

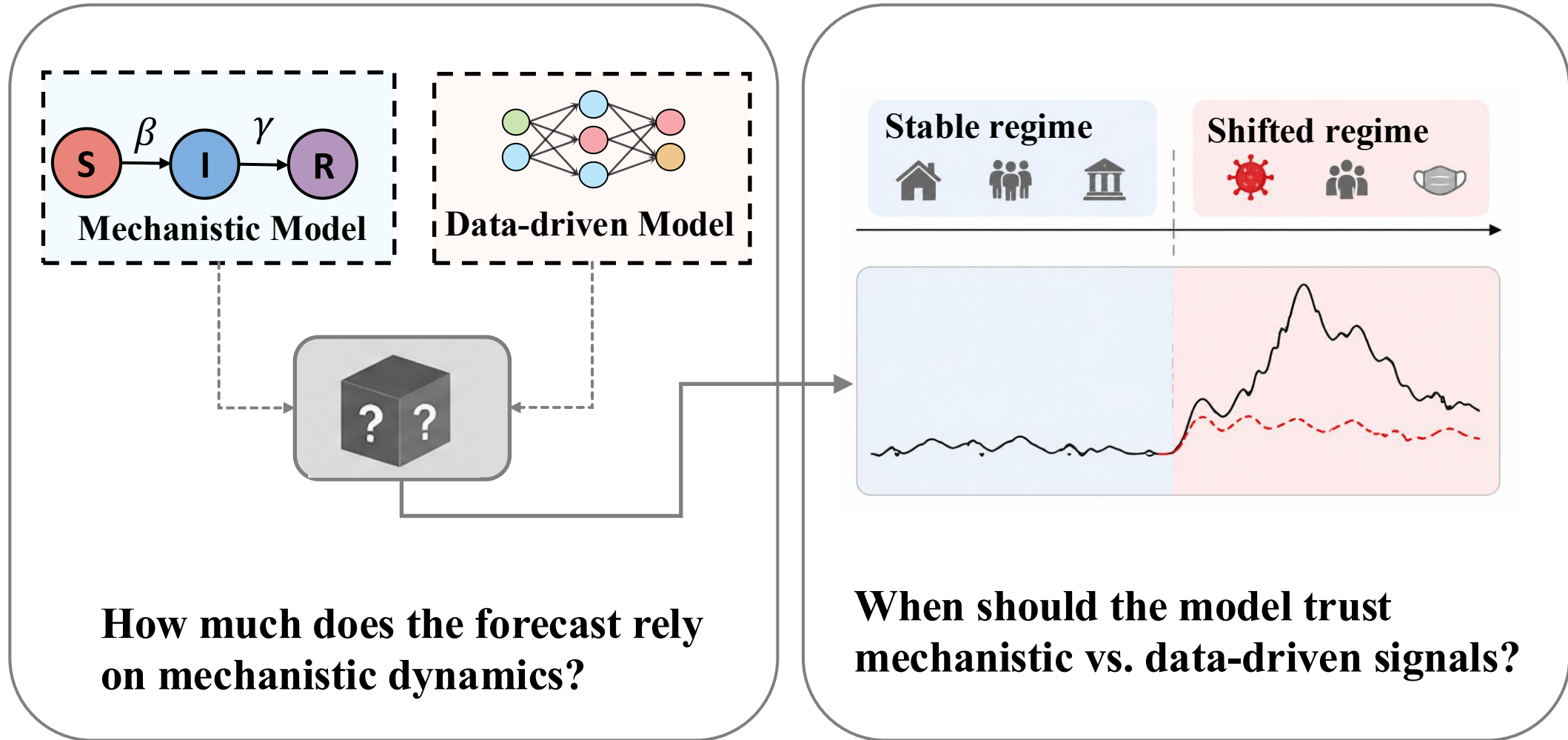
Bridging Dynamics and Data: A Unified Diffusion Framework for Mechanistically- Informed Epidemic Forecasting

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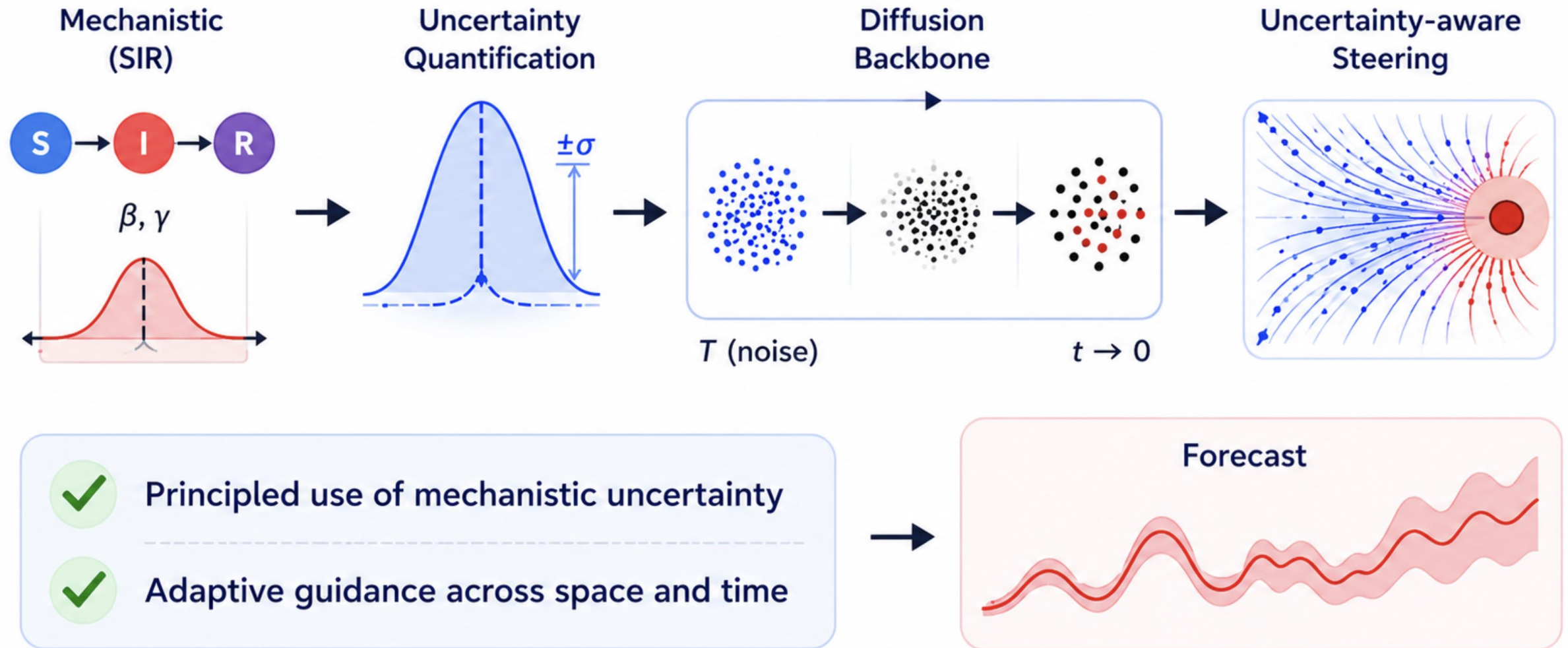
May 31, 2026



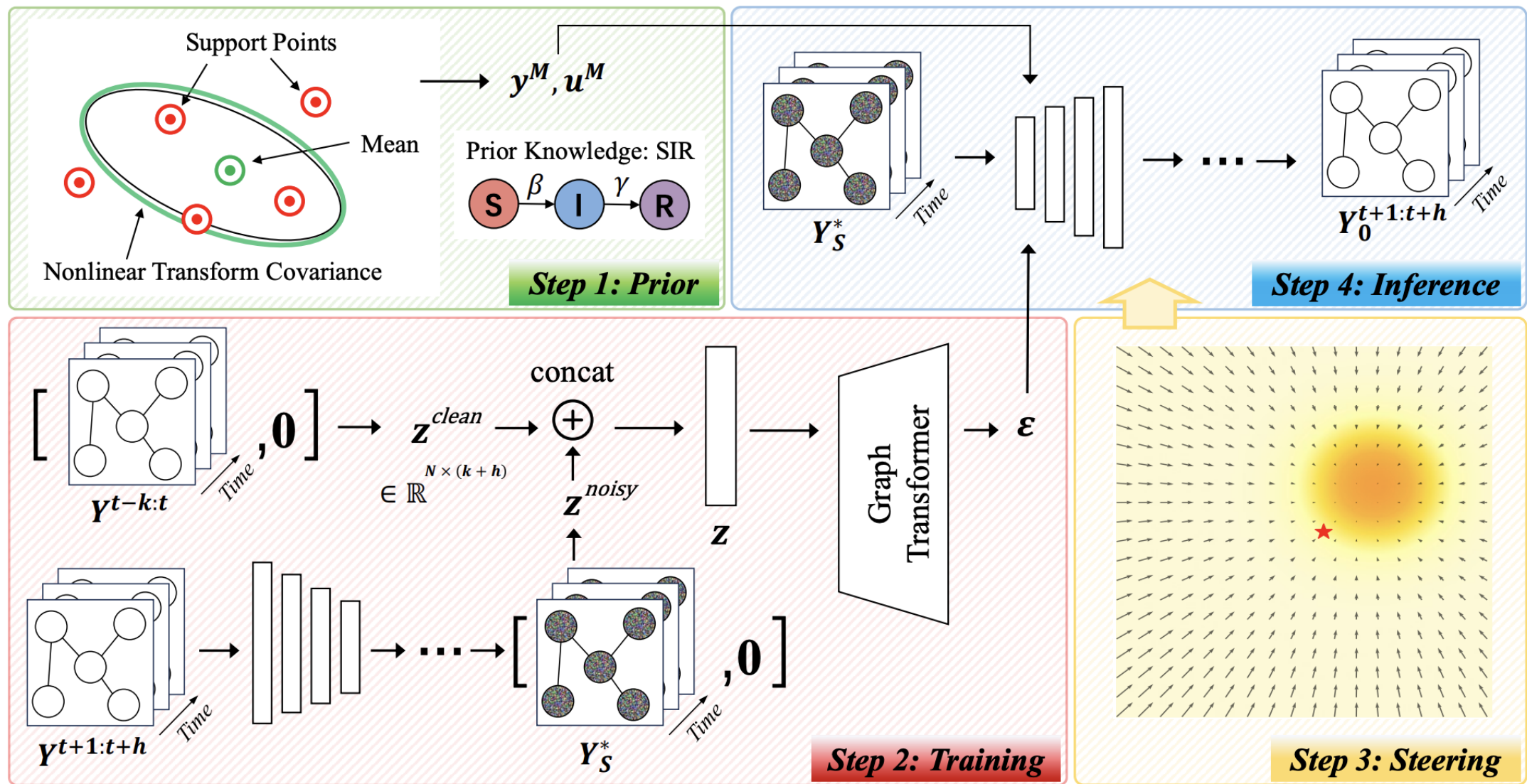
Motivation



Our Solution: EpiDiff

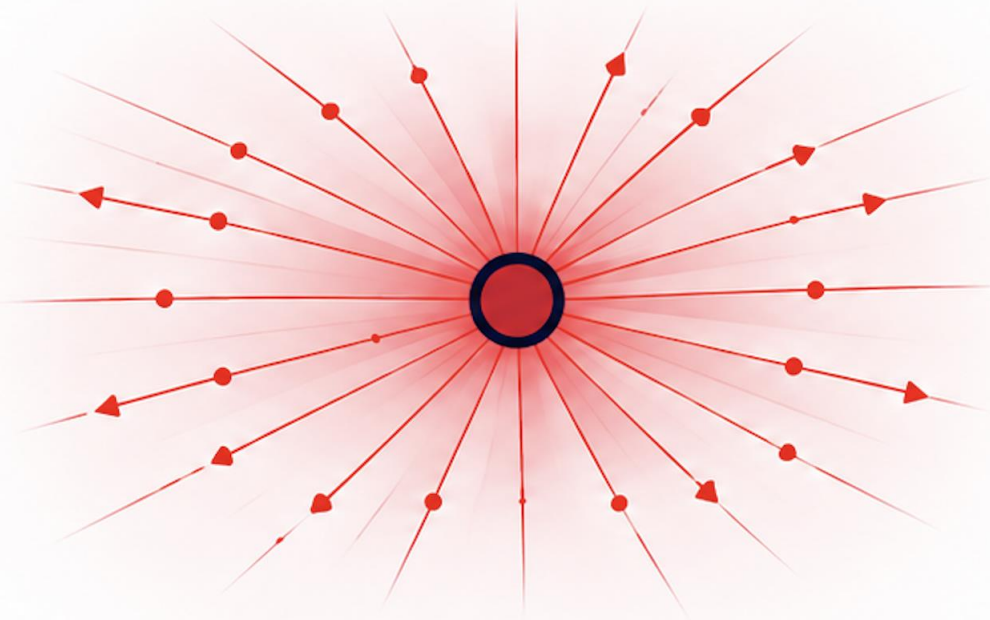


EpiDiff Pipeline

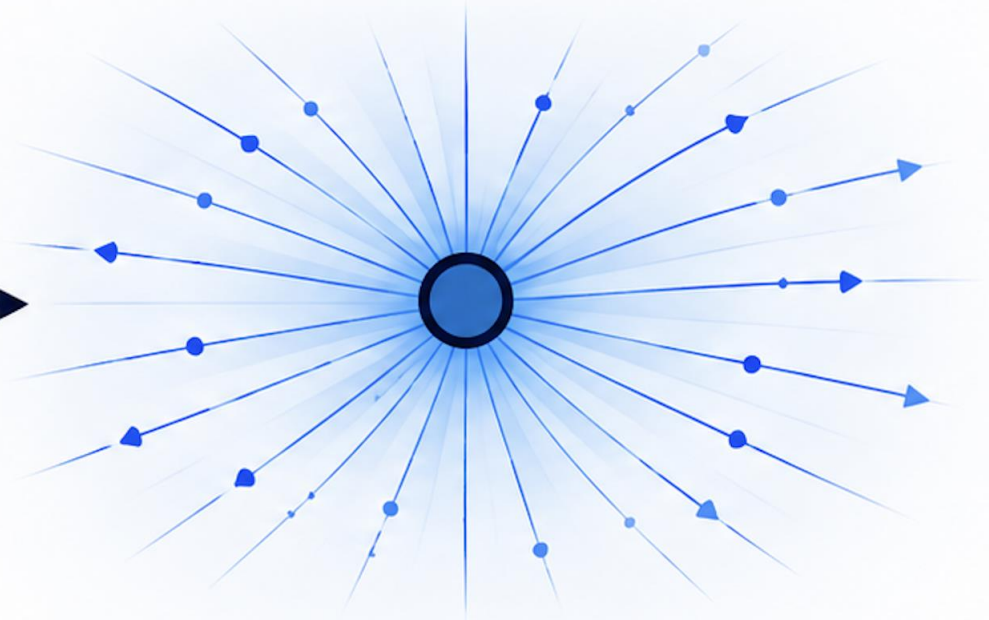


Intuition: Uncertainty Is the Switch

High uncertainty



Low uncertainty



Rely more on
data (diffusion)



Adaptive weight



Rely more on
mechanistic prior

Model Components



Mechanistic Prior (SIR)

$$\frac{dS}{dt} = -\beta SI, \quad \frac{dI}{dt} = \beta SI - \gamma I, \dots$$



Uncertainty Quantification

Posterior $p(\theta | D) \rightarrow$ mean μ , std σ



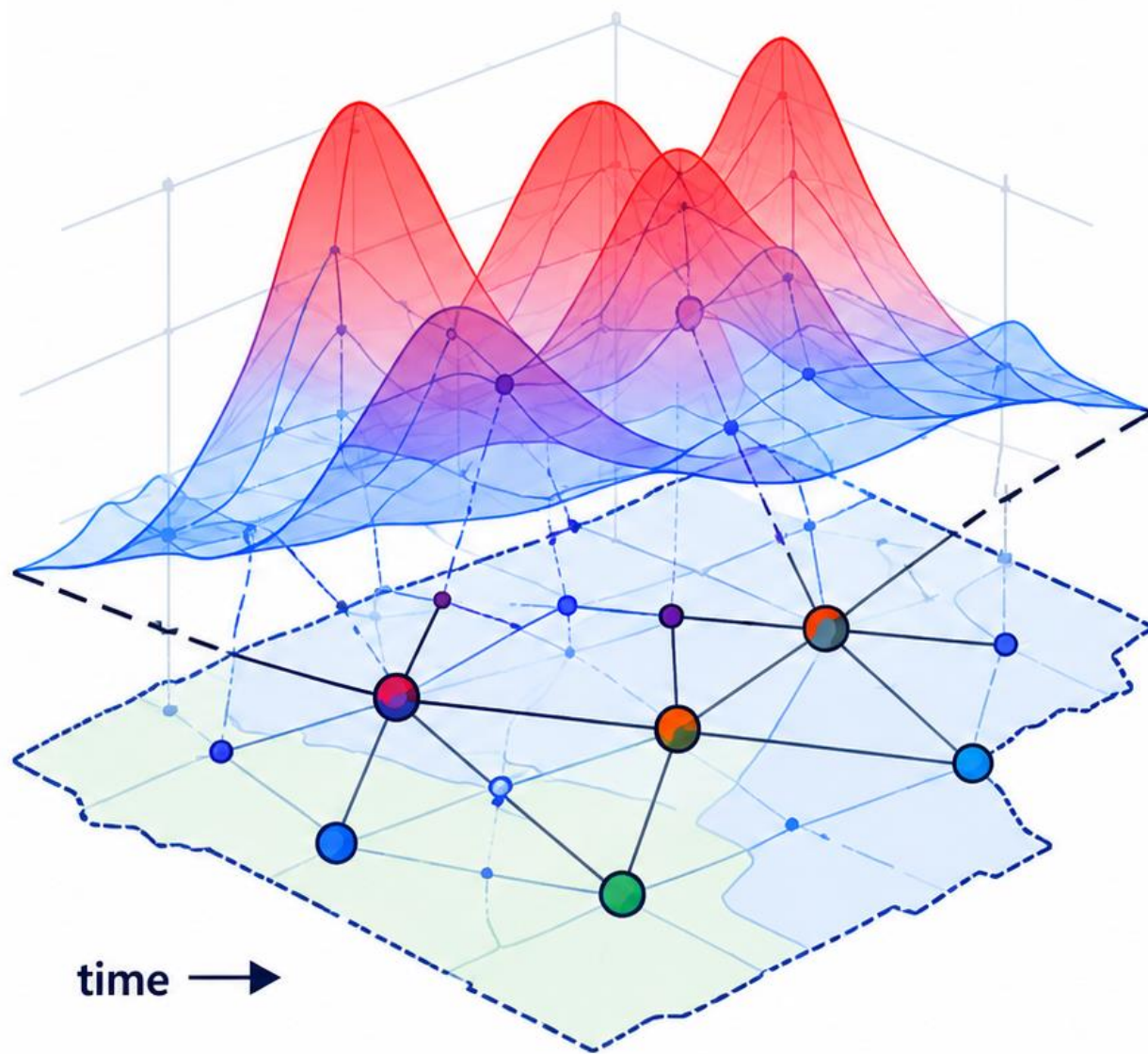
Diffusion Backbone

Spatiotemporal Graph Transformer



Steering Rule

Guidance $\propto$ Uncertainty $\sigma(x, t)$



Dataset

covid-US



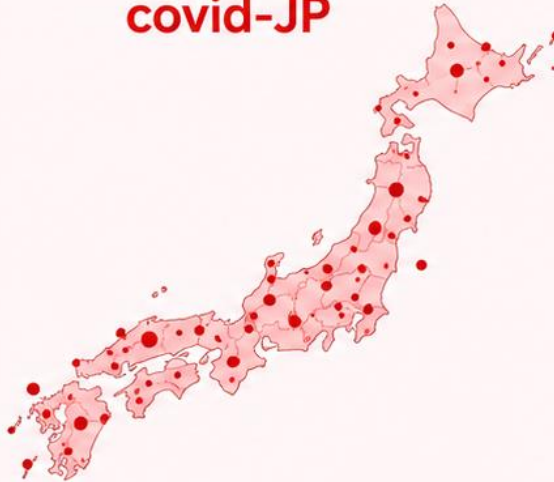
📅 Apr 15, 2020 – Apr 15, 2021

🕒 Daily cases

📍 County level

📊 Train [0, 262] Val [262, 313] Test [313, 366]

covid-JP



📅 Apr 1, 2020 – Sep 21, 2021

🕒 Daily cases

📍 Prefecture level

📊 Train [0, 327] Val [327, 433] Test [433, 539]

influenza-US



📅 Week 1, 2019 – Last week of 2021

🕒 Weekly estimated cases (ILI)

📍 State level

📊 Train [0, 98] Val [98, 127] Test [127, 158]



Non-stationary dynamics



Data sparsity



Policy & behavior shifts

Performance

Table 1. Performance comparison on *covid-US*, *covid-JP*, and *influenza-US* in RMSE and MAE ($\mathcal{R}$ and $\mathcal{P}$, lower is better). Best results are **in pink** and runner-ups are underlined.

Methods	<i>covid-US</i>				<i>covid-JP</i>				<i>influenza-US</i>			
	$h = 7$		$h = 14$		$h = 7$		$h = 14$		$h = 7$		$h = 14$	
	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$
PatchSIR	1791 $\pm$ 0	891 $\pm$ 0	2077 $\pm$ 0	886 $\pm$ 0	313 $\pm$ 0	98 $\pm$ 0	332 $\pm$ 0	106 $\pm$ 0	323 $\pm$ 0	<u>82</u> $\pm$ 0	454 $\pm$ 0	138 $\pm$ 0
AR	1791 $\pm$ 16	835 $\pm$ 16	1943 $\pm$ 23	865 $\pm$ 7	457 $\pm$ 57	159 $\pm$ 11	476 $\pm$ 11	158 $\pm$ 4	4579 $\pm$ 2429	2040 $\pm$ 1814	2328 $\pm$ 236	1172 $\pm$ 84
ARMA	1832 $\pm$ 29	843 $\pm$ 2	1906 $\pm$ 18	852 $\pm$ 9	460 $\pm$ 32	156 $\pm$ 8	496 $\pm$ 6	166 $\pm$ 3	4423 $\pm$ 1971	2289 $\pm$ 1271	2314 $\pm$ 150	1174 $\pm$ 68
LSTM	1797 $\pm$ 59	867 $\pm$ 48	1956 $\pm$ 49	962 $\pm$ 52	408 $\pm$ 10	135 $\pm$ 3	398 $\pm$ 19	135 $\pm$ 7	353 $\pm$ 29	119 $\pm$ 21	<u>354</u> $\pm$ 42	133 $\pm$ 7
GRU	1654 $\pm$ 81	771 $\pm$ 52	1949 $\pm$ 38	974 $\pm$ 30	364 $\pm$ 12	120 $\pm$ 3	383 $\pm$ 8	130 $\pm$ 3	377 $\pm$ 41	142 $\pm$ 13	498 $\pm$ 303	192 $\pm$ 75
STGCN	2131 $\pm$ 12	877 $\pm$ 23	2446 $\pm$ 51	1107 $\pm$ 12	315 $\pm$ 83	96 $\pm$ 23	326 $\pm$ 24	103 $\pm$ 9	414 $\pm$ 93	160 $\pm$ 42	588 $\pm$ 128	248 $\pm$ 56
ColaGNN	2740 $\pm$ 758	1175 $\pm$ 339	3055 $\pm$ 356	1403 $\pm$ 181	<u>213</u> $\pm$ 20	<u>63</u> $\pm$ 4	305 $\pm$ 47	<u>88</u> $\pm$ 11	250 $\pm$ 17	85 $\pm$ 9	347 $\pm$ 228	<u>129</u> $\pm$ 27
STGODE	1763 $\pm$ 96	719 $\pm$ 34	2716 $\pm$ 107	1010 $\pm$ 40	269 $\pm$ 10	79 $\pm$ 2	309 $\pm$ 23	96 $\pm$ 7	643 $\pm$ 194	261 $\pm$ 94	1415 $\pm$ 460	570 $\pm$ 200
DiffSTG	<u>1236</u> $\pm$ 28	<u>443</u> $\pm$ 11	1284 $\pm$ 20	<u>444</u> $\pm$ 14	246 $\pm$ 28	81 $\pm$ 9	423 $\pm$ 35	147 $\pm$ 13	344 $\pm$ 29	151 $\pm$ 69	520 $\pm$ 44	254 $\pm$ 54
STAN	1364 $\pm$ 196	1150 $\pm$ 174	<u>1265</u> $\pm$ 98	1046 $\pm$ 90	222 $\pm$ 87	127 $\pm$ 40	224 $\pm$ 88	128 $\pm$ 41	4599 $\pm$ 42	4306 $\pm$ 32	3750 $\pm$ 79	3270 $\pm$ 79
EARTH	2145 $\pm$ 324	603 $\pm$ 103	2664 $\pm$ 272	794 $\pm$ 79	303 $\pm$ 73	100 $\pm$ 13	394 $\pm$ 102	134 $\pm$ 21	539 $\pm$ 172	190 $\pm$ 47	2027 $\pm$ 1160	629 $\pm$ 302
EpiColaGNN	1825 $\pm$ 84	745 $\pm$ 58	2653 $\pm$ 22	1235 $\pm$ 9	282 $\pm$ 33	83 $\pm$ 11	327 $\pm$ 19	99 $\pm$ 6	1311 $\pm$ 67	693 $\pm$ 36	1514 $\pm$ 64	807 $\pm$ 36
EpiDiff	1096 $\pm$ 5	374 $\pm$ 1	1232 $\pm$ 5	422 $\pm$ 16	188 $\pm$ 7	62 $\pm$ 2	<u>247</u> $\pm$ 2	86 $\pm$ 1	<u>304</u> $\pm$ 1	82 $\pm$ 0.4	423 $\pm$ 1	125 $\pm$ 0.1

Ablation

Table 2. Ablation study of different variants. Best results are **in bold** and runner-ups are underlined.

Variants	<i>covid-US</i>				<i>covid-JP</i>				<i>influenza-US</i>			
	$h = 7$		$h = 14$		$h = 7$		$h = 14$		$h = 7$		$h = 14$	
	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$	$\mathcal{R} \downarrow$	$\mathcal{P} \downarrow$
w/o Diffusion	1791	891	2077	886	313	98	332	106	<u>323</u>	<u>82</u>	454	<u>138</u>
w/o Steering	1080 ± 2	348 ± 1	1199 ± 6	386 ± 11	370 ± 22	94 ± 5	417 ± 4	<u>121</u> ± 1	438 ± 19	224 ± 7	1787 ± 87	928 ± 59
w/o Uncertainty	1099 ± 5	376 ± 1	1253 ± 5	438 ± 18	<u>316</u> ± 5	<u>97</u> ± 2	<u>365</u> ± 1	122 ± 1	328 ± 1	147 ± 1	<u>446</u> ± 0.4	155 ± 0.3
EpiDiff	<u>1096</u> ± 5	<u>374</u> ± 1	<u>1232</u> ± 5	<u>422</u> ± 16	188 ± 7	62 ± 2	247 ± 2	86 ± 1	304 ± 1	82 ± 0.4	423 ± 1	125 ± 0.1

Mechanistic Uncertainty Visualization

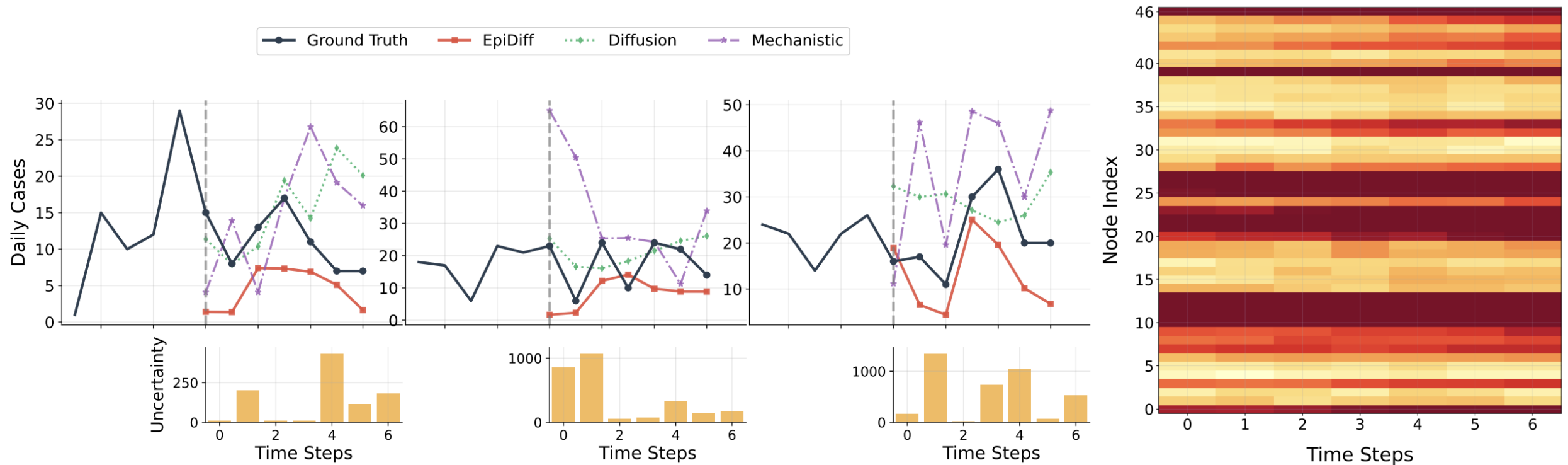


Figure 4. Left: Node 30 prediction and uncertainty at windows 30, 60, and 70 on *covid-JP*. **Right:** Uncertainty heatmap for nodes and steps on *covid-JP*. Darker squares indicate greater uncertainty.

Thanks for listening!