Figure 2 in the paper "Multi-Resolution, Multi-Horizon Distributed Solar PVPower Forecasting with Forecast Combinations" is created with the solar forecasting studies listed in Table 1. This table includes the time resolution and the forecasting horizon (in minutes) that has been studied in the papers.

Table 1: Resolution and horizon of 104 solar forecasting studies.

Resolution (min)	Horizon (min)	Author
60	4320	Lorenz et al. (2007)
60	60	Yona et al. (2007)
60	120	
60	180	
60	1440	Lorenz et al. (2008)
60	2880	
60	4320	
60	60	Bacher et al. (2009)
60	120	
60	180	
60	240	
60	300	
60	360	
60	420	
60	480	
60	540	
60	600	
60	660	
60	720	
60	780	
60	840	
60	900	
60	960	
60	1020	
60	1080	
60	1140	
60	1200	
60	1260	
60	1320	
60	1380	
60	1440	
60	1500	
60	1560	
60	1620	
60	1680	
60	1740	
60	1800	

60	1860	
60	1920	
60	1980	
60	2040	
60	2100	
60	2160	
60	1080	Kudo et al. (2009)
60	1860	
15	15	Li et al. (2009)
15	30	
15	40	
1	1	Hassanzadeh et al. (2010)
5	5	
10	10	
15	15	
30	30	
60	60	
60	1440	Mellit & Pavan (2010)
60	1440	Cai et al. (2010)
60	1440	Chen et al. (2011)
60	1440	Chupong & Plangklang (2011)
60	1440	Ding et al. (2011)
60	60	Da Silva Fonseca et al. (2014a)
60	120	Da Silva Fonseca et al. (2014b)
60	1500	
60	4320	Lorenz et al. (2011a)
60	2880	Lorenz et al. (2011b)
60	60	Pelland et al. (2011)
60	120	
60	180	
60	240	
60	300	
60	360	
60	420	
60	480	
60	540	
60	600	
60	660	
60	720	
60	780	
60	840	
60	900	
60	960	
60	1020	
60	1080	

60	1140	
60	1200	
60	1260	
60	1320	
60	1380	
60	1440	
60	1500	
60	1560	
60	1620	
60	1680	
60	1740	
60	1800	
60	1860	
60	1920	
60	1980	
60	2040	
60	2100	
60	2160	
60	2220	
60	2280	
60	2340	
60	2400	
60	2460	
60	2520	
60	2580	
60	2640	
60	2700	
60	2760	
60	2820	
60	2880	
60	1440	Shi et al. (2011)
10	10	Al-Messabi et al. (2012)
10	60	
10	10	Chow et al. (2012)
10	20	
60	1440	Fernandez-Jimenez et al. (2012)
60	720	Mandal et al. (2012)
60	60	Pedro & Coimbra (2012)
60	120	
60	1440	Simonov et al. (2012)
1	5	Yang & Xie (2012)
5	15	
60	60	
60	1440	
10	60	Bouzerdoum et al. (2013)

15	60	Bracale et al. (2013)
60	720	Haque et al. (2013)
60	360	Hossain et al. (2012)
15	60	Jafarzadeh et al. (2013)
15	120	
15	180	
15	15	Lonij et al. (2013)
15	30	
15	45	
15	60	
15	74	
15	90	
60	1440	Monteiro et al. (2013)
60	60	Monteiro et al. (2013)
60	120	
60	180	
60	240	
60	300	
60	360	
60	420	
60	480	
60	540	
60	600	
60	660	
60	720	
60	780	
60	840	
60	900	
60	960	
60	1020	
60	1080	
60	1140	
60	1200	
60	1260	
60	1320	
60	1380	
60	1440	
60	1440	Ogliari et al. (2013)
1440	10080	
1	30	Takahashi & Mori (2014)
0.5	5	Urquhart et al. (2013)
0.5	10	
0.5	15	
60	1440	Yona et al. (2013)

60	60	Zeng & Qiao (2013)
60	120	
60	180	
60	60	Almonacid et al. (2014)
60	1440	Da Silva Fonseca Junior et al. (2014a)
60	1440	Da Silva Fonseca et al. (2014a)
60	1440	Da Silva Fonseca Junior et al. (2014b)
60	1440	Da Silva Fonseca et al. (2014c)
60	1440	Gandelli et al. (2014)
60	1440	Li et al. (2014)
60	1440	Long et al. (2014)
60	2880	
60	4320	
15	15	Lorenz et al. (2014)
60	300	Lorenz et al. (2014)
60	1440	Masa-Bote et al. (2014)
60	1440	Mellit et al. (2014)
15	15	Russo et al. (2014)
15	30	
15	45	
15	60	
10	60	Wu et al. (2014)
15	60	
60	60	
60	1440	Yang et al. (2014)
60	1680	Zamo et al. (2014a)
60	1740	
60	1800	
60	1860	
60	1980	
60	2040	
60	2100	
60	2160	
60	2220	
60	2280	
60	2340	
60	2400	
60	2460	
60	2520	
60	2580	
60	2640	
60	2700	
1440	2880	Zamo et al. (2014b)
60	1440	Alessandrini et al. (2015)

60	2880	
60	4320	
60	60	Alhakeem et al. (2015)
60	180	
60	360	
60	1440	Almeida et al. (2015)
60	60	Bessa et al. (2015)
60	120	
60	180	
60	240	
60	300	
60	360	
5	5	Chu et al. (2015)
5	10	
5	15	
1440	1440	De Felice et al. (2015)
1440	7200	
1440	14400	
60	60	De Giorgi et al. (2016)
60	180	
60	360	
60	720	
60	1440	
60	1440	Dolara et al. (2015)
60	2880	
60	4320	
60	1440	da Silva Fonseca et al. (2015)
60	1440	Huang & Perry (2016)
60	1440	Leva et al. (2017)
43200	43200	Lin & Pai (2016)
0.33333333	0.33333333	Lipperheide et al. (2015)
0.33333333	0.66666667	
0.33333333	1	
0.33333333	1.33333333	
0.33333333	1.66666667	
0.33333333	2	
0.33333333	2.33333333	
0.33333333	2.66666667	
0.33333333	3	
60	1440	Lu et al. (2015)
60	1440	Ramsami & Oree (2015)
60	60	Mohammed et al. (2015)
60	120	
60	180	
60	240	

60	300	
60	360	
60	420	
60	480	
60	540	
60	600	
60	660	
60	720	
60	780	
60	840	
60	900	
60	960	
60	1020	
60	1080	
60	1140	
60	1200	
60	1260	
60	1320	
60	1380	
60	1440	
60	1440	Ramsami & Oree (2015)
1	30	Rana et al. (2015)
1	60	
5	120	
5	180	
30	240	
30	300	
60	720	Schmelas et al. (2015)
60	60	Zhang et al. (2015)
60	1440	
60	60	De Giorgi et al. (2016)
60	180	
60	360	
60	720	
60	1440	
60	60	Do et al. (2016)
1	10	Golestaneh et al. (2016)
1	60	
60	1440	Larson et al. (2016)
15	15	Li et al. (2016)
15	60	
15	1440	
60	60	Nagy et al. (2016)
60	1440	
5	5	Rana et al. (2016)

5	10	
5	15	
5	20	
5	25	
5	30	
5	35	
5	40	
5	45	
5	50	
5	55	
5	60	
1	1	Soubdhan et al. (2016)
5	5	
10	10	
30	30	
60	60	
60	1440	Sperati et al. (2016)
60	2880	
60	4320	
15	15	Vaz et al. (2016)
15	60	
60	60	Abdel-Nasser & Mahmoud (2017)
15	1440	Raza et al. (2018)
60	1440	Nespoli et al. (2019)
60	840	Lee & Kim (2019)
5	1440	Liu et al. (2019)
60	60	VanDeventer et al. (2019)
60	60	Acikgoz (2022)
60	120	
60	180	
43200	43200	Heo et al. (2021)
60	60	Korkmaz (2021)
60	120	
60	180	
5	1440	Qu et al. (2021)
60	60	Aicardi et al. (2022)
60	120	
60	180	
60	240	
60	300	
15	1440	De Hoog et al. (2021)
15	60	du Plessis et al. (2021)
15	120	
15	180	
15	240	

15	300	
15	360	
60	60	Kumari & Toshniwal (2021)
15	1440	Mayer & Gróf (2021)
15	2880	
60	1440	Zang et al. (2020)
60	1440	Lan et al. (2019)
60	60	Lai et al. (2021)
15	15	Cannizzaro et al. (2021)
15	30	
15	45	
15	60	
15	180	
15	720	
15	1440	
60	60	
60	180	
60	720	
60	1440	
15	15	Castangia et al. (2021)
15	30	
15	45	
15	60	
15	75	
15	90	
15	105	
15	120	
15	135	
15	150	
15	165	
15	180	
15	195	
15	210	
15	225	
15	240	

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