Challenges of Epidemiological Forecasting in a Pandemic

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- We were awarded CDC Center of Excellence in September 2019

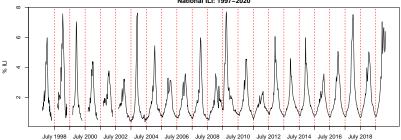
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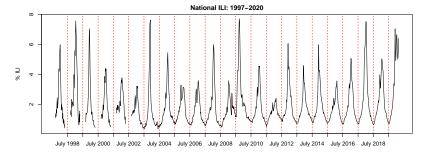
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National ILI: 1997-2020

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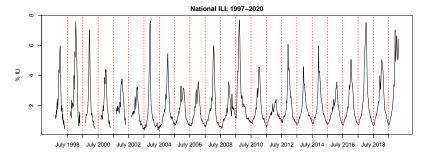
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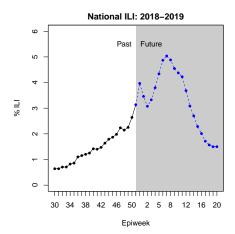
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- Data is subject to revisions, sometimes significant ones

Severity pyramid



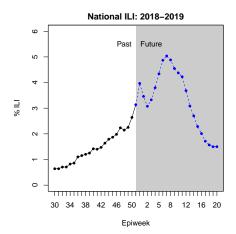
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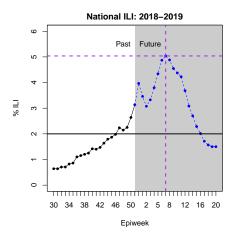
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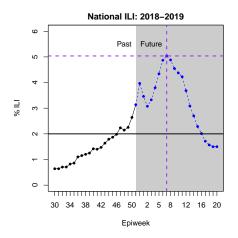
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Same for hospitalizations

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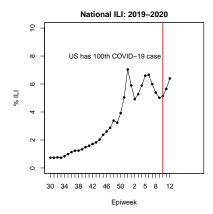
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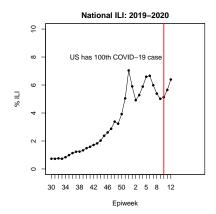
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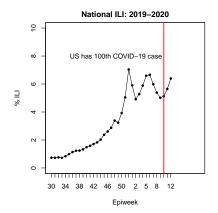
Our system is an ensemble of 6 separate forecasting models



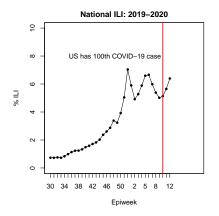
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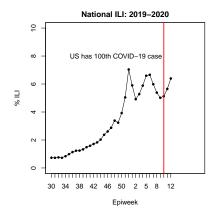


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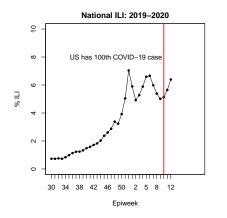
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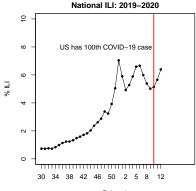
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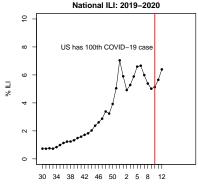
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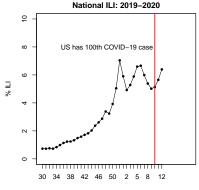
Epiweek

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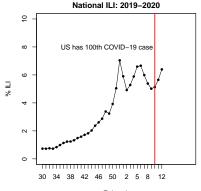
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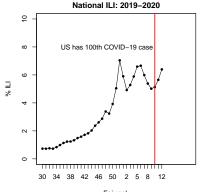
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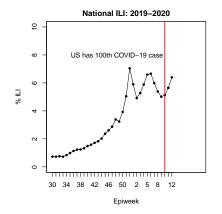


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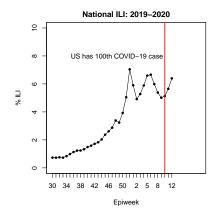
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