

# Domain Adaptation for Time Series Forecasting via Attention Sharing

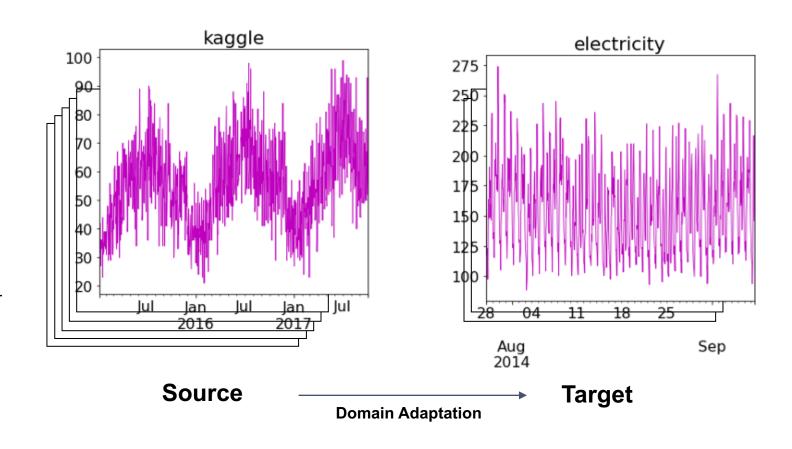
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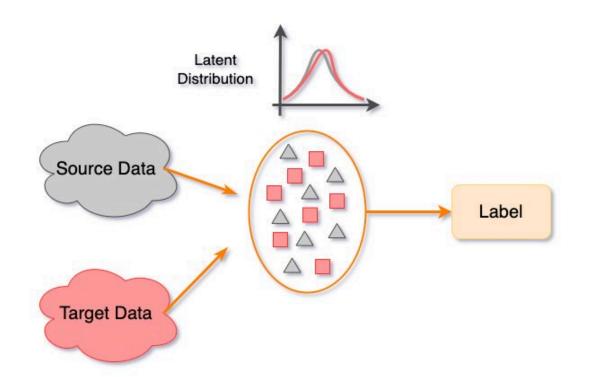
#### Problem

- Forecasting with data scarcity
  - Limited data from a target domain
  - Abundant data from a source domain
- Domain adaptation:
  - Learn a model mainly on the data-rich source domain
  - Transfer certain knowledge to the datascarce target domain by adaptation





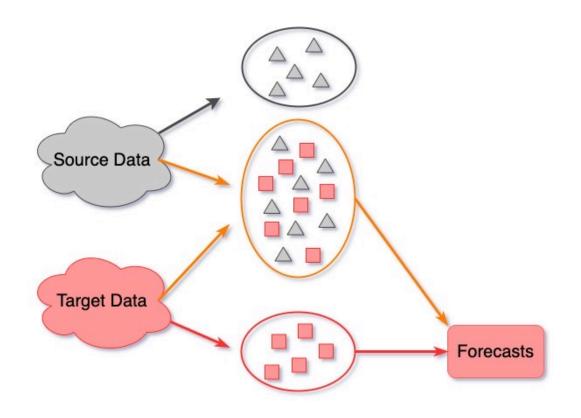
## Existing Domain Adaptation



- Learn domain-invariant features
- Distinctions between domains do not affect predictions from features
- Methods:
  - Metric-based regularization
  - Adversarial training



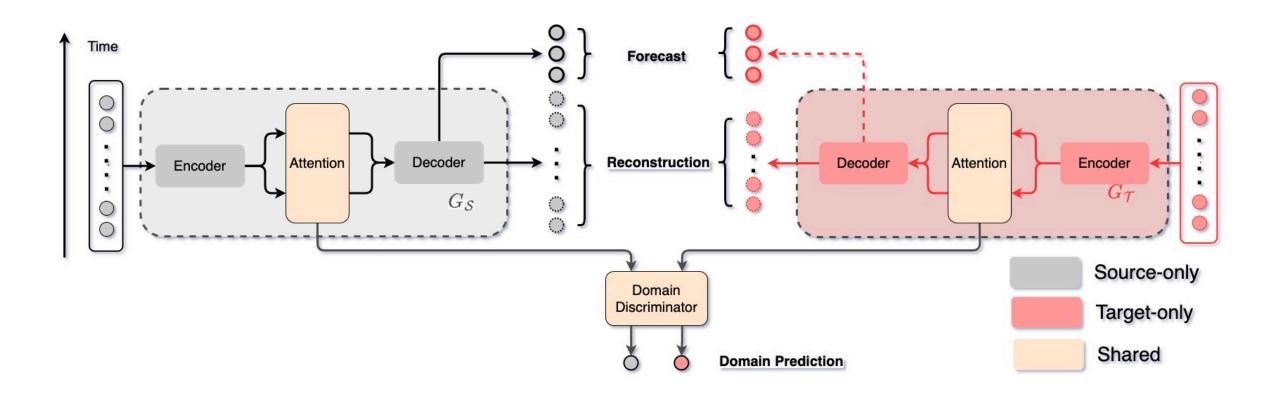
#### DA in Forecasting



- Domain-specific features are necessary for domain-dependent forecasts
- Domain-invariant features to connect both domains
  - Query-key matching in Attention
- Domain-specific features are retained to make domain-dependent forecasts
  - Value combination in Attention

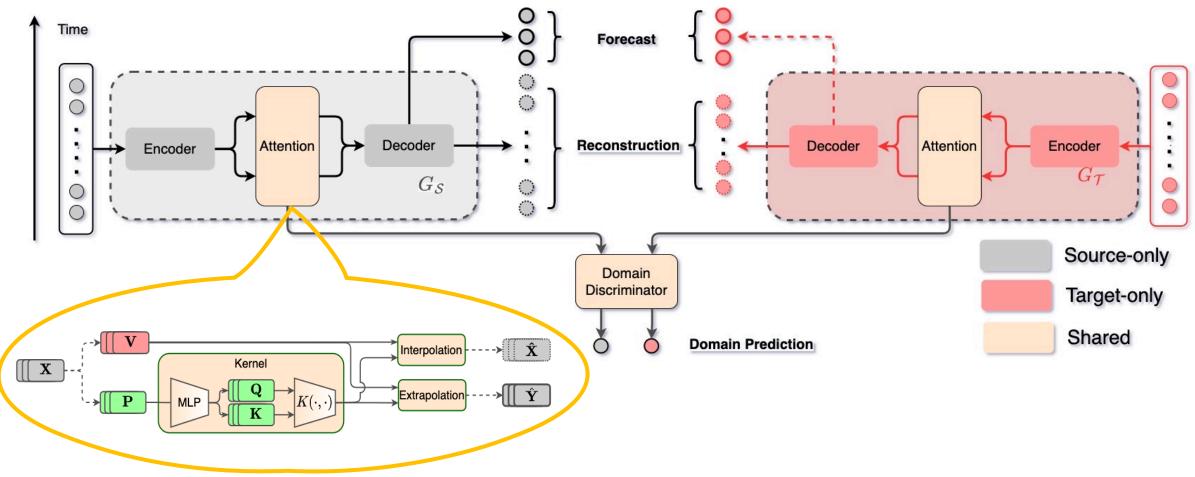


# Domain Adaptation Forecaster





### Attention Sharing Strategy

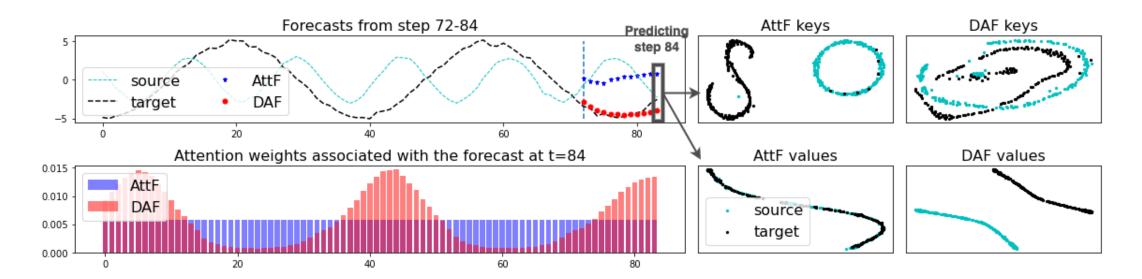




#### How it works

#### **Key observations**

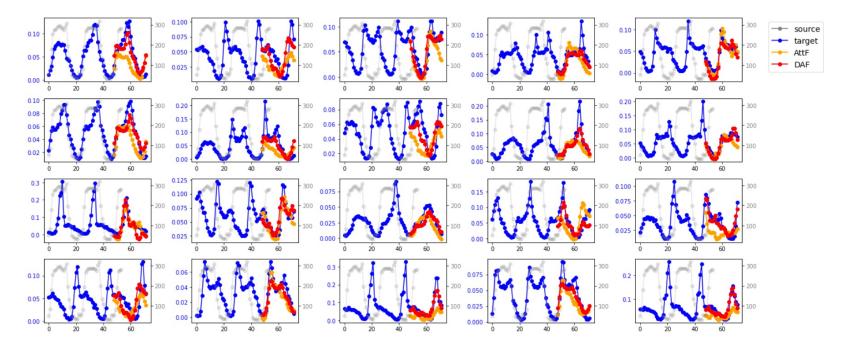
- Attention keys are aligned in DAF but not in single-domain Attention Forecaster (AttF)
- More reasonable attention weights in DAF than in AttF
- Better forecasts from DAF
- Attention values stay distinct across domains





#### Results

$\mathcal{D}_{\mathcal{T}}$	$\mathcal{D}_{\mathcal{S}}$	au	DAR	AttF	SASA	DATSING	RDA-DANN	RDA-ADDA	RDA-MMD	DAF
traf	elec wiki		0.205±0.015	$0.182 \pm 0.007$	0.177±0.004	$0.184{\pm}0.004$	$0.181 {\pm} 0.009$	$0.174 \pm 0.005$	$0.186{\pm}0.004$	0.169±0.002
uai		24			$0.197\pm0.001$	$0.189 \pm 0.005$	$0.180\pm0.004$	$0.181 \pm 0.003$	$0.179\pm0.004$	$0.176 \pm 0.004$
elec	traf	traf	0.141±0.023	0.137±0.005	$0.164\pm0.001$	$0.137 \pm 0.003$	$0.133 \pm 0.005$	$0.134 \pm 0.002$	$0.140\pm0.006$	$0.125 \pm 0.008$
elec	sales				$0.160\pm0.001$	$0.149 \pm 0.009$	$0.135 {\pm} 0.007$	$0.142 \pm 0.003$	$0.144 \pm 0.003$	$0.123 \pm 0.005$
wiki	traf		0.055±0.010	0.050±0.003	0.053±0.001	$0.049\pm0.002$	$0.047 \pm 0.005$	$0.045 \pm 0.003$	$0.045{\pm}0.003$	0.042±0.004
	sales				$0.053\pm0.001$	$0.052 \pm 0.004$	$0.053 \pm 0.002$	$0.049 \pm 0.003$	$0.052 \pm 0.004$	$0.049 \pm 0.003$
sales	elec	elec '	0.305±0.005	0.308±0.002	$0.451 \pm 0.001$	$0.301 \pm 0.008$	$0.297 \pm 0.004$	$0.281 \pm 0.001$	$0.291 \pm 0.004$	$0.277 \pm 0.005$
sales	wiki				$0.301\pm0.001$	$0.305 \pm 0.008$	$0.287 \pm 0.009$	$0.287 \pm 0.002$	$0.289 \pm 0.003$	0.280±0.007





#### Takeaways

- Knowledge of forecasting can be transferred from data-rich domains to data-scarce domains
- 2. Attention mechanism is suitable for domain adaptation, where
  - 1. queries/keys can be induced to be invariant across domains;
  - 2. values can stay distinct to make domain-dependent forecasts