

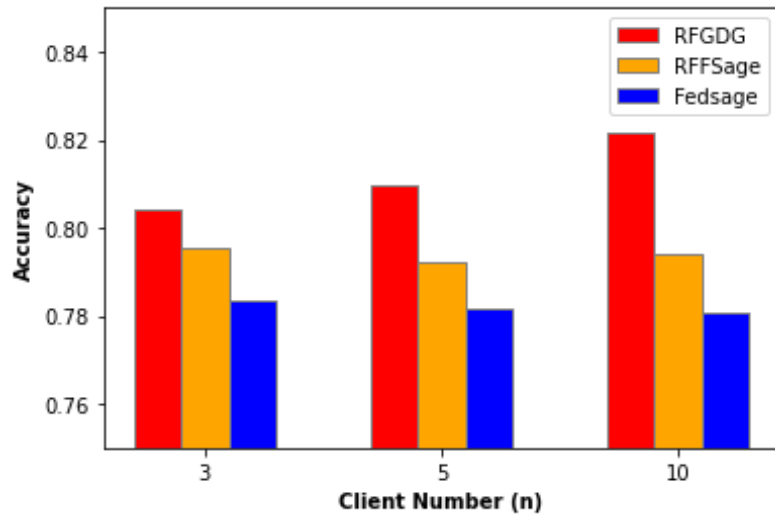
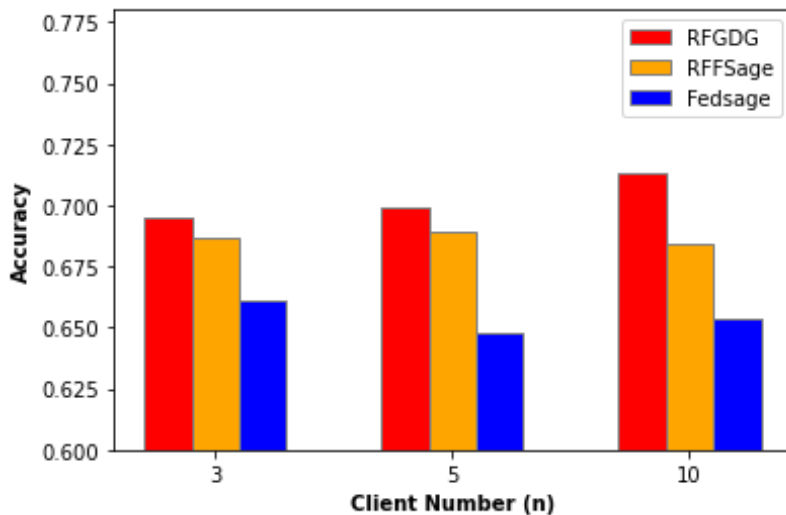
# Reinforced Federated Graph Domain Generalization with Dynamic Aggregation Strategy

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# Problems & Ideas

- Problems of federated graph learning approaches:
  - Existing federated graph methods suffer poor generalization from client data heterogeneity and strict privacy constraints.
  - Centralized graph model training is difficult due to strict privacy regulations and sensitive graph data.
- Ideas: A reinforced federated graph domain generalization (RFGDG) combining Fourier feature generalization and reinforcement learning-based aggregation improves accuracy across heterogeneous clients.



Classification accuracy comparisons on Twitter (left) and Weibo (right) datasets under varying numbers of clients. Our proposed method RFGDG consistently achieves higher accuracy compared to baseline methods (RFFSage and FedSage). Particularly, as client numbers increase, RFGDG shows more prominent performance improvement, demonstrating its superior adaptability and robustness in scenarios with higher client heterogeneity.

# Main Contributions

- Contributions:
  - A reinforced federated graph domain generalization (RFGDG) framework combining federated learning with Random Fourier Feature transformation, effectively improving feature generalization across heterogeneous clients;
  - A dynamic aggregation strategy utilizing reinforcement learning to adaptively optimize client contributions, significantly enhancing global model accuracy and robustness;
  - Extensive experiments conducted on real-world datasets validate that the proposed RFGDG outperforms state-of-the-art federated graph methods under various scenarios with different data heterogeneity.

**Table 1** Comparison of Test Accuracy on Cora and Citeseer

Dataset	Federated	Graph	3 Clients		5 Clients		10 Clients		
			Test Set	Validation Set	Test Set	Validation Set	Test Set	Validation Set	
Cora	FedAvg	GCN	0.8053±0.0021	0.7499±0.0125	0.8068±0.0087	0.7502±0.0085	0.7799±0.0038	0.6948±0.0066	
		GAT	0.7997±0.0128	0.7436±0.0041	0.7736±0.0205	0.7469±0.0099	0.7097±0.0121	0.6647±0.0202	
		SAGE	0.8143±0.0034	0.7558±0.0048	0.8025±0.0078	0.7576±0.0085	0.7745±0.0093	0.6857±0.0084	
	FedProx	GCN	0.8096±0.0037	0.7481±0.0052	0.8081±0.0037	0.7524±0.0063	0.7798±0.0092	0.6901±0.0041	
		GAT	0.8006±0.0127	0.7481±0.0072	0.7771±0.0074	0.7399±0.0207	0.7683±0.0124	0.6703±0.0109	
		SAGE	0.8193±0.0053	0.7587±0.0074	0.8009±0.0094	0.7561±0.0085	0.7777±0.0089	0.6868±0.0166	
	MOON	GCN	0.7932±0.0034	0.7328±0.0107	0.7941±0.0065	0.7478±0.0072	0.7691±0.0118	0.6894±0.0053	
		GAT	0.7912±0.0089	0.7386±0.0027	0.7846±0.0093	0.7493±0.0128	0.7583±0.0013	0.6632±0.0047	
		SAGE	0.8081±0.0035	0.7512±0.0084	0.7990±0.0048	0.7524±0.0038	0.7695±0.0116	0.6807±0.0058	
	FedHiSyn	GCN	0.8137±0.0143	0.7521±0.0056	0.8096±0.0017	0.7513±0.0078	0.7805±0.0028	0.6924±0.0032	
		GAT	0.8016±0.0106	0.7509±0.0063	0.7787±0.0165	0.7435±0.0171	0.7627±0.00074	0.6685±0.0158	
		SAGE	0.8174±0.0093	0.7544±0.0061	0.8035±0.0081	0.7610±0.0048	0.7718±0.0059	0.6952±0.0102	
			<b>RFGDG</b>	<b>0.8393±0.0118</b>	<b>0.8044±0.0041</b>	<b>0.7956±0.0007</b>	<b>0.7912±0.0051</b>	<b>0.7348±0.0052</b>	
	Citeseer	FedAvg	GCN	0.7250±0.0022	0.7156±0.0081	0.7231±0.0047	0.7027±0.0024	0.7297±0.0029	0.6863±0.0063
			GAT	0.7326±0.0054	0.7114±0.0018	0.7270±0.0043	0.6817±0.0099	0.7151±0.0024	0.6600±0.0131
			SAGE	0.7339±0.0017	0.7303±0.0054	0.7339±0.0011	0.6997±0.0039	0.7322±0.0034	0.6803±0.0063
		FedProx	GCN	0.7249±0.0023	0.7162±0.0045	0.7221±0.0026	0.7057±0.0054	0.7286±0.0052	0.6881±0.0045
			GAT	0.7306±0.0085	0.7141±0.0096	0.7263±0.0066	0.6835±0.0096	0.7132±0.0046	0.6573±0.0039
SAGE			0.7352±0.0046	0.7282±0.0058	0.7291±0.0047	0.7021±0.0153	0.7281±0.0034	0.6907±0.0033	
MOON		GCN	0.7234±0.0014	0.7131±0.0053	0.7208±0.0033	0.6993±0.0029	0.7248±0.0087	0.6819±0.0075	
		GAT	0.7298±0.0056	0.7085±0.0067	0.7226±0.0052	0.6784±0.0089	0.7115±0.0042	0.6547±0.0082	
		SAGE	0.7258±0.0085	0.7266±0.0033	0.7264±0.0068	0.6974±0.0045	0.7258±0.0097	0.6842±0.0048	
FedHiSyn		GCN	0.7275±0.0035	0.7192±0.0083	0.7256±0.0023	0.7084±0.0030	0.7308±0.0049	0.6911±0.0041	
		GAT	0.7315±0.0061	0.7148±0.0072	0.7295±0.0051	0.6866±0.0075	0.7192±0.0033	0.6635±0.0057	
		SAGE	0.7362±0.0022	0.7328±0.0047	0.7352±0.0028	0.7025±0.0085	0.7293±0.0058	0.6925±0.0075	
			<b>RFGDG</b>	<b>0.7583±0.0045</b>	<b>0.7443±0.0035</b>	<b>0.7595±0.0063</b>	<b>0.7684±0.0022</b>	<b>0.7532±0.0081</b>	

**Table 3** Comparison of Test Accuracy on Twitter and Weibo

Dataset	Federated	Graph	3 Clients		5 Clients		10 Clients		
			Test Set	Validation Set	Test Set	Validation Set	Test Set	Validation Set	
Twitter	FedAvg	GCN	0.6416±0.0229	0.7647±0.0012	0.8184±0.0006	0.5949±0.0086	0.8251±0.0003	0.5953±0.0025	
		GAT	0.6995±0.0127	0.7070±0.0140	0.7796±0.0046	0.5853±0.0113	0.7956±0.0025	0.5781±0.0004	
		SAGE	0.7391±0.0126	0.8278±0.0045	0.8310±0.0033	0.7142±0.0056	0.8476±0.0063	0.7205±0.0033	
	FedProx	GCN	0.6264±0.0015	0.7624±0.0312	0.8203±0.0064	0.6252±0.0054	0.8208±0.0120	0.8249±0.0075	
		GAT	0.6807±0.0226	0.7735±0.0143	0.7917±0.0027	0.5144±0.0187	0.8095±0.0010	0.5912±0.0041	
		SAGE	0.7385±0.0034	0.8160±0.0113	<b>0.8407±0.0034</b>	0.7944±0.0068	<b>0.8532±0.0095</b>	0.8113±0.0044	
	MOON	GCN	0.6324±0.0145	0.7532±0.0114	0.8194±0.0054	0.6035±0.0065	0.8035±0.0053	0.6481±0.0087	
		GAT	0.6871±0.0156	0.7426±0.0068	0.7843±0.0063	0.5674±0.0134	0.7948±0.0034	0.5861±0.0064	
		SAGE	0.7284±0.0079	0.8043±0.0014	0.8215±0.0015	0.7681±0.0023	0.8317±0.0073	0.7392±0.0018	
	FedHiSyn	GCN	0.6586±0.0151	0.7751±0.00218	0.8224±0.0058	0.6325±0.0087	0.8256±0.0041	0.7745±0.0063	
		GAT	0.7045±0.0132	0.7831±0.0039	0.7963±0.0072	0.6031±0.0074	0.8079±0.0056	0.6098±0.0017	
		SAGE	0.7407±0.0108	0.8243±0.0041	0.8352±0.0044	0.7822±0.0047	0.8454±0.0066	0.8051±0.0081	
			<b>RFGDG</b>	<b>0.7693±0.0184</b>	<b>0.8394±0.0046</b>	<b>0.7755±0.0006</b>	<b>0.8436±0.0078</b>	<b>0.8661±0.0068</b>	
	Weibo	FedAvg	GCN	0.5664±0.0072	0.5359±0.0048	0.6703±0.0043	0.5739±0.0182	0.6708±0.0095	0.5955±0.0014
			GAT	0.5841±0.0128	0.4911±0.0045	0.5703±0.0020	0.5567±0.0172	0.5785±0.0012	0.5220±0.0033
			SAGE	0.6420±0.0094	0.6795±0.0010	0.6871±0.0084	0.6141±0.0212	0.6936±0.0045	0.6620±0.0063
		FedProx	GCN	0.6170±0.0013	0.5378±0.0117	0.6810±0.0017	0.6084±0.0072	0.7083±0.0054	0.5936±0.0027
			GAT	0.6314±0.0102	0.5944±0.0034	0.5827±0.0067	0.5842±0.0017	0.5708±0.0093	0.5408±0.0043
SAGE			0.6458±0.0068	0.6714±0.0058	0.6796±0.0010	0.6776±0.0142	0.7077±0.0053	0.6864±0.0031	
MOON		GCN	0.5874±0.0051	0.5271±0.0063	0.6609±0.0016	0.5682±0.0079	0.6876±0.0012	0.5917±0.0045	
		GAT	0.6114±0.0067	0.5794±0.0021	0.5635±0.0044	0.5463±0.0078	0.5596±0.0077	0.5204±0.0068	
		SAGE	0.6378±0.0066	0.6702±0.0087	0.6743±0.0035	0.6334±0.0088	0.6897±0.0028	0.6785±0.0042	
FedHiSyn		GCN	0.6271±0.0075	0.5314±0.0063	0.6815±0.0024	0.6151±0.0085	0.7052±0.0062	0.6036±0.0055	
		GAT	0.6418±0.0052	0.6025±0.0012	0.5988±0.0011	0.5891±0.0076	0.6913±0.0023	0.6855±0.0091	
		SAGE	0.6545±0.0071	0.6753±0.0016	0.6854±0.0044	0.6718±0.0059	0.7032±0.0087	0.6825±0.0036	
			<b>RFGDG</b>	<b>0.7083±0.0045</b>	<b>0.6815±0.0218</b>	<b>0.7058±0.0167</b>	<b>0.6916±0.0304</b>	<b>0.7196±0.0024</b>	

Accuracy comparisons of federated graph methods, showing superior performance of RFGDG across multiple datasets and client numbers.