

## A APPENDIX

### A.1 EXPERIMENT DETAILS

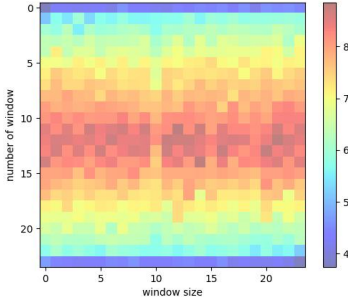
**Datasets characteristics** The dataset is characterised as follows in Appendix Table 7. We follow standard protocol ((Nie et al., 2023)) and split all datasets into training, validation and test set in chronological order by the ratio of 6:2:2 for the ETT dataset and 7:1:2 for the other datasets. Frequency indicates the sampling time difference between neighbouring time steps.

Table 7: Summary of experiment datasets.

Datasets	ETTM1	ETTM2	ETTh1	ETTh2	weather	ILI	exchange	traffic	ECL
Channels	7	7	7	7	21	7	7	862	321
Lengths	69680	69680	17420	17420	52696	966	7588	17544	26304
Frequency	15min	15min	1h	1h	10min	7day	1day	1h	1h

**Decomposition Correlation Block** Convolutional networks can extract the local correlation between period-trend and oscillation terms based on the weights of the convolutional kernel. A higher weight means the relatively higher correlation. As shown in Appendix Figure 5, we visualise the heat map after the convolution kernel and can see that the model focuses more on the local periodicity of the intermediate window from 5 to 18 for the 2D tensor. In terms of periods, our model pays more attention to the previous 10-15 periods when forecasting.

Figure 5: The features after the Decomposition Correlation Block. In the figure, the rows represent periodic windows and the columns represent the number of windows.



**Two-Dimensional Period Decomposition** We take real data from ETTh1 for TDPD and get the period-trend and oscillation terms. As demonstrated in Appendix Figure 6, each column of the period-trend keeps essentially the same trend. In contrast, the oscillation term is characterised by disorganized distributions.

### A.2 FULL RESULTS

Due to the space limitation of the main text, we place the full results of all experiments in the following: multivariate long-input length forecasting in Appendix Table 8, multivariate short-input length forecasting in Appendix Table 9, univariate long-input length forecasting in Appendix Table 10.

### A.3 PREDICTION VISUALIZATION

As shown in Appendix Figure 7, Figure 8, Figure 9, Figure 10, Figure 11, Figure 12, Figure 13, we visualize the long-term univariate prediction results of WinNet from the test set of all eight datasets. Here, we predict  $\{24, 36, 48, 60\}$  steps on ILI dataset and  $\{96, 192, 336, 720\}$  steps on other datasets. It can be seen that the proposed model can achieve the best results.

Figure 6: Showcase of the TDPD.

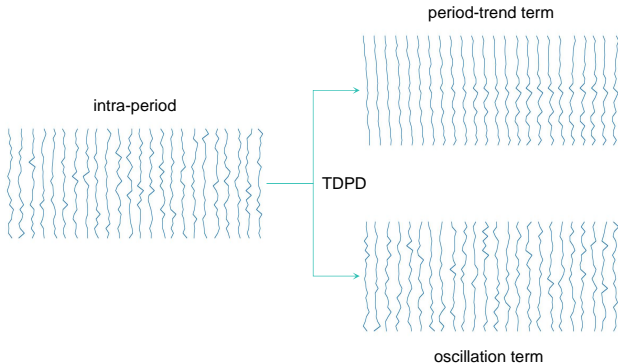
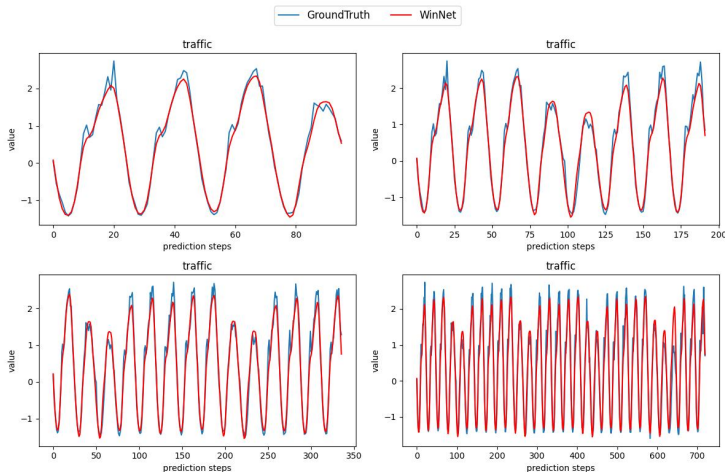


Figure 7: Visualization of prediction on traffic with the look-back window L=336.



#### A.4 MORE ABLATION STUDIES

**Window size** As shown in Appendix Figure 14, since the window size is approximated as a multiple of the multi-short periods, it can also have implications for experimental performance of the proposed model. For example, for a sequence with top-4 periods  $\{4, 6, 8, 12\}$ , we can take 24 as the window size, but it is equally possible to approximate the period of 8 as 9 and take 36 as the final window size. The top-4 periods for the datasets can be seen in Table 1. As demonstrated in Appendix Table 12, the proposed approach can achieve better performance for the periodic window of 24.

**CNN Kernel** The convolution kernel size determines the receptive field for extracting periodicity from the sequence. In this section, we consider the model performance with different kernel size to  $\{3, 5, 7\}$ , respectively, and the experimental results are shown in Appendix Table 13. It can be found that a larger convolution kernel causes lower prediction accuracy. The results can be attributed that after the proposed period decomposition, the period of the reshaped sequence is smaller and a larger kernel excessively extracts the temporal correlation from other periods to degrade the model performance. More ablation study results are available in Appendix Table 15.

Table 8: Full results for multivariate long-input length prediction. We compare extensive competitive models under different prediction lengths. The input sequence length is set to 104 for the ILI dataset and 512 for the others. Avg is the averaged result from all four prediction lengths. The 1st count indicates the numbers of best performance.

Methods	WinNet (Ours)		PatchTST (2023)		TimesNet* (2023)		MICN* (2023)		Crossformer (2023)		DLinear (2023)		FEDformer (2022)		Autoformer (2021)			
Metric	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE		
ETTm1	96	<b>0.283</b>	<b>0.335</b>	0.293	0.346	0.377	0.397	0.305	0.354	0.302	0.359	<b>0.290</b>	<b>0.342</b>	0.326	0.390	0.510	0.492	
	192	<b>0.324</b>	<b>0.360</b>	0.333	0.370	0.389	0.401	0.362	0.399	0.341	0.387	<b>0.332</b>	<b>0.369</b>	0.365	0.415	0.514	0.495	
	336	<b>0.357</b>	<b>0.379</b>	0.369	0.392	0.393	0.414	0.382	0.405	0.419	0.432	<b>0.366</b>	<b>0.392</b>	0.392	0.425	0.510	0.492	
	720	<b>0.416</b>	<b>0.411</b>	<b>0.416</b>	<b>0.420</b>	0.470	0.458	0.441	0.441	0.637	0.577	<b>0.420</b>	0.424	0.446	0.458	0.527	0.493	
	Avg	<b>0.345</b>	<b>0.371</b>	<b>0.352</b>	0.382	0.407	0.417	0.372	0.399	0.424	0.438	<b>0.352</b>	<b>0.381</b>	0.382	0.422	0.515	0.493	
ETTm2	96	<b>0.160</b>	<b>0.251</b>	<b>0.166</b>	<b>0.256</b>	0.201	0.280	0.189	0.287	0.305	0.361	0.167	0.260	0.180	0.271	0.205	0.293	
	192	<b>0.212</b>	<b>0.287</b>	<b>0.223</b>	<b>0.296</b>	0.242	0.313	0.239	0.323	0.355	0.391	0.224	0.303	0.252	0.318	0.278	0.336	
	336	<b>0.261</b>	<b>0.322</b>	<b>0.274</b>	<b>0.329</b>	0.310	0.356	0.348	0.385	0.420	0.431	0.281	0.342	0.324	0.364	0.343	0.379	
	720	<b>0.359</b>	<b>0.381</b>	<b>0.362</b>	<b>0.385</b>	0.381	0.396	0.421	0.434	0.592	0.548	0.397	0.421	0.410	0.420	0.414	0.419	
	Avg	<b>0.248</b>	<b>0.310</b>	<b>0.256</b>	<b>0.316</b>	0.283	0.336	0.299	0.357	0.418	0.432	0.267	0.331	0.291	0.343	0.310	0.356	
ETTth1	96	<b>0.362</b>	<b>0.390</b>	0.379	0.401	0.460	0.464	0.404	0.429	0.394	0.418	<b>0.375</b>	<b>0.399</b>	0.376	0.415	0.435	0.446	
	192	<b>0.394</b>	<b>0.410</b>	0.413	0.429	0.458	0.462	0.511	0.506	0.423	0.436	<b>0.412</b>	<b>0.420</b>	0.423	0.446	0.456	0.457	
	336	<b>0.419</b>	<b>0.426</b>	<b>0.435</b>	<b>0.436</b>	0.523	0.501	0.482	0.489	0.438	0.451	0.439	0.443	0.444	0.462	0.486	0.487	
	720	<b>0.436</b>	<b>0.453</b>	<b>0.446</b>	<b>0.464</b>	0.502	0.497	0.697	0.631	0.508	0.514	0.472	0.490	0.469	0.492	0.515	0.517	
	Avg	<b>0.402</b>	<b>0.419</b>	<b>0.418</b>	<b>0.432</b>	0.485	0.481	0.523	0.513	0.440	0.454	0.424	0.438	0.428	0.453	0.473	0.476	
ETTth2	96	<b>0.267</b>	<b>0.332</b>	<b>0.274</b>	<b>0.337</b>	0.338	0.397	0.290	0.356	0.395	0.417	0.289	0.353	0.332	0.374	0.332	0.368	
	192	<b>0.322</b>	<b>0.372</b>	<b>0.338</b>	<b>0.376</b>	0.422	0.446	0.415	0.441	0.427	0.438	0.383	0.418	0.407	0.446	0.426	0.434	
	336	<b>0.351</b>	0.401	<b>0.363</b>	<b>0.397</b>	0.431	0.460	0.627	0.573	0.449	0.459	0.448	0.465	0.400	0.447	0.477	0.479	
	720	<b>0.389</b>	0.436	<b>0.393</b>	<b>0.430</b>	0.467	0.480	1.340	0.858	0.501	0.509	0.605	0.551	0.412	0.469	0.453	0.490	
	Avg	<b>0.332</b>	<b>0.385</b>	<b>0.342</b>	<b>0.385</b>	0.414	0.445	0.668	0.557	0.443	0.455	0.431	0.446	0.387	<b>0.434</b>	0.422	0.442	
ILI	24	1.985	0.905	<b>1.522</b>	<b>0.814</b>	2.500	1.055	2.559	1.099	3.383	1.249	2.215	1.081	2.624	1.095	2.906	1.182	
	36	1.897	0.900	<b>1.430</b>	<b>0.834</b>	2.222	1.007	2.483	1.023	3.151	1.157	1.963	0.963	2.516	1.021	2.585	1.038	
	48	1.868	0.910	<b>1.673</b>	<b>0.854</b>	2.304	1.043	2.371	1.007	3.386	1.186	2.130	1.024	2.505	1.041	3.024	1.145	
	60	1.928	0.933	<b>1.529</b>	<b>0.862</b>	2.354	1.046	2.694	1.112	3.658	1.268	2.368	1.096	2.742	1.122	2.761	1.114	
	Avg	1.919	0.912	<b>1.538</b>	<b>0.841</b>	2.345	1.037	2.526	1.060	3.394	1.215	2.169	1.041	2.596	1.069	2.819	1.119	
weather	96	<b>0.143</b>	<b>0.198</b>	0.152	<b>0.199</b>	0.163	0.223	0.170	0.235	<b>0.147</b>	0.211	0.176	0.237	0.238	0.314	0.249	0.329	
	192	<b>0.188</b>	<b>0.240</b>	0.197	<b>0.243</b>	0.218	0.266	0.214	0.277	0.194	0.261	<b>0.192</b>	0.246	0.275	0.329	0.325	0.370	
	336	<b>0.235</b>	<b>0.280</b>	0.249	<b>0.283</b>	0.280	0.306	0.278	0.326	0.246	0.306	<b>0.240</b>	0.287	0.339	0.377	0.351	0.391	
	720	<b>0.310</b>	0.336	0.320	<b>0.335</b>	0.349	0.356	0.318	0.363	0.322	0.363	<b>0.316</b>	0.352	0.389	0.409	0.415	0.426	
	Avg	<b>0.219</b>	<b>0.263</b>	<b>0.229</b>	<b>0.265</b>	0.252	0.287	0.245	0.300	<b>0.227</b>	0.285	0.231	0.280	0.310	0.357	0.335	0.379	
traffic	96	0.394	0.274	<b>0.367</b>	<b>0.251</b>	0.603	0.328	0.461	0.290	0.489	0.276	0.410	0.282	0.576	0.359	0.597	0.371	
	192	0.407	0.279	<b>0.385</b>	<b>0.259</b>	0.610	0.329	0.482	0.302	0.503	0.281	0.423	0.287	0.610	0.380	0.607	0.382	
	336	0.416	0.283	<b>0.398</b>	<b>0.265</b>	0.619	0.330	0.487	0.300	0.528	0.292	0.436	0.296	0.608	0.375	0.623	0.387	
	720	0.453	0.305	<b>0.434</b>	<b>0.287</b>	0.632	0.352	0.527	0.310	0.593	0.326	0.466	0.315	0.621	0.375	0.639	0.395	
	Avg	0.417	0.285	<b>0.396</b>	<b>0.265</b>	0.616	0.334	0.489	0.300	0.528	0.293	0.433	0.295	0.603	0.372	0.616	0.383	
ECL	96	<b>0.130</b>	<b>0.226</b>	<b>0.130</b>	<b>0.222</b>	0.181	0.281	0.162	0.272	0.198	0.292	0.140	0.237	0.186	0.302	0.196	0.313	
	192	<b>0.147</b>	<b>0.240</b>	<b>0.148</b>	<b>0.240</b>	0.193	0.293	0.176	0.285	0.266	0.330	0.153	0.249	0.197	0.311	0.211	0.324	
	336	<b>0.163</b>	<b>0.257</b>	<b>0.167</b>	<b>0.261</b>	0.205	0.312	0.194	0.301	0.353	0.384	0.169	0.267	0.213	0.328	0.214	0.327	
	720	<b>0.198</b>	<b>0.290</b>	<b>0.202</b>	<b>0.291</b>	0.222	0.320	0.222	0.327	0.400	0.416	0.203	0.301	0.233	0.344	0.236	0.342	
	Avg	<b>0.159</b>	<b>0.253</b>	<b>0.161</b>	<b>0.253</b>	0.200	0.301	0.188	0.296	0.304	0.355	0.166	<b>0.263</b>	0.207	0.321	0.214	0.326	
1st count	<b>56</b>		<b>29</b>		0		0		0		0		0		0		0	

\* We replace the input length L=512 in TimesNet and MICN for a fair comparison. Other experimental results are taken from the PatchTST and PETformer (Lin et al., 2023).

Table 9: All results for multivariate short-input length prediction. The input sequence length is set to 36 for the ILI dataset and 96 for the others. Avg is the averaged result from all four prediction lengths. The 1st count indicates the numbers of best performance.

Methods		WinNet (Ours)	PatchTST* (2023)	TimesNet (2023)	MICN (2023)	Crossformer* (2023)	DLinear (2023)	FEDformer (2022)	Autoformer (2021)
Metric		MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE
ETTm1	96	<b>0.316 0.356</b>	0.320 0.359	<b>0.316</b> 0.362	0.338 0.375	0.355 0.391	0.345 0.372	0.379 0.419	0.505 0.475
	192	0.368 <b>0.383</b>	0.364 0.381	<b>0.363</b> 0.390	0.374 0.387	0.416 0.433	0.380 0.389	0.426 0.441	0.553 0.496
	336	<b>0.398 0.402</b>	<b>0.391 0.401</b>	0.408 0.426	0.410 0.411	0.486 0.479	0.413 0.413	0.445 0.459	0.621 0.537
	720	<b>0.443 0.431</b>	<b>0.455 0.439</b>	0.481 0.476	0.478 0.450	0.624 0.570	0.474 0.453	0.543 0.490	0.671 0.561
	Avg	<b>0.381 0.385</b>	<b>0.382 0.395</b>	0.392 0.413	0.400 0.405	0.470 0.468	0.403 0.406	0.448 0.452	0.587 0.517
ETTm2	96	<b>0.171 0.252</b>	0.181 0.265	<b>0.179</b> 0.275	0.187 0.267	0.356 0.388	0.193 0.292	0.203 0.287	0.255 0.339
	192	<b>0.240 0.300</b>	<b>0.247 0.307</b>	0.307 0.376	0.249 0.309	0.422 0.440	0.284 0.362	0.269 0.328	0.281 0.340
	336	<b>0.295 0.336</b>	0.308 0.348	0.325 0.388	0.321 0.351	0.507 0.494	0.369 0.427	0.325 0.366	0.339 0.372
	720	<b>0.399 0.395</b>	<b>0.406 0.401</b>	0.502 0.490	0.408 0.403	0.598 0.552	0.554 0.522	0.421 0.415	0.433 0.432
	Avg	<b>0.276 0.320</b>	<b>0.285 0.330</b>	0.328 0.382	0.291 0.332	0.470 0.468	0.350 0.400	0.304 0.349	0.327 0.370
ETTth1	96	<b>0.379 0.388</b>	<b>0.392 0.408</b>	0.421 0.431	0.389 0.412	0.409 0.432	0.386 0.400	<b>0.376</b> 0.419	0.449 0.459
	192	<b>0.432 0.418</b>	0.450 0.433	0.474 0.487	0.442 0.442	0.458 0.459	0.437 0.432	<b>0.420</b> 0.448	0.500 0.482
	336	<b>0.474 0.439</b>	0.518 0.477	0.569 0.551	0.491 0.469	0.509 0.492	0.481 <b>0.459</b>	<b>0.459</b> 0.465	0.521 0.496
	720	<b>0.472 0.457</b>	0.522 <b>0.490</b>	0.770 0.672	0.521 0.500	0.696 0.632	0.519 0.516	0.506 0.507	0.514 0.512
	Avg	<b>0.439 0.425</b>	0.470 0.452	0.558 0.535	0.460 0.455	0.518 0.503	0.455 <b>0.451</b>	<b>0.440</b> 0.459	0.496 0.487
ETTth2	96	<b>0.289 0.337</b>	<b>0.297 0.347</b>	0.299 0.364	0.340 0.374	0.402 0.425	0.333 0.387	0.358 0.397	0.346 0.388
	192	<b>0.375 0.391</b>	<b>0.390 0.403</b>	0.441 0.454	0.402 0.414	0.452 0.456	0.477 0.476	0.429 0.439	0.456 0.452
	336	<b>0.416 0.428</b>	<b>0.417 0.429</b>	0.654 0.567	0.452 0.452	0.533 0.506	0.594 0.541	0.496 0.487	0.482 0.486
	720	<b>0.423 0.445</b>	<b>0.432 0.448</b>	0.956 0.716	0.462 0.468	0.577 0.557	0.831 0.657	0.463 0.474	0.515 0.511
	Avg	<b>0.375 0.400</b>	<b>0.384 0.406</b>	0.587 0.525	0.414 0.427	0.491 0.486	0.558 0.515	0.436 <b>0.449</b>	0.449 0.459
ILI	24	2.445 0.963	<b>1.743 0.814</b>	2.684 1.112	<b>2.317 0.934</b>	4.721 1.524	2.398 1.040	3.228 1.260	3.483 1.287
	36	2.465 1.008	<b>1.579 0.804</b>	2.667 1.068	<b>1.972 0.920</b>	4.148 1.379	2.646 1.088	2.679 1.080	3.103 1.148
	48	2.296 0.961	<b>2.199 0.897</b>	2.558 1.052	<b>2.238 0.940</b>	4.023 1.354	2.614 1.086	2.622 1.078	2.669 1.085
	60	2.348 0.977	<b>1.813 0.868</b>	2.747 1.110	<b>2.027 0.928</b>	4.114 1.369	2.804 1.146	2.857 1.157	2.770 1.125
	Avg	2.388 0.977	<b>1.833 0.845</b>	2.664 1.085	<b>2.138 0.930</b>	4.251 1.406	2.615 1.090	2.846 1.143	3.006 1.161
exchange	96	<b>0.082 0.198</b>	<b>0.082 0.198</b>	0.099 0.240	0.107 0.234	0.253 0.364	<b>0.088 0.218</b>	0.148 0.278	0.197 0.323
	192	<b>0.173 0.294</b>	<b>0.173 0.295</b>	0.198 0.354	0.226 0.344	0.482 0.517	<b>0.176</b> 0.315	0.271 0.380	0.300 0.369
	336	0.327 <b>0.412</b>	0.333 <b>0.415</b>	<b>0.302</b> 0.447	0.367 0.448	0.908 0.748	<b>0.313</b> 0.427	0.460 0.500	0.509 0.524
	720	0.911 0.722	0.880 0.700	<b>0.738 0.662</b>	0.964 0.746	1.414 0.975	<b>0.839 0.695</b>	1.195 0.841	1.447 0.941
	Avg	0.373 <b>0.406</b>	<b>0.367 0.402</b>	<b>0.334</b> 0.425	0.416 0.443	0.764 0.651	<b>0.354</b> 0.413	0.518 0.499	0.613 0.539
weather	96	<b>0.164</b> 0.223	0.177 <b>0.218</b>	<b>0.161</b> 0.229	0.172 <b>0.220</b>	<b>0.162</b> 0.231	0.196 0.255	0.217 0.296	0.266 0.336
	192	<b>0.213</b> 0.268	0.225 <b>0.259</b>	0.220 0.281	<b>0.219 0.261</b>	<b>0.211</b> 0.281	0.237 0.296	0.276 0.336	0.307 0.367
	336	<b>0.271</b> 0.313	0.275 <b>0.296</b>	0.278 0.331	0.280 <b>0.306</b>	<b>0.270</b> 0.328	0.283 0.335	0.339 0.380	0.359 0.395
	720	0.354 0.370	0.351 <b>0.346</b>	<b>0.311 0.356</b>	0.365 0.359	0.352 0.382	<b>0.345</b> 0.381	0.403 0.428	0.419 0.428
	Avg	0.250 0.293	0.257 <b>0.279</b>	<b>0.242</b> 0.299	0.259 <b>0.286</b>	<b>0.248</b> 0.305	0.265 0.316	0.308 0.360	0.337 0.381
traffic	96	<b>0.421</b> 0.344	0.540 0.357	0.519 <b>0.309</b>	0.593 0.321	<b>0.512 0.288</b>	0.650 0.396	0.587 0.366	0.613 0.388
	192	<b>0.452</b> 0.350	<b>0.536</b> 0.352	0.537 <b>0.315</b>	0.617 0.336	0.538 <b>0.297</b>	0.598 0.370	0.604 0.373	0.616 0.382
	336	<b>0.475</b> 0.362	0.547 0.355	<b>0.534 0.313</b>	0.629 0.336	0.569 <b>0.315</b>	0.605 0.373	0.621 0.383	0.622 0.337
	720	<b>0.483</b> 0.371	<b>0.541 0.330</b>	0.577 <b>0.325</b>	0.640 0.350	0.613 0.336	0.645 0.394	0.626 0.382	0.660 0.408
	Avg	<b>0.457</b> 0.356	<b>0.541</b> 0.348	<b>0.541 0.315</b>	0.619 0.335	0.558 <b>0.309</b>	0.624 0.383	0.609 0.376	0.627 0.378
ECL	96	<b>0.167 0.259</b>	0.208 0.297	<b>0.164 0.269</b>	0.168 0.272	0.224 0.310	0.197 0.282	0.193 0.308	0.201 0.317
	192	<b>0.177 0.265</b>	0.198 0.288	<b>0.177 0.285</b>	<b>0.184</b> 0.289	0.281 0.345	0.196 <b>0.285</b>	0.201 0.315	0.222 0.334
	336	<b>0.193 0.282</b>	0.200 <b>0.285</b>	<b>0.193</b> 0.304	<b>0.198</b> 0.300	0.351 0.394	0.209 0.301	0.214 0.329	0.231 0.338
	720	0.232 <b>0.317</b>	0.241 <b>0.318</b>	<b>0.212</b> 0.321	<b>0.220</b> 0.320	0.426 0.439	0.245 0.333	0.246 0.355	0.254 0.361
	Avg	<b>0.192 0.280</b>	0.211 0.297	<b>0.186 0.294</b>	<b>0.192</b> 0.295	0.320 0.372	0.211 0.300	0.213 0.326	0.227 0.337
1st count		<b>50</b>	<b>21</b>	16	0	5	0	3	0

\* We replace the input length L=96 in PatchTST and Crossformer for a fair comparison. Other experimental results are taken from the TimesNet and MICN.

Table 10: All results for univariate long-input length prediction. The input sequence length is set to 104 for the ILI dataset and 336 for the others. Avg is the averaged result from all four prediction lengths. The 1st count indicates the numbers of the performance.

Methods		WinNet (Ours)		PatchTST (2023)		TimesNet* (2023)		MICN* (2023)		DLinear (2023)		FEDformer (2022)		Autoformer (2021)	
Metric		MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
ETTm1	96	<b>0.025</b>	<b>0.120</b>	0.026	0.121	0.028	0.126	0.027	0.123	0.028	0.123	0.033	0.140	0.056	0.183
	192	<b>0.038</b>	<b>0.148</b>	0.039	0.150	0.048	0.167	0.043	0.154	0.045	0.156	0.058	0.186	0.081	0.216
	336	<b>0.051</b>	<b>0.171</b>	0.053	0.173	0.060	0.188	0.052	0.173	0.061	0.182	0.084	0.231	0.076	0.218
	720	<b>0.062</b>	<b>0.189</b>	0.074	0.207	0.076	0.213	0.075	0.206	0.080	0.210	0.102	0.250	0.110	0.267
	Avg	<b>0.044</b>	<b>0.157</b>	0.048	0.162	0.053	0.173	0.049	0.164	0.053	0.167	0.069	0.201	0.080	0.221
ETTm2	96	<b>0.062</b>	<b>0.181</b>	0.065	0.186	0.076	0.206	<b>0.063</b>	0.183	<b>0.063</b>	<b>0.183</b>	0.067	0.198	0.065	0.189
	192	<b>0.090</b>	<b>0.224</b>	0.094	0.231	0.107	0.251	0.091	0.225	<b>0.092</b>	<b>0.227</b>	0.102	0.245	0.118	0.256
	336	<b>0.116</b>	<b>0.258</b>	0.120	0.265	0.135	0.284	0.121	0.265	<b>0.119</b>	<b>0.261</b>	0.130	0.279	0.154	0.305
	720	<b>0.168</b>	<b>0.318</b>	0.171	0.322	0.210	0.362	0.172	0.317	0.175	<b>0.320</b>	0.178	0.325	0.182	0.335
	Avg	<b>0.109</b>	<b>0.245</b>	0.112	0.251	0.132	0.275	0.111	0.247	<b>0.112</b>	<b>0.247</b>	0.119	0.261	0.129	0.271
ETTth1	96	<b>0.052</b>	<b>0.176</b>	0.055	0.179	0.062	0.195	0.059	0.190	0.056	0.180	0.079	0.215	0.071	0.206
	192	<b>0.068</b>	<b>0.203</b>	0.071	0.205	0.080	0.225	0.087	0.235	0.071	<b>0.204</b>	0.104	0.245	0.114	0.262
	336	<b>0.080</b>	<b>0.225</b>	0.081	<b>0.225</b>	<b>0.075</b>	<b>0.215</b>	0.089	0.237	0.098	0.244	0.119	0.270	0.107	0.258
	720	<b>0.079</b>	<b>0.225</b>	0.087	<b>0.232</b>	<b>0.079</b>	<b>0.225</b>	0.176	0.343	0.189	0.359	0.142	0.299	0.126	0.283
	Avg	<b>0.069</b>	<b>0.207</b>	0.073	0.210	0.074	0.215	0.102	0.251	0.103	0.246	0.111	0.257	0.104	0.252
ETTth2	96	<b>0.128</b>	<b>0.277</b>	0.129	0.282	0.151	0.310	0.128	0.271	0.131	0.279	<b>0.128</b>	<b>0.271</b>	0.153	0.306
	192	<b>0.168</b>	<b>0.324</b>	<b>0.168</b>	<b>0.328</b>	0.179	0.337	0.175	0.328	<b>0.176</b>	0.329	0.185	0.330	0.204	0.351
	336	<b>0.194</b>	<b>0.355</b>	<b>0.185</b>	<b>0.351</b>	0.195	0.356	0.192	0.354	0.209	0.367	0.231	0.378	0.246	0.389
	720	<b>0.222</b>	<b>0.380</b>	0.224	0.383	<b>0.195</b>	<b>0.363</b>	0.268	0.418	0.276	0.426	0.278	0.420	0.268	0.409
	Avg	<b>0.178</b>	<b>0.334</b>	<b>0.176</b>	<b>0.336</b>	0.180	0.341	0.190	0.342	0.198	0.350	0.205	0.349	0.217	0.363
weather	96	<b>0.0010</b>	<b>0.0237</b>	0.0012	0.0256	0.0012	0.0257	0.0059	0.0645	0.0055	0.0617	0.0042	0.0533	0.0034	0.0467
	192	<b>0.0012</b>	0.0269	0.0013	<b>0.0268</b>	0.0013	0.0281	0.0063	0.0674	0.0061	0.0659	0.0067	0.0669	0.0039	0.0482
	336	<b>0.0014</b>	0.0284	<b>0.0014</b>	<b>0.0283</b>	0.0015	0.0298	0.0075	0.0748	0.0064	0.0678	0.0024	0.0394	0.0077	0.0633
	720	<b>0.0019</b>	0.0333	<b>0.0019</b>	<b>0.0324</b>	0.0020	0.0345	0.0059	0.0634	0.0068	0.0706	0.0038	0.0510	0.0103	0.0743
	Avg	<b>0.0013</b>	<b>0.0280</b>	0.0014	0.0282	0.0015	0.0295	0.0064	0.0675	0.0062	0.0665	0.0042	0.0526	0.0063	0.0581
exchange	96	<b>0.098</b>	<b>0.232</b>	0.162	0.314	0.156	0.293	0.150	0.328	<b>0.111</b>	<b>0.262</b>	0.369	0.475	0.418	0.516
	192	<b>0.210</b>	<b>0.341</b>	0.273	0.411	0.264	<b>0.375</b>	<b>0.211</b>	0.377	0.229	0.391	0.511	0.551	0.538	0.591
	336	<b>0.412</b>	<b>0.480</b>	0.479	0.538	0.501	0.536	0.440	0.524	<b>0.427</b>	<b>0.514</b>	0.713	0.635	0.978	0.760
	720	<b>1.218</b>	<b>0.836</b>	<b>0.912</b>	<b>0.770</b>	1.411	0.937	1.337	0.943	1.500	1.009	1.307	0.888	1.223	0.857
	Avg	<b>0.484</b>	<b>0.472</b>	<b>0.456</b>	<b>0.508</b>	0.583	0.535	0.534	0.543	0.566	0.544	0.725	0.637	0.789	0.681
ECL	96	<b>0.202</b>	<b>0.314</b>	0.247	0.348	0.227	0.338	0.244	0.355	<b>0.208</b>	<b>0.321</b>	0.396	0.471	0.482	0.519
	192	<b>0.239</b>	<b>0.341</b>	0.315	0.389	0.286	0.382	0.295	0.395	<b>0.238</b>	<b>0.342</b>	0.376	0.460	0.574	0.568
	336	<b>0.276</b>	<b>0.369</b>	0.490	0.494	0.326	0.401	0.333	0.428	<b>0.273</b>	<b>0.369</b>	0.522	0.543	0.587	0.581
	720	<b>0.317</b>	<b>0.413</b>	0.550	0.540	<b>0.392</b>	<b>0.455</b>	0.397	0.472	<b>0.310</b>	<b>0.411</b>	0.531	0.555	0.564	0.567
	Avg	<b>0.258</b>	<b>0.359</b>	0.400	0.442	0.307	0.394	0.317	0.412	<b>0.257</b>	<b>0.360</b>	0.456	0.507	0.551	0.558
traffic	96	<b>0.126</b>	<b>0.207</b>	0.134	0.211	0.144	0.232	0.150	0.249	0.139	0.230	0.250	0.355	0.280	0.388
	192	<b>0.131</b>	<b>0.214</b>	0.139	0.219	0.142	0.228	0.146	0.225	0.140	0.232	0.216	0.325	0.292	0.398
	336	<b>0.129</b>	<b>0.217</b>	0.138	0.221	0.139	0.229	0.152	0.239	0.142	0.236	0.331	0.428	0.241	0.354
	720	<b>0.143</b>	<b>0.233</b>	0.154	0.244	0.153	0.250	0.161	0.249	0.157	0.254	0.414	0.487	0.240	0.341
	Avg	<b>0.132</b>	<b>0.217</b>	0.141	0.223	0.144	0.234	0.152	0.240	0.144	0.238	0.302	0.398	0.263	0.370
ILI	24	<b>0.619</b>	<b>0.557</b>	0.845	0.651	0.698	0.661	2.356	1.209	<b>0.742</b>	0.661	1.024	0.880	0.928	0.834
	36	<b>0.677</b>	<b>0.602</b>	0.698	0.624	0.708	0.680	0.646	0.616	<b>0.553</b>	<b>0.613</b>	1.009	0.897	0.942	0.838
	48	<b>0.683</b>	<b>0.633</b>	0.806	0.711	0.792	0.740	0.955	0.859	<b>0.700</b>	<b>0.706</b>	1.014	0.886	1.032	0.881
	60	<b>0.683</b>	<b>0.663</b>	0.827	0.753	0.911	0.814	1.162	0.970	0.863	0.801	1.384	1.025	1.657	1.171
	Avg	<b>0.665</b>	<b>0.613</b>	0.794	0.684	0.777	0.723	1.279	0.913	<b>0.714</b>	0.695	1.107	0.922	1.139	0.931
1st count	<b>67</b>		15		8		0		9		2		0		

\* We replace the input length L=336 in TimesNet and MICN for a fair comparison. Other experimental results are taken from the PatchTST.



Figure 8: Visualization of prediction on ETTm2 with the look-back window  $L=336$ .

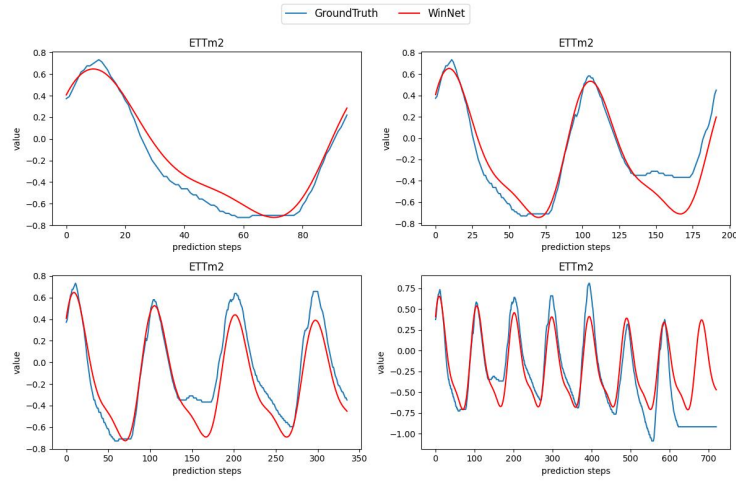


Figure 9: Visualization of prediction on ETTh1 with the look-back window  $L=336$ .

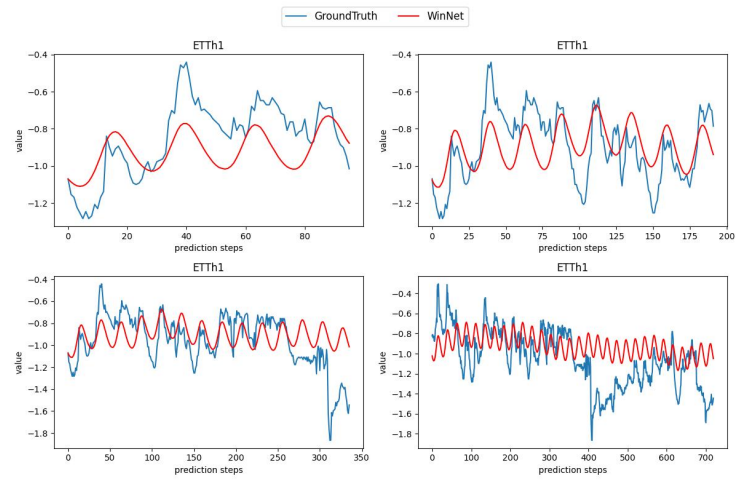


Figure 10: Visualization of prediction on ECL with the look-back window  $L=336$ .

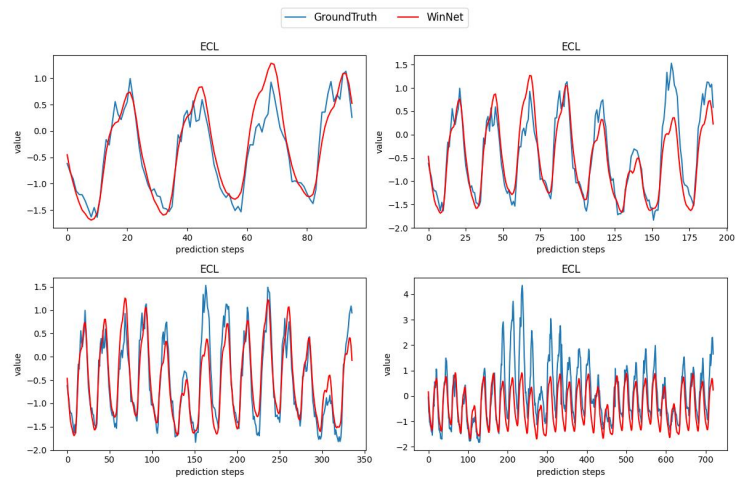


Figure 11: Visualization of prediction on ETTh2 with the look-back window  $L=336$ .

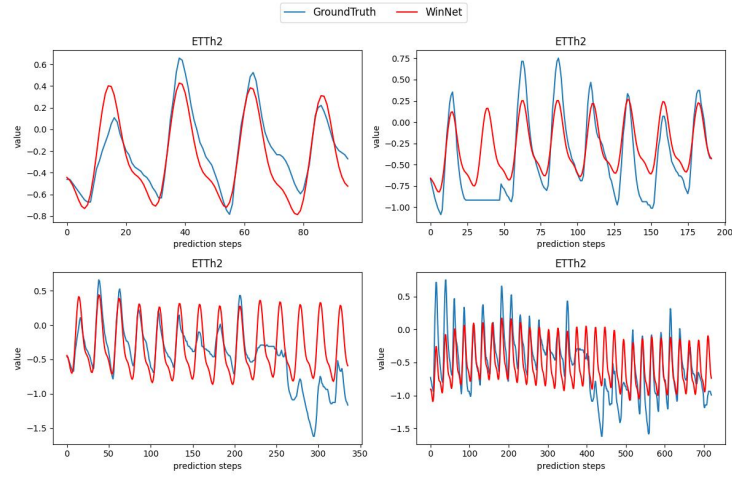


Figure 12: Visualization of prediction on ETTm1 with the look-back window  $L=336$ .

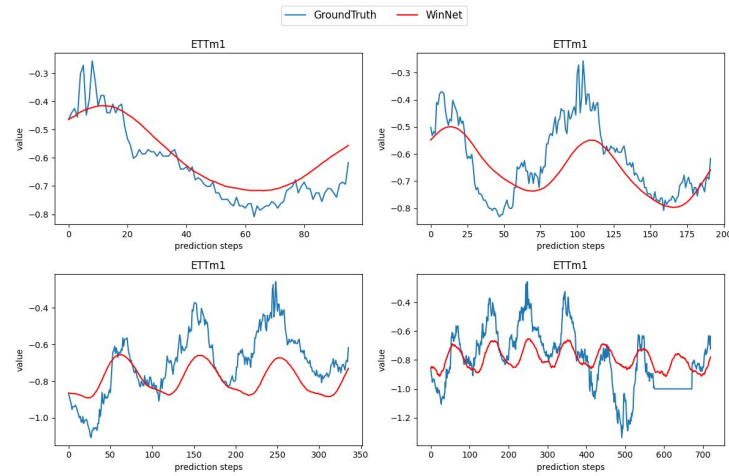


Figure 13: Visualization of prediction on ILI with the look-back window  $L=104$ .

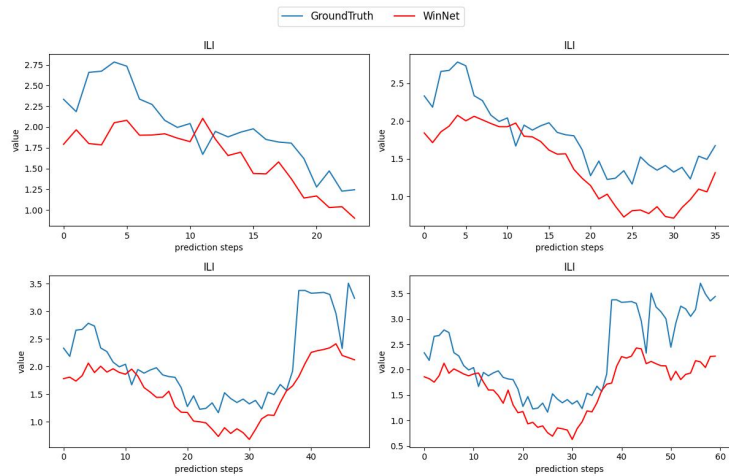


Table 11: Period frequencies obtained by FFT for each dataset before MLP. The 1st period represents the period with the highest number of occurrences, and so on. Period represents the specific value of periods and count represents the number of occurrences. It can be seen that most datasets have large periods, difficult to capture the periodicity through CNNs.

topk periods	1st period		2nd period		3rd period		4th period	
	period	count	period	count	period	count	period	count
weather	168	759	336	759	112	759	67	759
ETTm1	112	711	84	711	336	711	168	711
ECL	168	374	336	294	84	235	112	234
ETTm2	336	711	112	711	84	711	168	711
ETTh1	24	171	336	171	12	171	168	171
traffic	24	246	12	246	8	246	168	246
exchange	336	101	168	101	112	101	84	101
ETTh2	336	171	24	171	168	171	112	171

Figure 14: Period Window. The samples with same colour in each row represent time steps in a same period. A sequence can obtain its top-k period values by FFT. The 1st period indicates the most dominant period in the sequence, and so on. For example, the top-3 periods are  $\{6, 4, 3\}$  and approximated multiples of these periods are selected as the size of the periodic window, i.e. 12.

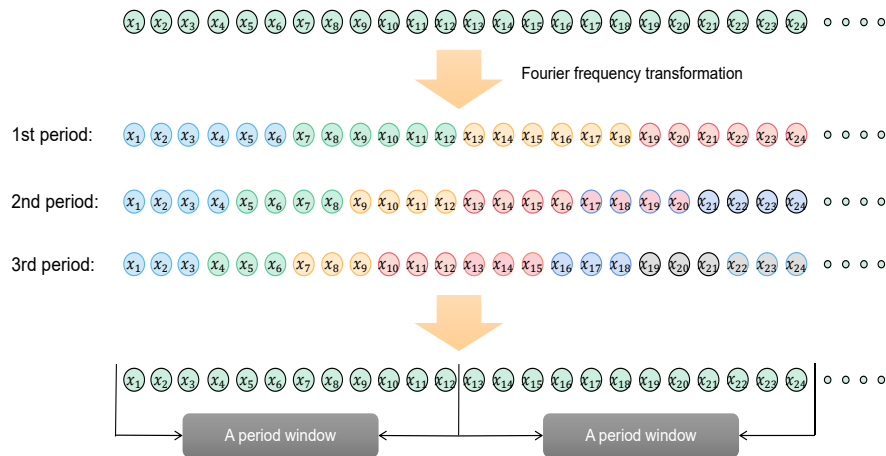




Table 12: Ablation study with different sizes of the periodic window.

Methods		WinNet						TimesNet*		DLinear	
		18		24		32					
Metric		MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
ETTm1	96	0.282	0.336	0.283	<b>0.335</b>	<b>0.279</b>	0.334	0.377	0.397	0.290	0.342
	192	<b>0.319</b>	<b>0.357</b>	0.324	0.360	0.323	0.361	0.389	0.401	0.332	0.369
	336	<b>0.354</b>	<b>0.379</b>	0.357	<b>0.379</b>	<b>0.354</b>	0.382	0.393	0.414	0.366	0.392
	720	0.419	0.412	0.416	<b>0.411</b>	<b>0.408</b>	0.414	0.470	0.458	0.420	0.424
ETTm2	96	0.163	0.252	<b>0.160</b>	<b>0.251</b>	0.162	0.251	0.201	0.280	0.167	0.260
	192	0.216	0.290	<b>0.212</b>	<b>0.287</b>	0.216	0.291	0.242	0.313	0.224	0.303
	336	0.271	0.325	<b>0.261</b>	<b>0.322</b>	0.268	0.323	0.310	0.356	0.281	0.342
	720	0.363	0.385	<b>0.359</b>	<b>0.381</b>	0.360	0.386	0.381	0.396	0.397	0.421
ETTh1	96	0.367	0.392	<b>0.362</b>	<b>0.390</b>	0.374	0.397	0.460	0.464	0.375	0.399
	192	0.402	0.413	<b>0.394</b>	<b>0.410</b>	0.409	0.420	0.458	0.462	0.412	0.420
	336	0.426	0.428	<b>0.419</b>	<b>0.426</b>	0.427	0.430	0.523	0.501	0.439	0.443
	720	0.442	0.454	0.436	0.453	<b>0.434</b>	<b>0.451</b>	0.502	0.497	0.472	0.490
ETTh2	96	0.271	0.333	<b>0.267</b>	<b>0.332</b>	0.275	0.337	0.338	0.397	0.289	0.353
	192	0.326	0.374	<b>0.322</b>	<b>0.372</b>	0.337	0.377	0.422	0.446	0.383	0.418
	336	0.356	<b>0.400</b>	<b>0.351</b>	0.401	0.368	0.414	0.431	0.460	0.448	0.465
	720	0.395	0.437	<b>0.389</b>	<b>0.436</b>	0.406	0.448	0.467	0.480	0.605	0.551
weather	96	0.151	0.207	<b>0.143</b>	<b>0.198</b>	0.146	0.202	0.163	0.223	0.176	0.237
	192	0.196	0.253	<b>0.188</b>	<b>0.240</b>	0.191	0.245	0.218	0.266	0.192	0.246
	336	0.245	0.288	<b>0.235</b>	<b>0.280</b>	0.239	0.289	0.280	0.306	0.240	0.287
	720	0.319	0.341	<b>0.310</b>	<b>0.336</b>	0.315	0.345	0.349	0.356	0.316	0.352
ECL	96	0.141	0.238	<b>0.130</b>	<b>0.226</b>	0.142	0.240	0.181	0.281	0.140	0.237
	192	0.155	0.251	<b>0.147</b>	<b>0.240</b>	0.157	0.253	0.193	0.293	0.153	0.249
	336	0.172	0.268	<b>0.163</b>	<b>0.257</b>	0.174	0.270	0.205	0.312	0.169	0.267
	720	0.211	0.300	<b>0.198</b>	<b>0.290</b>	0.212	0.301	0.222	0.320	0.203	0.301
ILI	24	1.987	0.906	<b>1.985</b>	<b>0.905</b>	2.031	0.934	2.500	1.055	2.215	1.081
	36	1.928	0.915	1.897	<b>0.900</b>	<b>1.889</b>	0.902	2.222	1.007	1.963	0.963
	48	1.904	0.919	<b>1.868</b>	<b>0.910</b>	1.902	0.922	2.304	1.043	2.130	1.024
	60	2.066	0.976	<b>1.928</b>	<b>0.933</b>	2.012	0.958	2.354	1.046	2.368	1.096

\* We replace the input length L=512 in TimesNet and MICN for a fair comparison.

Table 13: Prediction error (MSE &amp; MAE) with different kernel size on 3 datasets: weather, ECL and traffic.

Methods		WinNet						TimesNet*		DLinear	
		3x3		5x5		7x7					
Metric		MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
weather	96	0.143	<b>0.198</b>	<b>0.141</b>	0.239	0.150	0.203	0.163	0.223	0.176	0.237
	192	<b>0.188</b>	<b>0.240</b>	0.193	0.245	0.193	0.244	0.218	0.266	0.192	0.246
	336	<b>0.235</b>	<b>0.280</b>	0.243	0.285	0.244	0.287	0.280	0.306	0.240	0.287
	720	<b>0.310</b>	<b>0.336</b>	0.323	0.346	0.318	0.340	0.349	0.356	0.316	0.352
ECL	96	<b>0.130</b>	<b>0.226</b>	0.141	0.239	0.140	0.238	0.181	0.281	0.140	0.237
	192	<b>0.147</b>	<b>0.240</b>	0.161	0.258	0.161	0.258	0.193	0.293	0.153	0.249
	336	<b>0.163</b>	<b>0.257</b>	0.178	0.274	0.178	0.274	0.205	0.312	0.169	0.267
	720	<b>0.198</b>	<b>0.290</b>	0.212	0.301	0.216	0.306	0.222	0.320	0.203	0.301
traffic	96	<b>0.394</b>	<b>0.274</b>	0.414	0.287	0.415	0.289	0.603	0.328	0.410	0.282
	192	<b>0.407</b>	<b>0.279</b>	0.426	0.294	0.428	0.295	0.610	0.329	0.423	0.287
	336	<b>0.416</b>	<b>0.283</b>	0.436	0.295	0.439	0.297	0.619	0.330	0.436	0.296
	720	<b>0.453</b>	<b>0.305</b>	0.464	0.310	0.467	0.313	0.632	0.352	0.466	0.315

\* We replace the input length L=512 in TimesNet for a fair comparison.

Table 14: Ablation study the fusion mode of inter-period and intra-period on all datasets in WinNet. NF means the normal fusion, while DF means no transposition, direct fusion.

Methods		WinNet				TimeNet*		DLinear	
		DF		NF					
Metric		MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
ETTm1	96	<b>0.283</b>	<b>0.335</b>	0.286	0.337	0.377	0.397	0.290	0.342
	192	<b>0.324</b>	<b>0.360</b>	0.326	0.360	0.389	0.401	0.332	0.369
	336	<b>0.357</b>	<b>0.379</b>	0.366	0.381	0.393	0.414	0.366	0.392
	720	<b>0.416</b>	<b>0.411</b>	0.421	0.414	0.470	0.458	0.420	0.424
ETTm2	96	<b>0.160</b>	<b>0.251</b>	0.162	0.252	0.201	0.280	0.167	0.260
	192	<b>0.212</b>	<b>0.287</b>	0.217	0.290	0.242	0.313	0.224	0.303
	336	<b>0.261</b>	<b>0.322</b>	0.273	0.326	0.310	0.356	0.281	0.342
	720	<b>0.359</b>	<b>0.381</b>	0.363	0.384	0.381	0.396	0.397	0.421
ETTh1	96	<b>0.362</b>	<b>0.390</b>	0.369	0.393	0.460	0.464	0.375	0.399
	192	<b>0.394</b>	<b>0.410</b>	0.409	0.419	0.458	0.462	0.412	0.420
	336	<b>0.419</b>	<b>0.426</b>	0.432	0.432	0.523	0.501	0.439	0.443
	720	<b>0.436</b>	<b>0.453</b>	0.438	0.456	0.502	0.497	0.472	0.490
ETTh2	96	<b>0.267</b>	<b>0.332</b>	0.270	0.335	0.338	0.397	0.289	0.353
	192	<b>0.322</b>	<b>0.372</b>	0.330	0.376	0.422	0.446	0.383	0.418
	336	<b>0.351</b>	<b>0.401</b>	0.385	0.413	0.431	0.460	0.448	0.465
	720	<b>0.389</b>	<b>0.436</b>	0.424	0.464	0.467	0.480	0.605	0.551
weather	96	<b>0.143</b>	<b>0.198</b>	0.154	0.206	0.163	0.223	0.176	0.237
	192	<b>0.188</b>	<b>0.240</b>	0.193	0.246	0.218	0.266	0.192	0.246
	336	<b>0.235</b>	<b>0.280</b>	0.244	0.285	0.280	0.306	0.240	0.287
	720	<b>0.310</b>	<b>0.336</b>	0.321	0.342	0.349	0.356	0.316	0.352
ECL	96	<b>0.130</b>	<b>0.226</b>	0.141	0.239	0.181	0.281	0.140	0.237
	192	<b>0.147</b>	<b>0.240</b>	0.156	0.252	0.193	0.293	0.153	0.249
	336	<b>0.163</b>	<b>0.257</b>	0.173	0.269	0.205	0.312	0.169	0.267
	720	<b>0.198</b>	<b>0.290</b>	0.211	0.300	0.222	0.320	0.203	0.301
traffic	96	<b>0.394</b>	<b>0.274</b>	0.414	0.288	0.603	0.328	0.410	0.282
	192	<b>0.407</b>	<b>0.279</b>	0.427	0.294	0.610	0.329	0.423	0.287
	336	<b>0.416</b>	<b>0.283</b>	0.435	0.296	0.619	0.330	0.436	0.296
	720	<b>0.453</b>	<b>0.305</b>	0.465	0.312	0.632	0.352	0.466	0.315
ILI	96	<b>1.985</b>	<b>0.905</b>	2.003	0.923	2.500	1.055	2.215	1.081
	192	<b>1.897</b>	<b>0.900</b>	1.906	0.909	2.222	1.007	1.963	0.963
	336	<b>1.868</b>	<b>0.910</b>	1.925	0.930	2.304	1.043	2.130	1.024
	720	<b>1.928</b>	<b>0.933</b>	1.937	0.934	2.354	1.046	2.368	1.096

\* We replace the input length L=512 in TimesNet for a fair comparison.

Table 15: Prediction error (MSE &amp; MAE) with different kernel sizes on 5 datasets: ETTm1, ETTm2, ETTh1, ETTh2 and ILI.

Methods		WinNet						TimesNet*		DLinear	
		3x3		5x5		7x7					
Metric		MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
ETTm1	96	<b>0.283</b>	<b>0.335</b>	0.288	0.339	0.286	0.338	0.377	0.397	0.290	0.342
	192	<b>0.324</b>	<b>0.360</b>	0.335	0.368	0.329	0.364	0.389	0.401	0.332	0.369
	336	<b>0.357</b>	<b>0.379</b>	0.370	0.385	0.366	0.380	0.393	0.414	0.366	0.392
	720	<b>0.416</b>	<b>0.411</b>	0.423	0.414	0.422	0.413	0.470	0.458	0.420	0.424
ETTm2	96	<b>0.160</b>	<b>0.251</b>	0.163	0.252	0.162	0.252	0.201	0.280	0.167	0.260
	192	<b>0.212</b>	<b>0.287</b>	0.216	0.289	0.217	0.290	0.242	0.313	0.224	0.303
	336	<b>0.261</b>	0.322	0.267	<b>0.321</b>	0.269	0.324	0.310	0.356	0.281	0.342
	720	<b>0.359</b>	<b>0.381</b>	0.361	0.382	0.375	0.387	0.381	0.396	0.397	0.421
ETTh1	96	<b>0.362</b>	<b>0.390</b>	0.368	0.392	0.366	0.392	0.460	0.464	0.375	0.399
	192	<b>0.394</b>	<b>0.410</b>	0.410	0.420	0.425	0.434	0.458	0.462	0.412	0.420
	336	<b>0.419</b>	<b>0.426</b>	0.436	0.434	0.442	0.442	0.523	0.501	0.439	0.443
	720	0.436	0.453	<b>0.435</b>	<b>0.453</b>	0.436	<b>0.453</b>	0.502	0.497	0.472	0.490
ETTh2	96	<b>0.267</b>	<b>0.332</b>	0.274	0.337	0.270	0.334	0.338	0.397	0.289	0.353
	192	<b>0.322</b>	<b>0.372</b>	0.337	0.378	0.360	0.399	0.422	0.446	0.383	0.418
	336	<b>0.351</b>	<b>0.401</b>	0.376	0.409	0.364	0.409	0.431	0.460	0.448	0.465
	720	<b>0.389</b>	<b>0.436</b>	0.502	0.532	0.418	0.458	0.467	0.480	0.605	0.551
ILI	24	<b>1.985</b>	<b>0.905</b>	2.034	0.923	1.966	0.900	2.500	1.055	2.215	1.081
	36	<b>1.897</b>	<b>0.900</b>	1.906	0.909	1.900	0.902	2.222	1.007	1.963	0.963
	48	<b>1.868</b>	<b>0.910</b>	1.925	0.930	1.906	0.922	2.304	1.043	2.130	1.024
	60	<b>1.928</b>	<b>0.933</b>	1.937	0.934	1.945	0.937	2.354	1.046	2.368	1.096

\* We replace the input length L=512 in TimesNet for a fair comparison.

Table 16: Ablation study on all ETT datasets of the proposed modules including the I2PE, TDPD and DCB in WinNet. Four cases are included: (a) all the three modules are included in model (Final: I2PE+TDPD+DCB); (b) only the TDPD; (c) TDPD+DCB; (d) the original version with the common CNN and one-dimensional trend decomposition.

Methods		WinNet						TimeNet*		DLinear	
		Final		TDPD+DCB		TDPD					
Metric		MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
ETTm1	96	<b>0.283</b>	<b>0.335</b>	0.288	0.341	0.289	0.341	0.292	0.352	0.377	0.397
	192	<b>0.324</b>	<b>0.360</b>	0.328	0.363	0.330	0.365	0.327	0.374	0.389	0.401
	336	<b>0.357</b>	<b>0.379</b>	0.362	0.385	0.363	0.384	0.357	0.391	0.393	0.414
	720	0.416	<b>0.411</b>	<b>0.409</b>	0.413	0.428	0.421	<b>0.409</b>	0.423	0.470	0.458
ETTm2	96	0.160	0.251	<b>0.159</b>	<b>0.247</b>	0.161	0.249	0.359	0.390	0.201	0.280
	192	<b>0.212</b>	<b>0.287</b>	0.217	0.288	0.217	0.290	0.611	0.511	0.242	0.313
	336	<b>0.261</b>	<b>0.322</b>	0.270	0.326	0.268	0.324	0.855	0.613	0.310	0.356
	720	0.359	0.381	<b>0.358</b>	<b>0.380</b>	0.359	0.383	1.002	0.674	0.381	0.396
ETTh1	96	<b>0.362</b>	<b>0.390</b>	0.363	0.393	0.377	0.404	0.379	0.415	0.460	0.464
	192	<b>0.394</b>	<b>0.410</b>	0.397	0.415	0.402	0.418	0.411	0.438	0.458	0.462
	336	<b>0.419</b>	<b>0.426</b>	0.422	0.431	0.448	0.450	0.440	0.461	0.523	0.501
	720	<b>0.436</b>	<b>0.453</b>	0.437	0.456	0.449	0.461	0.506	0.525	0.502	0.497
ETTh2	96	0.267	<b>0.332</b>	<b>0.264</b>	<b>0.332</b>	0.269	0.333	0.842	0.622	0.338	0.397
	192	0.322	<b>0.372</b>	<b>0.319</b>	<b>0.372</b>	0.330	0.375	0.971	0.682	0.422	0.446
	336	0.351	<b>0.401</b>	<b>0.349</b>	<b>0.401</b>	0.382	0.416	1.098	0.728	0.431	0.460
	720	<b>0.389</b>	<b>0.436</b>	0.391	0.437	0.398	0.443	1.428	0.854	0.467	0.480

\* We replace the input length L=512 in TimesNet for a fair comparison.

Table 17: Efficiency of our model on the ECL dataset vs other methods in multivariate prediction. We set the input length to 720 and the prediction length to 720. See relevant computational efficiency with thop, torchsummary and memory\_allocated functions.

Method	FLOPs	Parameter	Time	Memory
<b>WinNet</b>	<b>273M</b>	<b>836.8K</b>	0.42s	<b>12MiB</b>
PatchTST	51.1G	18.18M	0.17s	85MiB
TimesNet	1620.1G	226.3M	0.70s	878MiB
MICN	5.95G	19.07M	0.04s	88MiB
Crossformer	146.4G	11.1M	0.54s	96MiB
DLinear	<b>333M</b>	<b>1.04M</b>	<b>0.01s</b>	<b>14MiB</b>
FEDformer	2.35G	4.76M	0.56s	39MiB
Autoformer	2.35G	3.19M	0.26s	32MiB
Informer	1.84G	3.35M	0.14s	33MiB
Transformer	2.17G	2.95M	<b>0.03s</b>	31MiB

Table 18: Efficiency of our model on the ETTm1 dataset vs. other methods in multivariate prediction. We set the input length to 720 and the prediction length to 720. See relevant computational efficiency with thop, torchsummary and memory\_allocated functions.

Method	FLOPs	Parameter	Time	Memory
<b>WinNet</b>	<b>5.96M</b>	<b>830.9K</b>	28ms	<b>11MiB</b>
PatchTST	309M	8.69M	26ms	44MiB
TimesNet	406.2G	57.28M	500ms	239MiB
MICN	5.33G	18.75M	<b>25ms</b>	85MiB
Crossformer	3.46G	11.09M	62ms	57MiB
DLinear	<b>7.26M</b>	<b>1.04M</b>	<b>15ms</b>	<b>12MiB</b>
FEDformer	1.75G	3.96M	450ms	24MiB
Autoformer	1.75G	2.39M	270ms	27MiB
Informer	1.42G	2.78M	150ms	29MiB
Transformer	1.75G	2.39M	45ms	27MiB