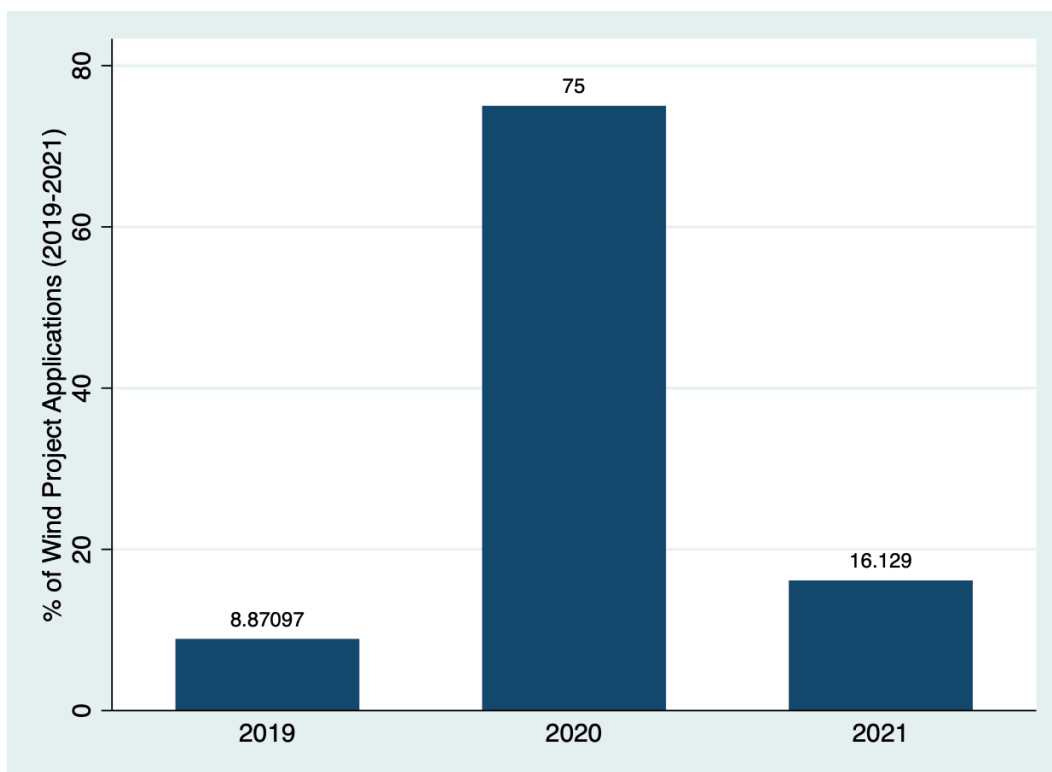
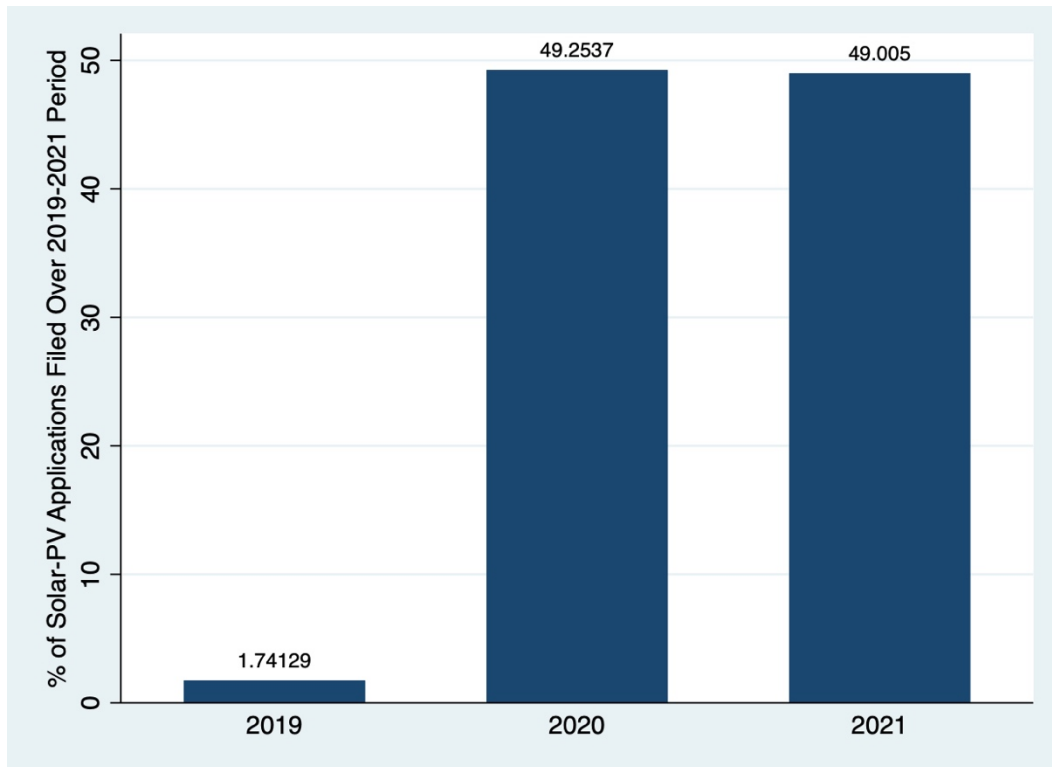
Annexes: “Renovables sí, però així no?”

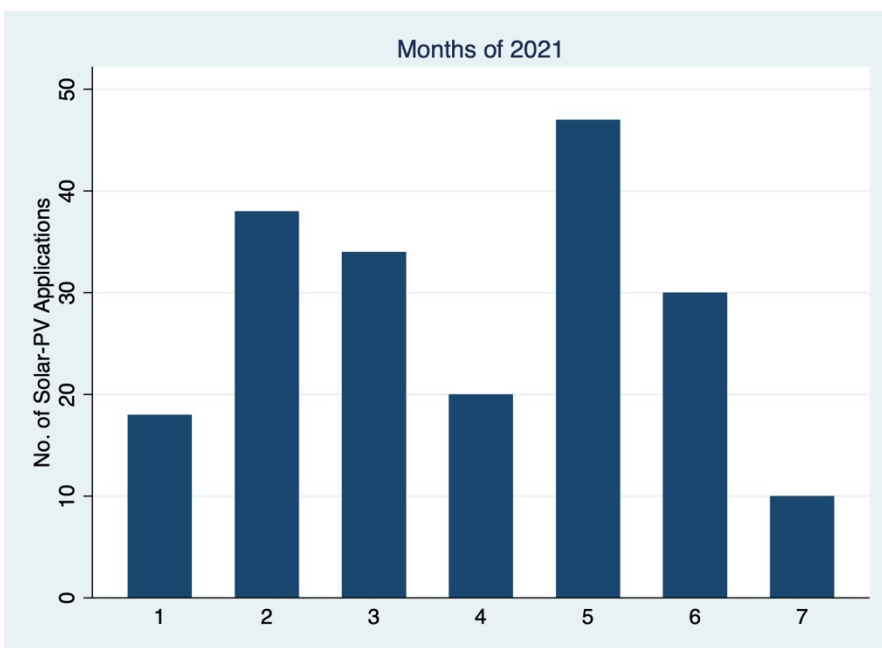
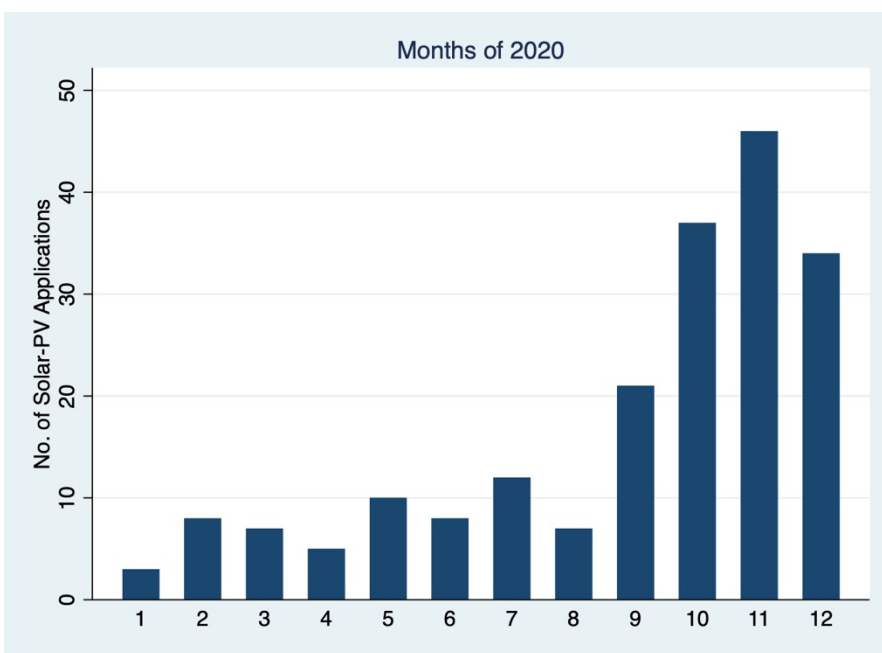
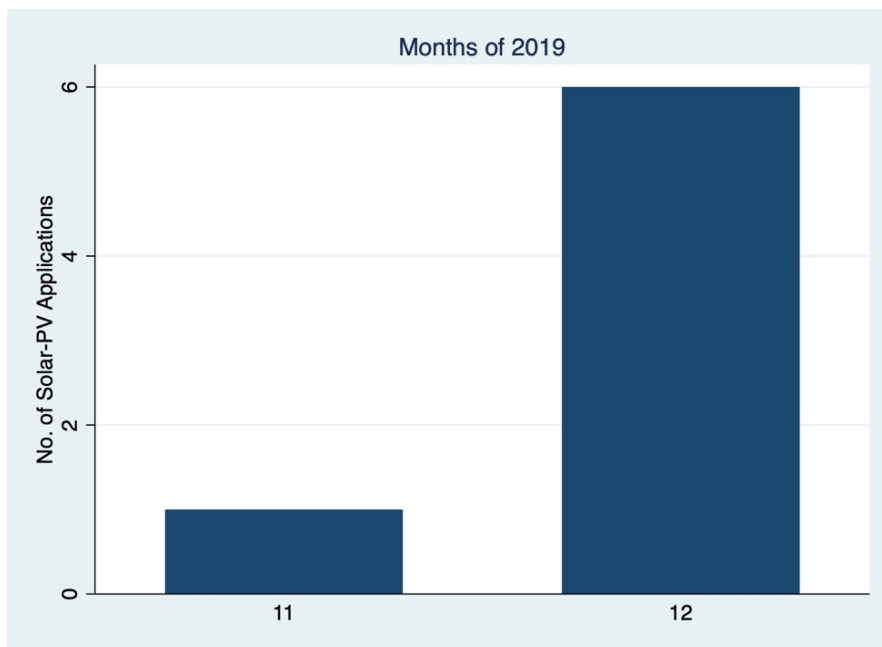
Annex 0

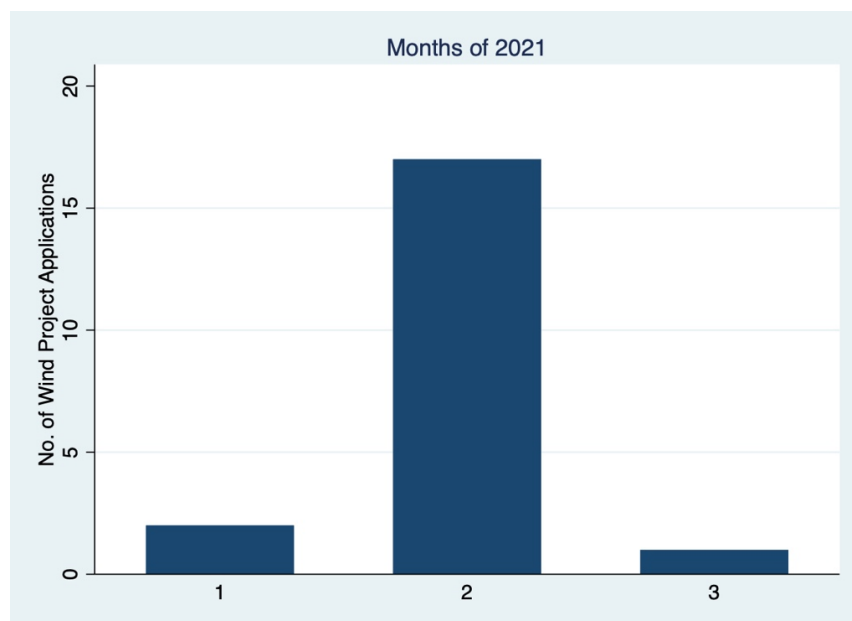
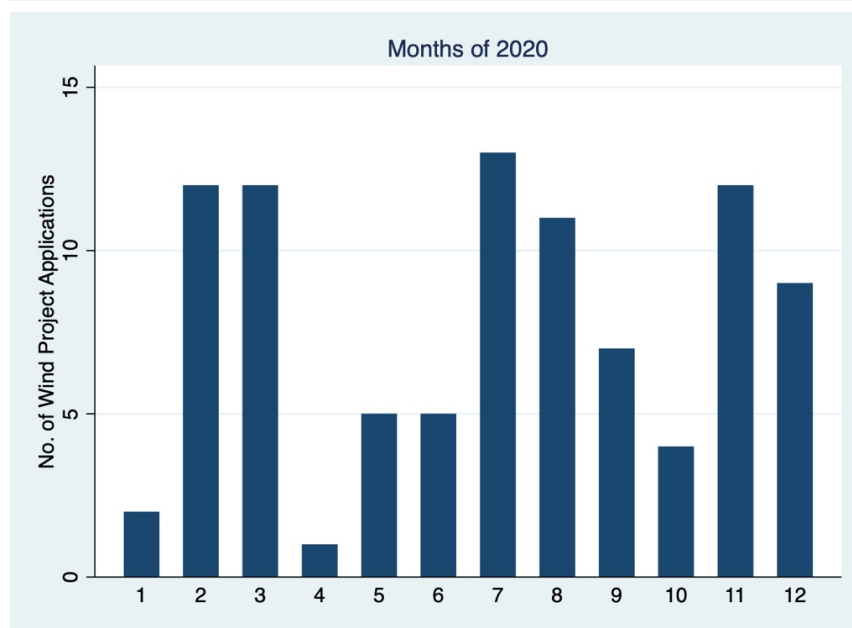
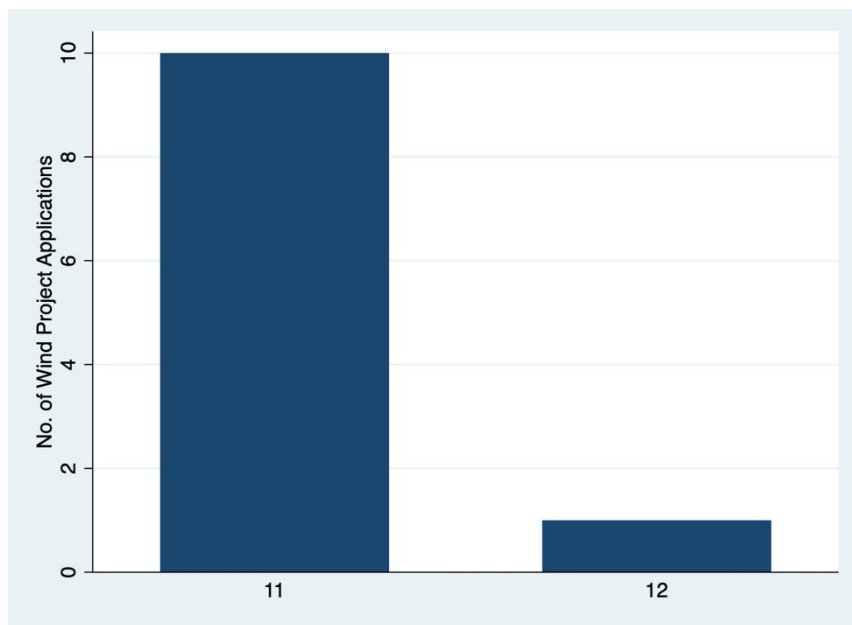


Source: IRENA., 2020. Renewable Power Generation Costs in 2019, International Renewable Energy Agency, Abu Dhabi.

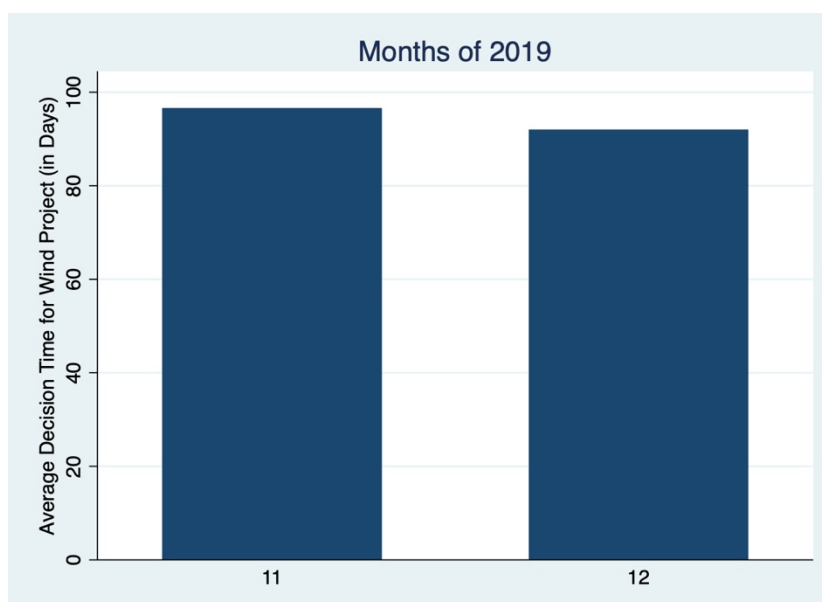
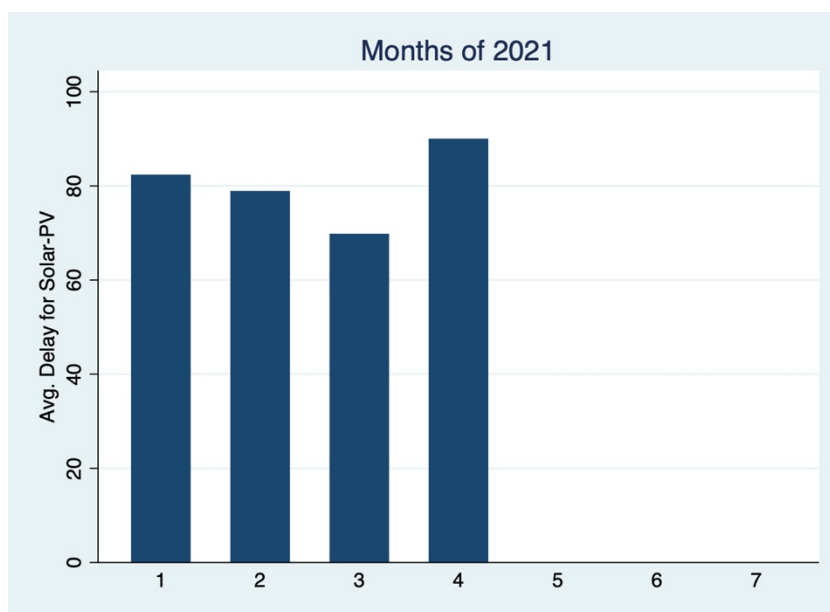
Annex 1A

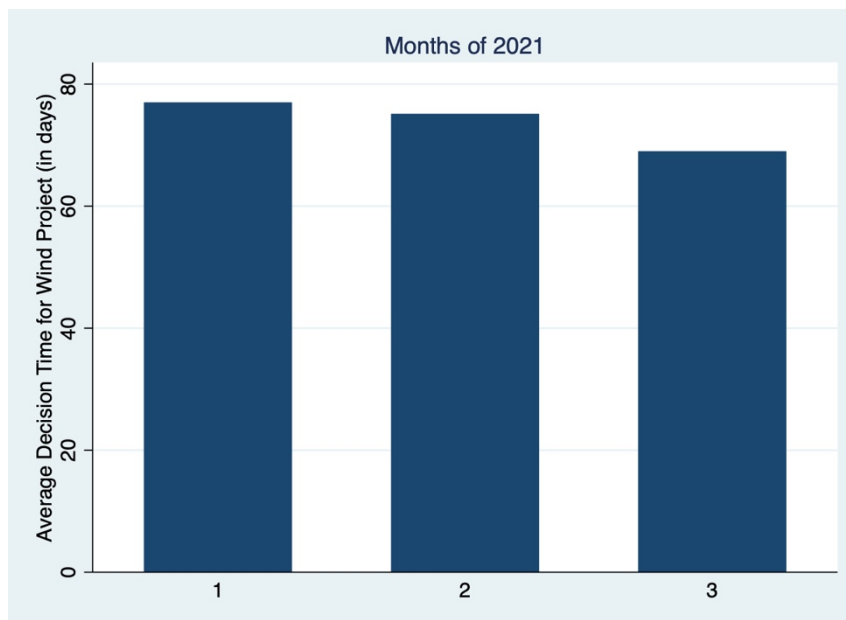
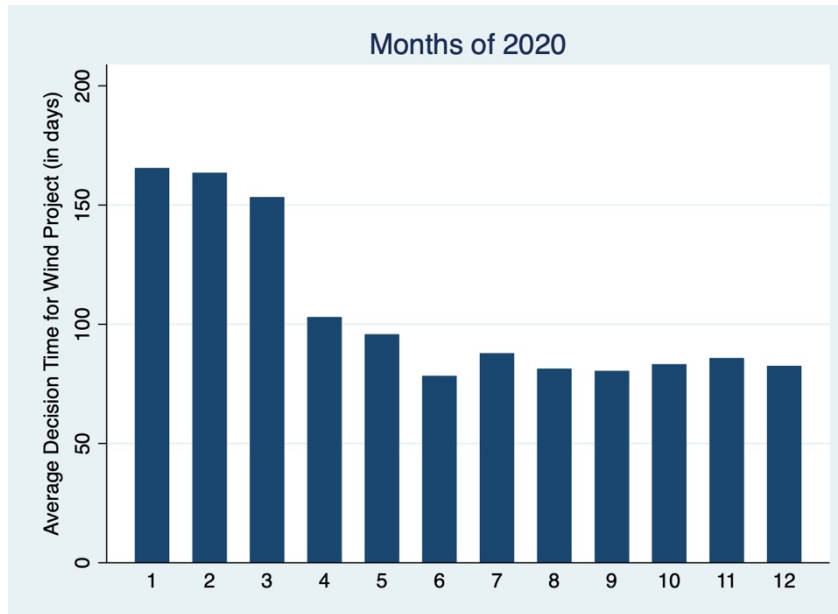






Annex 1B





Annex 2A

Our samples of solar-PV and wind project applications from DATASET C are compatible with the following assumptions.

1. Our samples are representative because it contains observations on solar-PV and wind projects applications from all regions of Catalonia. To the best of our knowledge, no. criteria or bias affecting their inclusion or exclusion in DATASET C. Our analysis also avoids any bias. Our analysis naturally filters out observations with missing values on the decision time variable(*deldec*) but there was no systemic selection bias weakening the randomness and representativeness of the sample.
2. Our samples are also sufficiently large ($N > 50$) with 266 and 124 non-missing observations on *lag* for solar-PV and wind project applications respectively. “Because our sample size is sufficiently large ($N > 50$), we know, based on the central limit theorem, that the sampling distribution of the mean will be approximately normal” **Leon-Guerrero and Frankfort-Nachmias, 2017, p.159**). Therefore, the normality assumption is also met.
3. Bartlett’s test shows that our samples on wind and solar-PV project applications are very likely to have unequal variances on the *deldec* variable when grouped and compared based on the year of filing (variable *datenew*). Therefore, to adjust for these unequal variances in calculating confidence intervals and conducting pairwise comparison of difference of means, we use methods developed by Games and Howell (cited in Tamhane, 1979) and other authors (see next page).

Bartlett’s equal-variances test for the sample of wind project applications. The P value ($\text{prob} > \chi^2$) indicates that we can reject the null hypothesis of equal variances at the 0.01 significance level.

```
. oneway deldec datenew
```

Source	Analysis of variance			F	Prob > F
	SS	df	MS		
Between groups	15483.3556	2	7741.67779	8.31	0.0004
Within groups	112690.862	121	931.329439		
Total	128174.218	123	1042.06681		

Bartlett's equal-variances test: $\chi^2(2) = 77.2656$ $\text{Prob} > \chi^2 = 0.000$

Bartlett's equal-variances test for the sample of solar-PV applications. The P value (prob>chi2) indicates that we can reject the null hypothesis of equal variances at the 0.01 significance level.

```
. oneway deldec datenew
```

Analysis of variance					
Source	SS	df	MS	F	Prob > F
Between groups	29816.1188	2	14908.0594	15.18	0.0000
Within groups	258286.287	263	982.077138		
Total	288102.406	265	1087.17889		

Bartlett's equal-variances test: chi2(2) = 95.2165 Prob>chi2 = 0.000

Annex 2B

(FOR SOLAR-PV)
Pairwise comparisons of means (unequal variances)

lag	Diff.	Std.Err	Dunnett's C [95% Conf. Interval]	
datenew				
2020 vs 2019	-48.80231	14.35708	-92.54487	-5.059751
2021 vs 2019	-64.21546	14.18622	-107.6771	-20.75381
2021 vs 2020	-15.41315	2.770149	-21.97606	-8.850235

lag	Diff.	Std.Err	Games and Howell [95% Conf. Interval]	
datenew				
2020 vs 2019	-48.80231	14.35708	-92.09845	-5.506163
2021 vs 2019	-64.21546	14.18622	-107.5686	-20.86234
2021 vs 2020	-15.41315	2.770149	-21.94398	-8.882313

lag	Diff.	Std.Err	Tamhane's T2 [95% Conf. Interval]	
datenew				
2020 vs 2019	-48.80231	14.35708	-94.82865	-2.775971
2021 vs 2019	-64.21546	14.18622	-110.4332	-17.99771
2021 vs 2020	-15.41315	2.770149	-22.07156	-8.754732

(FOR WIND)

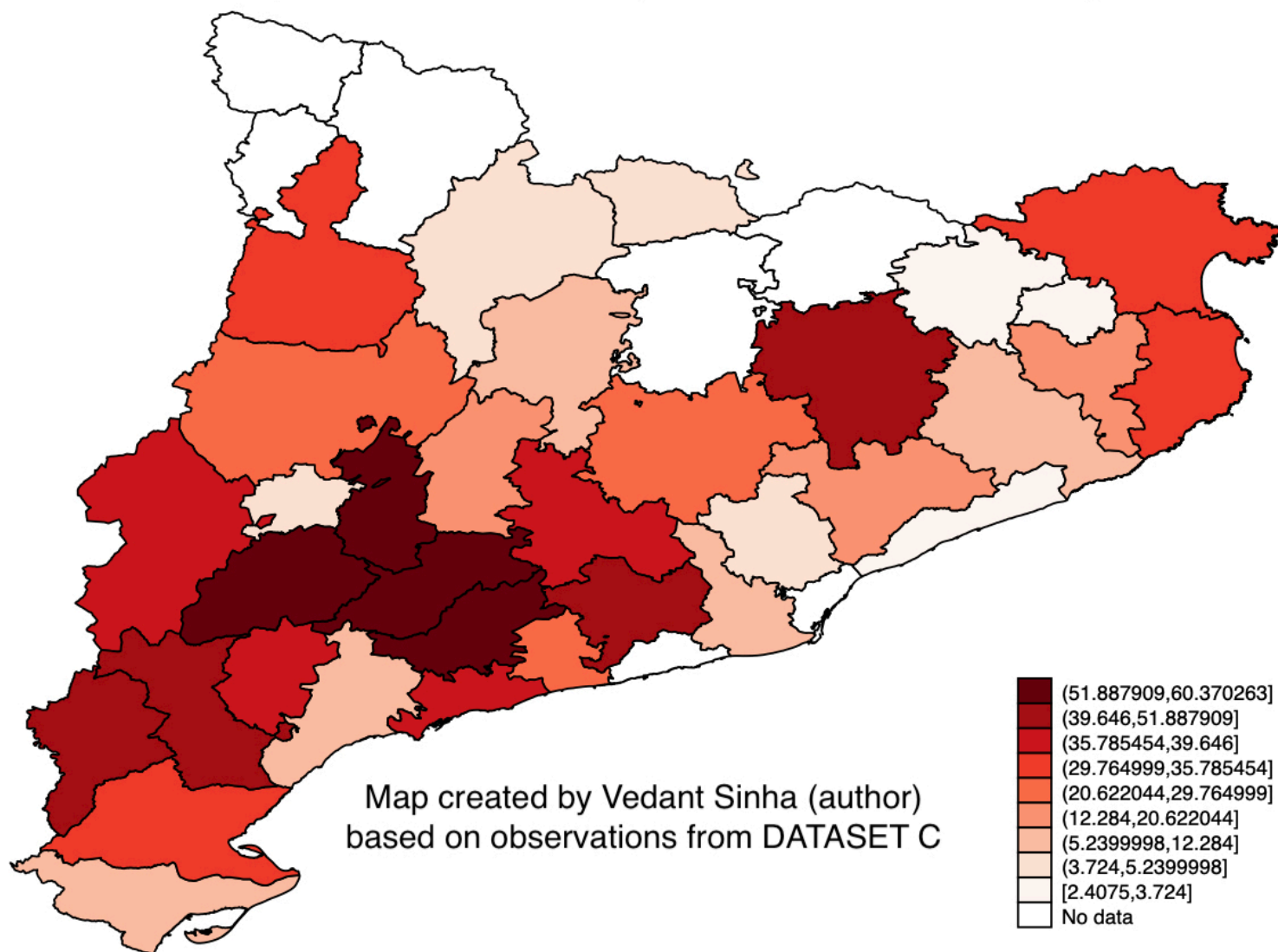
Pairwise comparisons of means (unequal variances)

lag	Diff.	Std.Err	Dunnett's C [95% Conf. Interval]	
datenew				
2020 vs 2019	9.334311	3.734991	.3513619	18.31726
2021 vs 2019	-21.18182	1.752659	-25.73608	-16.62756
2021 vs 2020	-30.51613	3.904711	-39.90656	-21.12569

lag	Diff.	Std.Err	Games and Howell [95% Conf. Interval]	
datenew				
2020 vs 2019	9.334311	3.734991	.4492169	18.2194
2021 vs 2019	-21.18182	1.752659	-25.51345	-16.85018
2021 vs 2020	-30.51613	3.904711	-39.79271	-21.23955

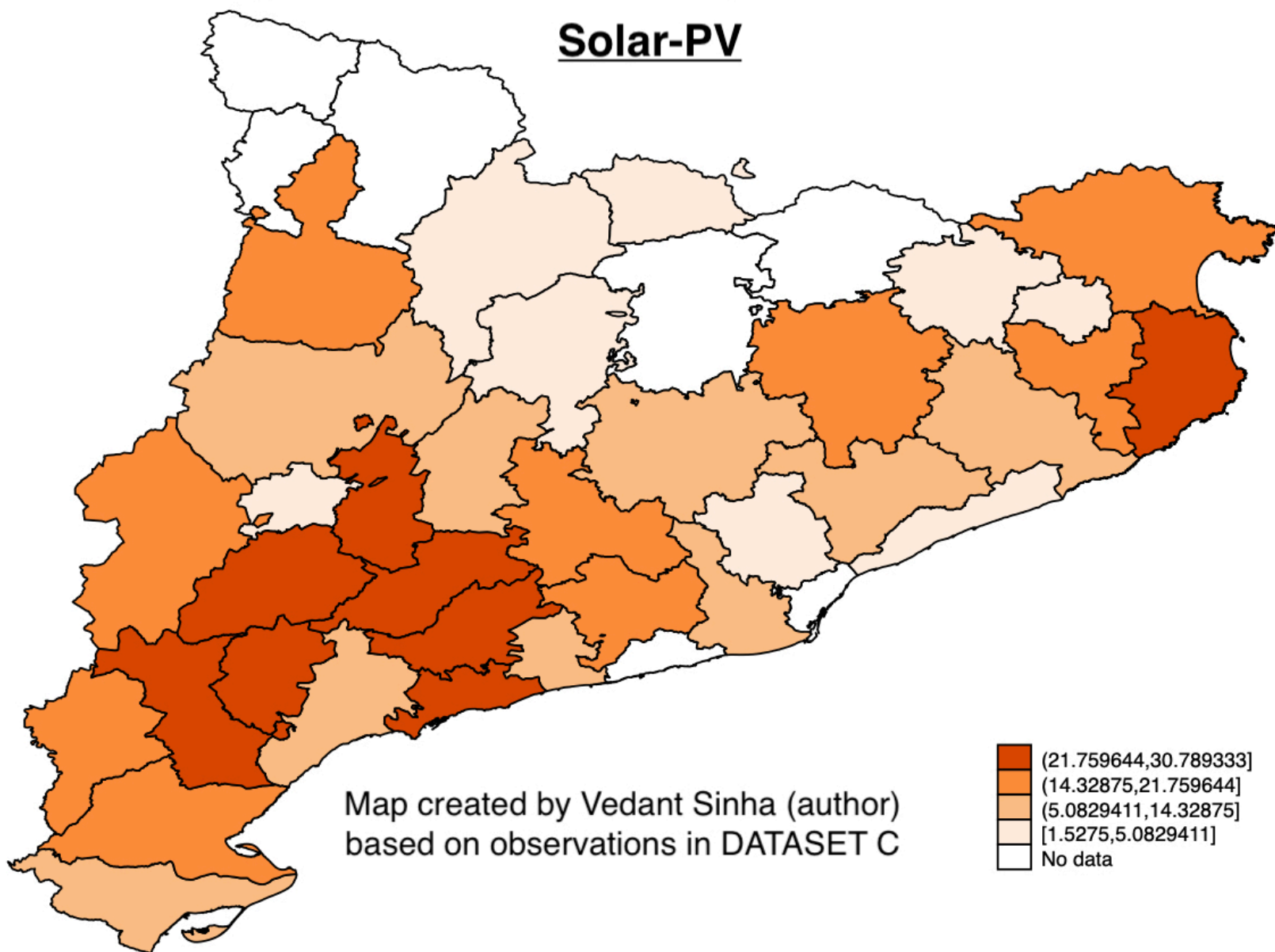
lag	Diff.	Std.Err	Tamhane's T2 [95% Conf. Interval]	
datenew				
2020 vs 2019	9.334311	3.734991	.2653074	18.40331
2021 vs 2019	-21.18182	1.752659	-25.62609	-16.73755
2021 vs 2020	-30.51613	3.904711	-39.98312	-21.04914

Average Land Area Requested For Solar Project (in Ha)

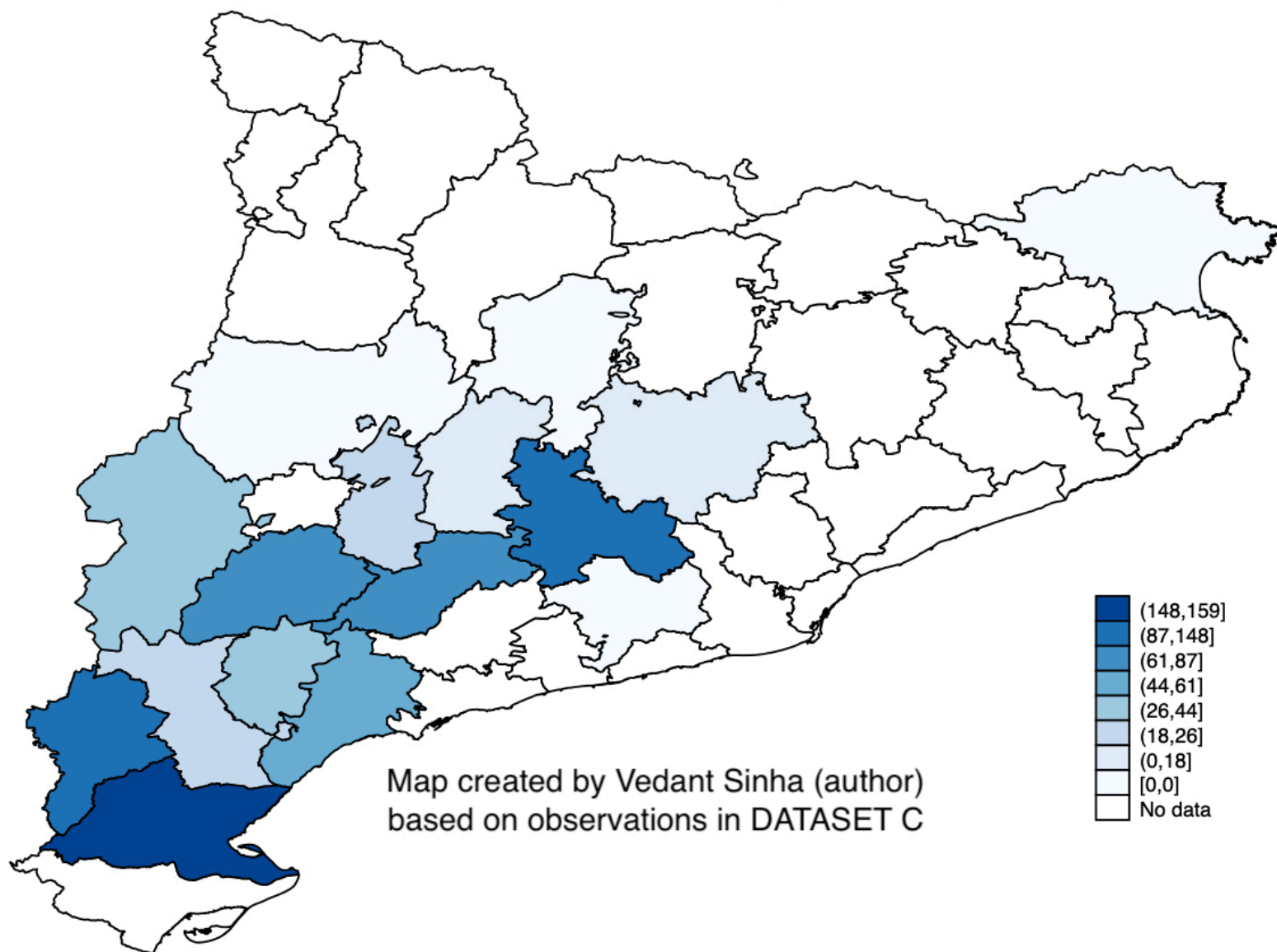


Average Production Capacity Of Installed Project (MW)

Solar-PV

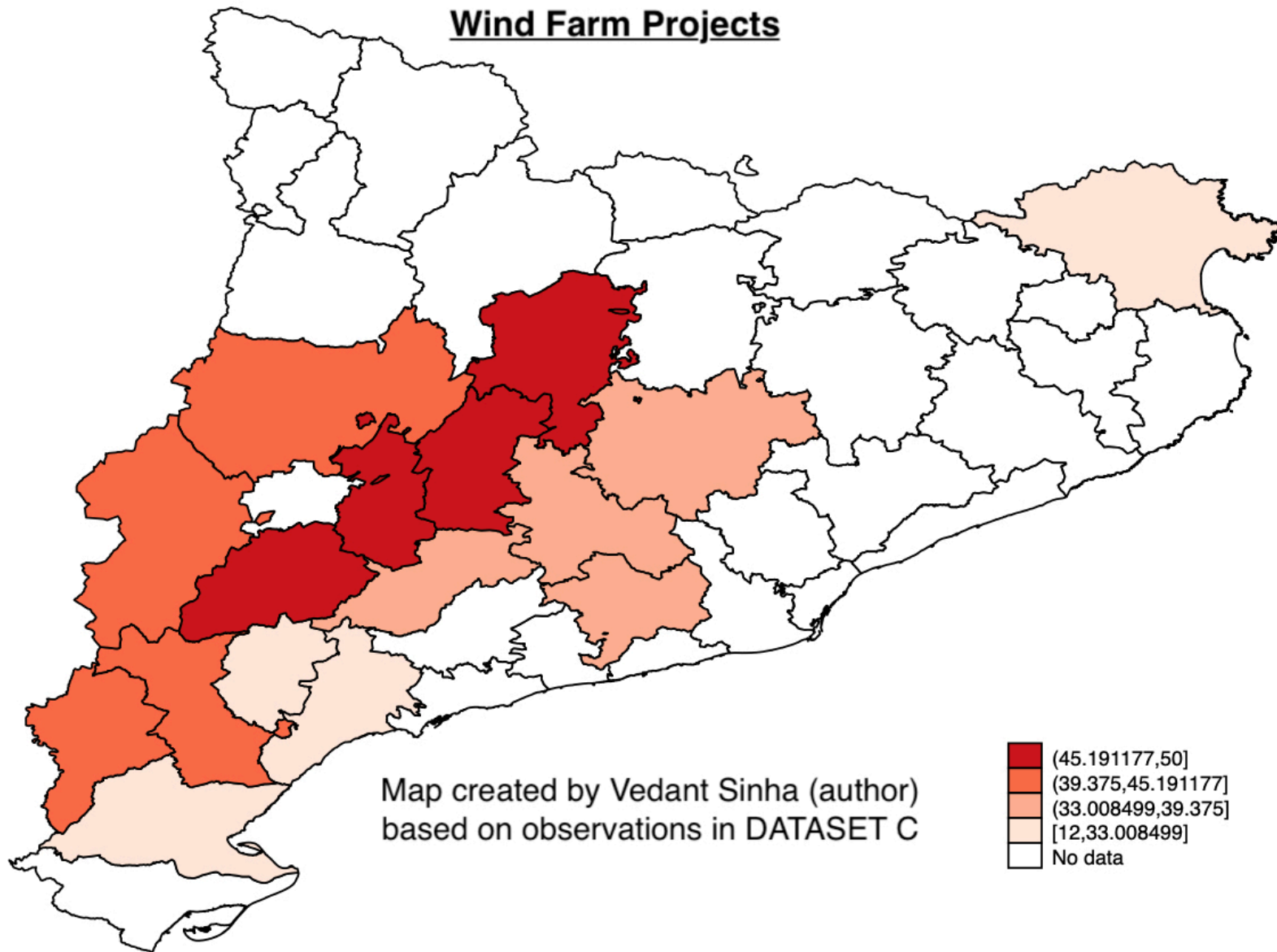


Wind Turbines Already Installed



Average Production Capacity (in MW)

Wind Farm Projects



Annex 3B

Solar-PV project applications filed after the turning point of 439 days. Only 29 non-missing observations on decision delays are available for such applications.

```
. misstable summarize deldec
```

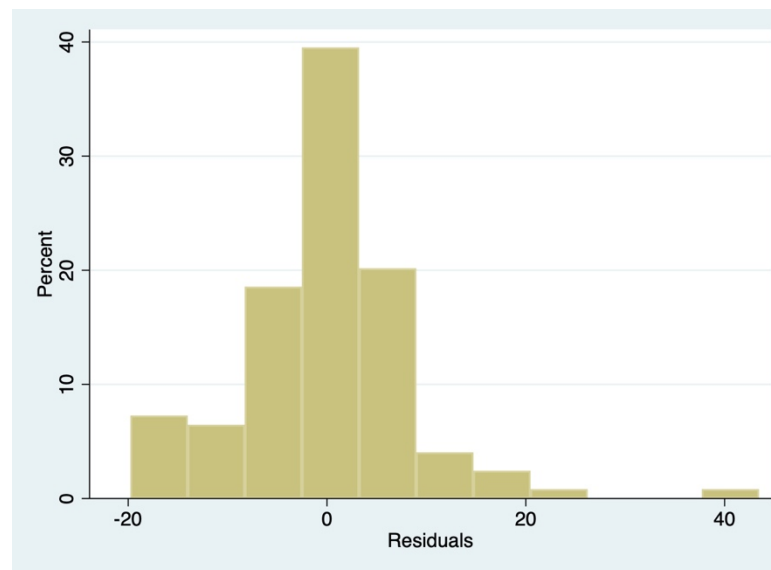
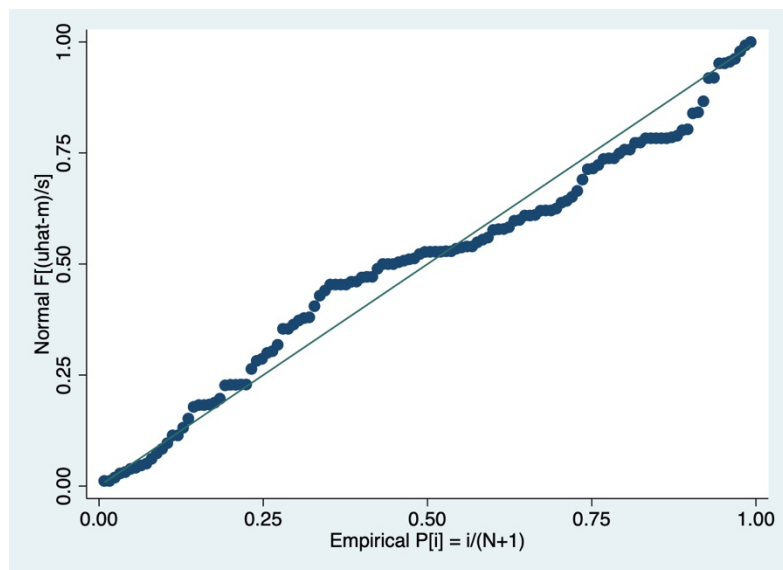
Variable	Obs=.	Obs>.	Obs<.	Obs<.		
				Unique values	Min	Max
deldec	136		29	12	47	90

54 non-missing observations on decision times for wind project applications filed after the turning point of 268 days.

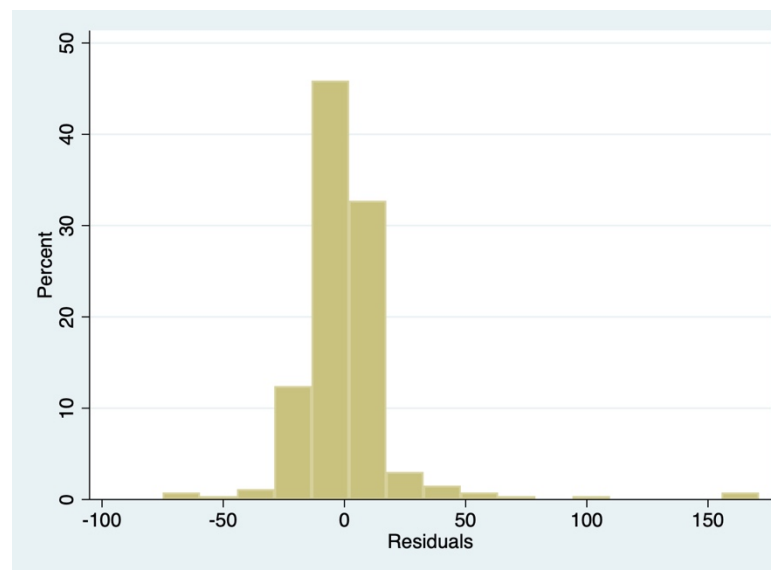
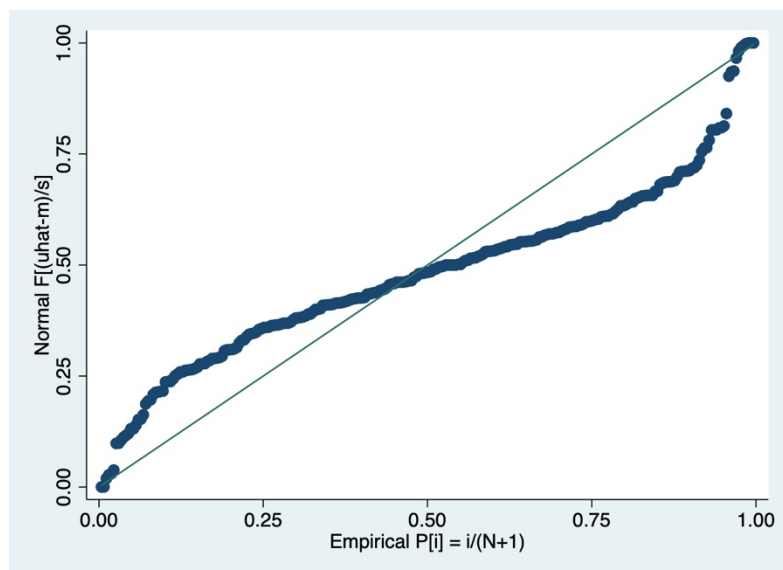
```
. sum deldec
```

Variable	Obs	Mean	Std. dev.	Min	Max
deldec	54	80.22222	8.437711	45	94

Annex 3C



For wind project model-



For solar-PV project
model-

The normal probability plots and the histograms show above that the errors for both our solar-PV and wind farm project model are approximately normally distributed

Annex 3D

```
. pwcorr powerinmw restrixlaw turbinesinop
```

	poweri~w	restri~w	turbin~p
powerinmw	1.0000		
restrixlaw	0.0763	1.0000	
turbinesinop	-0.0908	-0.0501	1.0000

```
. pwcorr restrixlaw powerinmw landarea
```

	restri~w	poweri~w	landarea
restrixlaw	1.0000		
powerinmw	0.1059	1.0000	
landarea	0.0989	0.8992	1.0000

Annex 3E

For wind project. The dategap variable represents the time that passes (in days) between the filing of two successive applications. The smaller the time gap between two applications, an old and a new one, the higher would be the workload and backlog of government bodies processing those applications.

dategap						(unlabeled)
Type: Numeric (float)						
Range: [0,40]			Units: 1			
Unique values: 20			Missing .: 1/124			
Mean: 3.78862						
Std. dev.: 6.88102						
Percentiles:	10%	25%	50%	75%	90%	
	0	0	0	6	13	

dategap						(unlabeled)
Type: Numeric (float)						
Range: [0,15]			Units: 1			
Unique values: 16			Missing .: 1/402			
Mean: 1.5187						
Std. dev.: 2.59428						
Percentiles:	10%	25%	50%	75%	90%	
	0	0	0	2	4	

Wind Model full table

Linear regression

Number of obs = 124
F(18, 102) = .
Prob > F = .
R-squared = 0.9273
Root MSE = 9.5567

deldec	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
restrixlaw	-1.072155	.0667205	-16.07	0.000	-1.204495	-.9398153
quadrestrix	.0017633	.0001553	11.35	0.000	.0014553	.0020714
turbinesinop	.4474756	.2662776	1.68	0.096	-.0806848	.975636
quadrturb	-.0020176	.0016447	-1.23	0.223	-.0052799	.0012448
powerinmw	-.6121404	.4232165	-1.45	0.151	-1.451588	.2273076
quadpower	.0093501	.0064123	1.46	0.148	-.0033687	.022069
location						
Alt Penedès	9.92347	6.149091	1.61	0.110	-2.273222	22.12016
Anoia	-10.92578	6.413519	-1.70	0.092	-23.64696	1.795409
Bages	1.92803	3.324511	0.58	0.563	-4.666122	8.522181
Baix Camp	-12.80927	8.508076	-1.51	0.135	-29.685	4.066463
Baix Ebre	-11.22812	5.791327	-1.94	0.055	-22.71519	.258952
Conca de Barberà	-9.585641	8.272523	-1.16	0.249	-25.99415	6.822869
Garrigues	-6.832387	7.948855	-0.86	0.392	-22.5989	8.93413
Noguera	6.254489	6.27136	1.00	0.321	-6.184724	18.6937
Priorat	-10.64926	5.708908	-1.87	0.065	-21.97285	.6743287
Ribera d'Ebre	6.921531	4.727849	1.46	0.146	-2.456136	16.2992
Segarra	4.604729	4.660298	0.99	0.325	-4.638951	13.84841
Segrià	2.06601	5.230005	0.40	0.694	-8.30768	12.4397
Solsonès	-.2890543	6.471086	-0.04	0.964	-13.12442	12.54631
Terra Alta	0	(omitted)				
Urgell	0	(omitted)				
datenew						
2020	138.0631	5.483324	25.18	0.000	127.187	148.9393
2021	102.9888	6.314202	16.31	0.000	90.46461	115.513
_cons	96.43695	7.454863	12.94	0.000	81.65027	111.2236

Solar-PV Model

Linear regression

Number of obs = 266
 F(36, 225) = .
 Prob > F = .
 R-squared = 0.5596
 Root MSE = 23.747

deldec	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
powerinmw	.1860426	.8260271	0.23	0.822	-1.441696	1.813781
quadpower	-.0065614	.0110529	-0.59	0.553	-.0283418	.015219
landarea	-.0653808	.3137449	-0.21	0.835	-.6836351	.5528734
quadlandarea	.0007447	.0014106	0.53	0.598	-.0020349	.0035243
restrixlaw	-.8785809	.1619558	-5.42	0.000	-1.197725	-.5594367
quadrestrix	.0013459	.0002875	4.68	0.000	.0007793	.0019124
location						
Alt Empordà	-13.56887	5.805878	-2.34	0.020	-25.00972	-2.128015
Alt Penedès	1.472279	3.964817	0.37	0.711	-6.340644	9.285202
Alt Urgell	.4088271	7.527289	0.05	0.957	-14.42417	15.24183
Anoia	-5.250082	5.220072	-1.01	0.316	-15.53656	5.036401
Bages	.4973076	4.513211	0.11	0.912	-8.396261	9.390876
Baix Camp	-7.287811	5.158941	-1.41	0.159	-17.45383	2.878208
Baix Ebre	-3.763875	3.712682	-1.01	0.312	-11.07995	3.5522
Baix Llobregat	2.569615	12.93527	0.20	0.843	-22.92015	28.05938
Baix Penedès	-3.253485	4.515161	-0.72	0.472	-12.1509	5.643925
Cerdanya	-10.85674	4.516303	-2.40	0.017	-19.7564	-1.957077
Conca de Barberà	6.095682	4.262157	1.43	0.154	-2.303168	14.49453
Garrigues	5.118873	3.854348	1.33	0.185	-2.476365	12.71411
Garrotxa	-85.83584	20.83899	-4.12	0.000	-126.9004	-44.77129
Gironès	-40.28623	9.121548	-4.42	0.000	-58.26082	-22.31164
Maresme	.806951	4.205124	0.19	0.848	-7.479513	9.093415
Moianès	-23.00398	8.43462	-2.73	0.007	-39.62493	-6.383023
Montsià	-9.723511	16.96871	-0.57	0.567	-43.16143	23.7144
Noguera	.1546365	7.034368	0.02	0.982	-13.70703	14.0163
Osona	-2.73995	7.682369	-0.36	0.722	-17.87854	12.39865
Pallars Jussà	9.114916	6.388073	1.43	0.155	-3.473186	21.70302
Pla d'Urgell	28.13161	40.98937	0.69	0.493	-52.64055	108.9038
Pla de l'Estany	-26.64214	7.217982	-3.69	0.000	-40.86563	-12.41865
Priorat	-9.795503	5.561366	-1.76	0.080	-20.75453	1.163522
Ribera d'Ebre	6.204813	9.294224	0.67	0.505	-12.11004	24.51967
Segarra	-13.00125	5.874599	-2.21	0.028	-24.57753	-1.424984
Segrià	1.570026	4.182681	0.38	0.708	-6.672212	9.812265
Selva	-19.34814	7.918523	-2.44	0.015	-34.95209	-3.74419
Solsonès	.451848	5.850709	0.08	0.939	-11.07734	11.98104
Tarragonès	-8.756682	4.787092	-1.83	0.069	-18.18995	.6765873
Terra Alta	-2.220539	3.182613	-0.70	0.486	-8.492079	4.051001
Urgell	-2.258106	5.573511	-0.41	0.686	-13.24106	8.72485
Vallès Oriental	-2.028232	4.318234	-0.47	0.639	-10.53759	6.481122
datenew						
2020	47.57517	23.52487	2.02	0.044	1.217925	93.93241
2021	29.97618	22.48579	1.33	0.184	-14.3335	74.28586
_cons	179.1222	21.17868	8.46	0.000	137.3883	220.8561

Annex 3F

Linear regression

Number of obs = 123
 F(20, 99) = .
 Prob > F = .
 R-squared = 0.9313
 Root MSE = 9.4253

deldec	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
restrixlaw	-1.075396	.0643327	-16.72	0.000	-1.203046	-.9477462
quadrestrix	.0017688	.0001489	11.88	0.000	.0014734	.0020643
turbinesinop	.3208294	.2549896	1.26	0.211	-.1851253	.8267841
quadturb	-.0012207	.0015552	-0.78	0.434	-.0043067	.0018652
powerinmw	-.6272191	.4411419	-1.42	0.158	-1.50254	.2481022
quadpower	.009889	.0065305	1.51	0.133	-.0030689	.0228469
dategap	-.5481272	.4112315	-1.33	0.186	-1.3641	.2678454
quaddategap	.0105552	.0162656	0.65	0.518	-.0217193	.0428296
location						
Alt Penedès	7.534995	5.942971	1.27	0.208	-4.257149	19.32714
Anoia	-7.365766	6.021606	-1.22	0.224	-19.31394	4.582407
Bages	1.261214	3.161044	0.40	0.691	-5.010984	7.533412
Baix Camp	-8.329844	7.738161	-1.08	0.284	-23.68403	7.024347
Baix Ebre	-11.30707	5.544749	-2.04	0.044	-22.30905	-.3050837
Conca de Barberà	-5.33006	7.59703	-0.70	0.485	-20.40422	9.744096
Garrigues	-2.661374	7.977534	-0.33	0.739	-18.49053	13.16778
Noguera	4.582527	6.455422	0.71	0.479	-8.226431	17.39148
Priorat	-8.680212	5.728073	-1.52	0.133	-20.04595	2.685528
Ribera d'Ebre	8.595999	4.514721	1.90	0.060	-.362186	17.55418
Segarra	4.750536	4.447065	1.07	0.288	-4.073405	13.57448
Segrià	4.502006	4.921293	0.91	0.363	-5.262907	14.26692
Solsonès	3.937466	8.190579	0.48	0.632	-12.31442	20.18935
Terra Alta	0	(omitted)				
Urgell	0	(omitted)				
datenew						
2020	138.5112	5.044697	27.46	0.000	128.5014	148.521
2021	103.2169	6.125673	16.85	0.000	91.06228	115.3716
_cons	98.57711	8.001041	12.32	0.000	82.70131	114.4529

The increase or decrease in backlog is represented by the dategap variable. It represents the time difference (in days) between two successive applications. It was calculated by subtracting the date of entry for two applications with one succeeding the other. A high value on dategap implies that new applications were coming in quickly and adding to the backlog because of a short time interval between new applications. From the regression table above, we can see that after controlling for a variety of factors, the dategap variable did not have a statistically significant impact on decision delay at the 0.05 significance level (represented by the deldec variable). Even after including a quadratic specific of the dategap variable, we can see that it has no statistically significant impact on deldec at the 0.05 significance level.