



Ada-MSHyper: Adaptive Multi-Scale Hypergraph Transformer for Time Series Forecasting

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Motivations

Two key challenges limit the performance of Transformer in multi-scale temporal pattern interaction modeling.

- **Semantic information sparsity.** Individual time points contain less semantic information, and **pair-wise interactions** may cause the information utilization bottleneck.
- **Temporal variations entanglement.** Multiple inherent temporal variations (e.g., rising, falling, and fluctuating) entangled in temporal pattern, bringing challenges for time series forecasting.

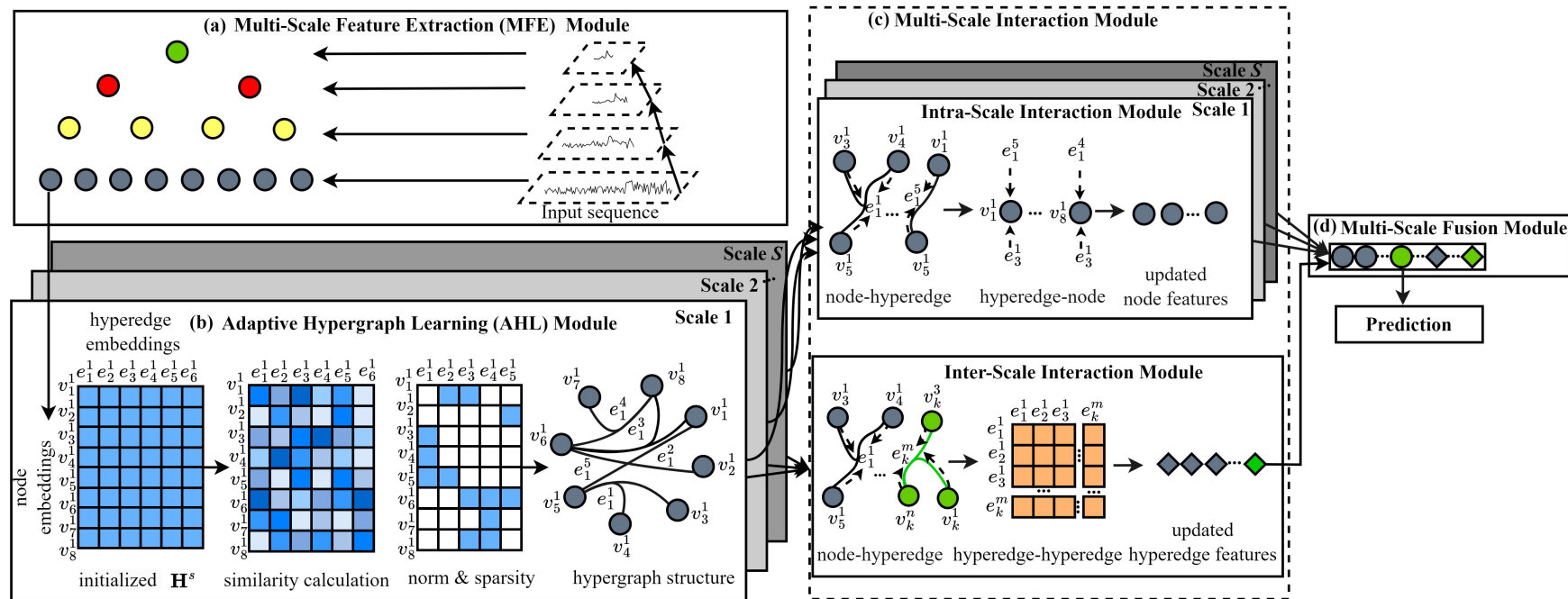
Contributions

Ada-MSHyper is **the first work** that incorporates adaptive hypergraph modeling into time series forecasting.

- An adaptive hypergraph learning module is designed to model abundant and implicit **group-wise node interactions** at different scales.
- A node and hyperedge constraint mechanism is introduced to cluster nodes with **similar semantic information** and **differentiate the temporal variations** within each scales.
- Experimental results on 11 real-world datasets demonstrate that Ada-MSHyper achieves **state-of-the-art** performance.

Method: Ada-MSHyper

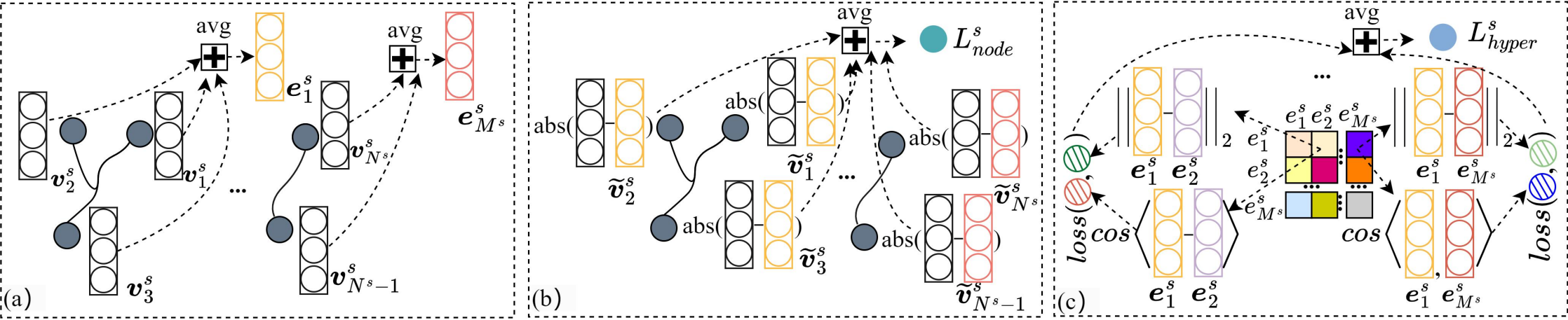
Framework



(a) The MFE module maps the input sequence into subsequences at different scales. (b) The AHL module provides foundations for modeling group-wise interactions. (c) The multi-scale interaction module models group-wise pattern interactions at different scales.

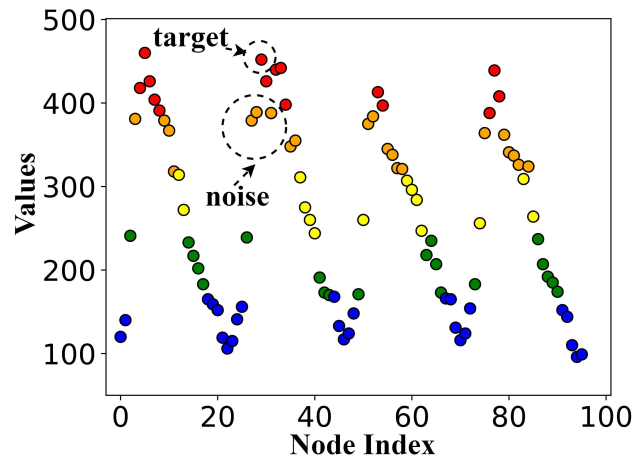
Method: Ada-MSHyper

Node and hyperedge constraint (NHC) mechanism

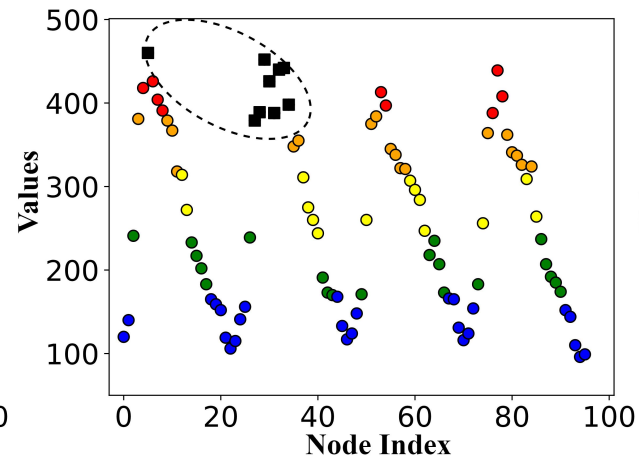


(a) The generation of hyperedge features. (b) The generation of node loss. (c) The generation of hyperedge loss.

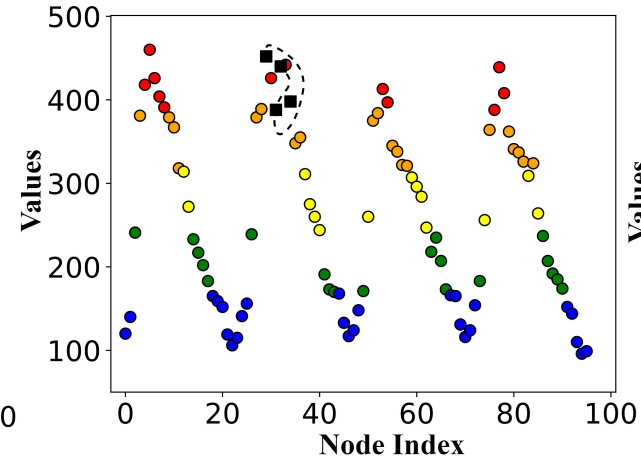
Visualization



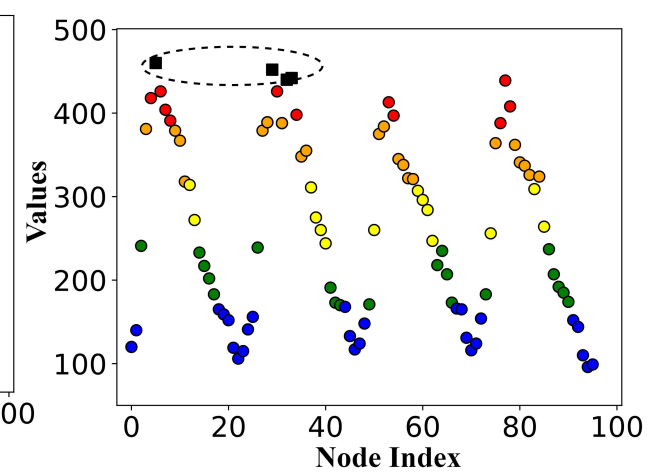
(a) Input sequence



(b) -w/o NHC



(c) -w/o NC



(d) Ada-MSHyper

With the NHC mechanism, Ada-MSHyper can not only cluster nodes with similar semantic information but also reduce noise interference.

Evaluations

The results of long-range time series forecasting under multivariate settings.

The results of short-range time series forecasting under multivariate settings.

Models	Ada-MSHyper (Ours)		iTransformer* (2024)		MSHyper* (2024)		TimeMixer* (2024)		MSGNet* (2024)		CrossGNN* (2023)		PatchTST (2023)		Crossformer (2023)		TimesNet (2023)		DLinear (2023)		FiLM* (2022)		FEDformer (2022)		Autoformer (2021)		
	Metric	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
Weather	96	0.157	0.195	0.174	0.214	0.174	0.223	0.163	0.210	0.163	0.212	0.159	0.218	0.177	0.218	0.158	0.230	0.172	0.220	0.196	0.255	0.199	0.262	0.217	0.296	0.266	0.336
	192	0.218	0.259	0.221	0.254	0.218	0.253	0.212	0.257	0.212	0.254	0.211	0.266	0.225	0.259	0.206	0.277	0.219	0.261	0.237	0.296	0.228	0.288	0.276	0.336	0.307	0.367
	336	0.251	0.252	0.278	0.296	0.269	0.300	0.269	0.292	0.272	0.299	0.267	0.310	0.278	0.297	0.272	0.335	0.280	0.306	0.283	0.335	0.267	0.323	0.339	0.380	0.359	0.395
720	0.304	0.328	0.358	0.347	0.343	0.341	0.343	0.345	0.350	0.348	0.352	0.362	0.354	0.348	0.398	0.418	0.365	0.359	0.345	0.381	0.319	0.361	0.403	0.428	0.419	0.428	
Electricity	96	0.135	0.238	0.148	0.240	0.176	0.261	0.153	0.247	0.165	0.274	0.173	0.275	0.181	0.270	0.219	0.314	0.168	0.272	0.197	0.282	0.198	0.274	0.193	0.308	0.201	0.317
	192	0.152	0.239	0.162	0.253	0.173	0.260	0.166	0.256	0.184	0.292	0.195	0.288	0.188	0.274	0.231	0.322	0.184	0.289	0.196	0.285	0.198	0.278	0.201	0.315	0.222	0.334
	336	0.168	0.266	0.178	0.269	0.195	0.297	0.185	0.277	0.195	0.302	0.206	0.300	0.204	0.293	0.246	0.337	0.198	0.300	0.209	0.301	0.217	0.300	0.214	0.329	0.231	0.338
720	0.212	0.293	0.225	0.317	0.219	0.315	0.225	0.310	0.231	0.332	0.231	0.335	0.246	0.324	0.280	0.363	0.220	0.320	0.245	0.333	0.278	0.356	0.246	0.355	0.254	0.361	
ETTh1	96	0.372	0.393	0.386	0.405	0.392	0.407	0.385	0.402	0.390	0.411	0.382	0.398	0.414	0.419	0.423	0.448	0.384	0.402	0.386	0.400	0.438	0.433	0.376	0.419	0.449	0.459
	192	0.433	0.417	0.441	0.436	0.440	0.426	0.443	0.430	0.442	0.442	0.427	0.425	0.460	0.445	0.471	0.474	0.436	0.429	0.437	0.432	0.493	0.466	0.420	0.448	0.500	0.482
	336	0.422	0.433	0.487	0.458	0.480	0.453	0.434	0.443	0.480	0.468	0.465	0.445	0.501	0.466	0.570	0.546	0.491	0.469	0.481	0.459	0.547	0.495	0.459	0.465	0.521	0.496
720	0.445	0.459	0.503	0.491	0.508	0.493	0.498	0.476	0.494	0.488	0.472	0.468	0.500	0.488	0.653	0.621	0.521	0.500	0.519	0.516	0.586	0.538	0.506	0.507	0.514	0.512	
ETTh2	96	0.283	0.332	0.297	0.349	0.300	0.351	0.296	0.347	0.328	0.371	0.309	0.359	0.302	0.348	0.475	0.584	0.340	0.374	0.333	0.387	0.322	0.364	0.358	0.397	0.346	0.388
	192	0.258	0.374	0.380	0.400	0.384	0.400	0.376	0.394	0.402	0.414	0.390	0.406	0.388	0.400	0.877	0.656	0.402	0.414	0.477	0.476	0.404	0.414	0.429	0.439	0.456	0.452
	336	0.282	0.437	0.428	0.432	0.434	0.438	0.434	0.443	0.435	0.443	0.426	0.444	0.426	0.433	1.043	0.731	0.452	0.452	0.594	0.541	0.435	0.445	0.496	0.487	0.482	0.486
720	0.413	0.432	0.427	0.445	0.412	0.441	0.464	0.464	0.417	0.441	0.445	0.464	0.441	0.446	1.104	0.763	0.462	0.468	0.831	0.657	0.447	0.458	0.463	0.474	0.515	0.511	
ETTh1	96	0.341	0.354	0.334	0.368	0.319	0.369	0.318	0.366	0.319	0.366	0.335	0.373	0.329	0.367	0.404	0.426	0.338	0.375	0.345	0.372	0.353	0.370	0.379	0.419	0.505	0.475
	192	0.345	0.375	0.377	0.391	0.392	0.391	0.366	0.385	0.376	0.397	0.372	0.390	0.367	0.385	0.550	0.451	0.374	0.387	0.380	0.389	0.389	0.387	0.426	0.441	0.553	0.496
	336	0.375	0.397	0.426	0.420	0.426	0.410	0.396	0.404	0.417	0.422	0.403	0.411	0.399	0.410	0.399	0.410	0.411	0.413	0.413	0.421	0.408	0.445	0.459	0.621	0.537	
720	0.437	0.435	0.491	0.459	0.483	0.448	0.454	0.441	0.481	0.458	0.461	0.442	0.454	0.439	0.666	0.589	0.478	0.450	0.474	0.453	0.481	0.441	0.543	0.490	0.671	0.561	
ETTh2	96	0.165	0.257	0.180	0.264	0.183	0.267	0.175	0.258	0.177	0.262	0.176	0.266	0.175	0.259	0.287	0.366	0.187	0.267	0.193	0.292	0.183	0.266	0.203	0.287	0.255	0.339
	192	0.230	0.307	0.250	0.309	0.257	0.313	0.241	0.304	0.247	0.307	0.240	0.307	0.241	0.302	0.414	0.492	0.249	0.309	0.284	0.362	0.248	0.305	0.269	0.328	0.281	0.340
	336	0.282	0.328	0.311	0.348	0.335	0.361	0.303	0.343	0.312	0.346	0.304	0.345	0.305	0.343	0.597	0.542	0.321	0.351	0.369	0.427	0.309	0.345	0.325	0.366	0.339	0.372
720	0.375	0.396	0.412	0.407	0.410	0.402	0.391	0.394	0.414	0.403	0.406	0.400	0.402	0.400	1.730	1.042	0.408	0.403	0.554	0.522	0.410	0.400	0.421	0.415	0.433	0.432	
Traffic	96	0.384	0.248	0.395	0.268	0.413	0.272	0.473	0.288	0.605	0.344	0.570	0.310	0.462	0.295	0.522	0.290	0.593	0.321	0.650	0.396	0.647	0.384	0.587	0.366	0.613	0.388
	192	0.401	0.258	0.417	0.276	0.422	0.274	0.473	0.296	0.613	0.359	0.577	0.321	0.466	0.296	0.530	0.293	0.617	0.336	0.598	0.370	0.600	0.361	0.604	0.373	0.616	0.382
	336	0.423	0.261	0.433	0.283	0.438	0.292	0.508	0.312	0.642	0.376	0.588	0.324	0.482	0.304	0.558	0.305	0.629	0.336	0.605	0.373	0.610	0.367	0.621	0.383	0.622	0.337
720	0.453	0.282	0.467	0.302	0.457	0.292	0.512	0.318	0.702	0.401	0.597	0.337	0.514	0.322	0.589	0.328	0.640	0.350	0.645	0.394	0.691	0.425	0.626	0.382	0.660	0.408	

Models	Metric	Ada-MSHyper (Ours)	iTransformer* (2024)	MSHyper* (2024)	TimeMixer* (2024)	WITRAN* (2023)	PatchTST* (2023)	Dlinear* (2023)	Crossformer* (2023)	FEDformer* (2022)	Pyraformer* (2022)	Autoformer* (2021)	
		MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	MSE MAE	
ETTh1	1080	0.534	0.509	0.562	0.521	0.557	0.517	0.682	0.569	0.602	0.660	0.549	0.512
	1440	0.616	0.498	0.620	0.556	0.667	0.578	0.793	0.625	0.705	0.878	0.619	0.553
	1800	0.689	0.627	0.780	0.631	0.758	0.624	0.877	0.643	0.775	0.623	0.746	0.627
2160	0.779	0.635	1.102	0.736	0.998	0.721	1.007	0.686	0.852	1.171	0.851	0.665	
ETTh2	1080	0.426	0.461	0.486	0.488	0.464	0.469	0.483	0.480	0.432	0.474	0.453	0.468
	1440	0.465	0.437	0.512	0.507	0.524	0.506	0.547	0.510	0.472	0.443	0.513	0.501
	1800	0.503	0.505	0.565	0.529	0.522	0.496	0.606	0.544	0.517	0.503	1.327	0.840
2160	0.527	0.515	0.600	0.546	0.542	0.510	0.616	0.557	0.657	0.619	1.670	0.919	
ETTm1	1080	0.460	0.445	0.534	0.483	0.520	0.465	0.502	0.465	0.464	0.459	0.494	0.459
	1440	0.473	0.449	0.556	0.495	0.542	0.477	0.523	0.488	0.508	0.467	0.534	0.491
	1800	0.492	0.475	0.571	0.501	0.564	0.490	0.526	0.487	0.550	0.497	0.556	0.507
2160	0.510	0.483	0.555	0.499	0.550	0.487	0.542	0.491	0.569	0.481	0.507	0.481	
ETTm2	1080	0.404	0.416	0.463	0.438	0.464	0.439	0.450	0.432	0.415	0.434	0.449	0.432
	1440	0.413	0.429	0.475	0.452	0.475	0.449	0.471	0.452	0.475	0.452	0.699	0.519
	1800												



Thank you for your listening!

Presenter: Zongjiang Shang