

Sea Ice Outlook
2021 July Report
Individual Outlook

Name of contributor or name of contributing organization:

UPenn Group 1 (FELR)

Is this contribution from a person or group not affiliated with a research organization?

Name and organization for all contributors. Indicate primary contact and total number of people who may have contributed to your Outlook, even if not included on the author list.

Upenn Group 1 (FELR)

Do you want your June contribution to automatically be included in subsequent reports? (If yes, you may still update your contribution via the submission form.)

Include this submission in this month's report ONLY

What is the type of your Outlook projection?

Statistical

Starting in 2017 we are accepting both pan-Arctic and pan-Antarctic sea ice extent (either one or both) of the September monthly mean. As in 2016, we are also collecting Alaskan regional sea ice extent. To be consistent with the validating sea ice extent index from NSIDC, if possible, please first compute the average sea ice concentration for the month and then compute the extent as the sum of cell areas > 15%.

a) Pan-Arctic September extent prediction in million square kilometers.

3.95

b) same as in (a) but for pan-Antarctic. If your method differs substantially from that for the Arctic, please enter it as a separate submission.

c) same as in (b) but for the Alaskan region. Please also tell us maximum possible extent if every ocean cell in your region were ice covered.

"Executive summary" of your Outlook contribution (using 300 words or less) describe how and why your contribution was formulated. To the extent possible, use non-technical language.

The UPenn group is composed of economists and statisticians interested in predictive modeling of many aspects of climate in its relation to economic activity. The Arctic -- and Arctic sea ice in particular -- is of particular interest to us. As is well known, the Arctic is warming about twice as fast as the global average, and the Arctic amplification in surface air temperature is of course closely connected to the dramatic multi-decade reduction in Northern sea ice. This loss of sea ice is one of the most conspicuous warning signs of \textit{current} climate change, and it also plays an integral role in the timing and intensity of \textit{future} global climate change. Not surprisingly then, we are keenly interested in predictive modeling of Arctic sea ice, particularly summer ice.

Brief explanation of Outlook method (using 300 words or less).

We have supplied a forecast based on a statistical model with trend, a feed-forward loop, and stochastic shocks, estimated by direct projection. In the modeling process we explore different levels of aggregation of the underlying high-frequency (daily) concentration data and associated sea ice extent, and we tune the aggregation to optimize the predictive bias/variance tradeoff in forecasting September extent. It turns out that our in-sample forecast errors (residuals) are approximately Gaussian, which we exploit in making our out-of-sample forecast for September. The predictive density is Gaussian, with the mean 3.77 million square kilometers and standard deviation 0.415 million square kilometers. (By symmetry, the mean and median coincide.) The approximate 95% interval that we report is the mean plus or minus 2 standard deviations.

Tell us the dataset used for your initial Sea Ice Concentration (SIC).

na

Tell us the dataset used for your initial Sea Ice Thickness (SIT) used. Include name and date.

na

If you use a dynamic model, please specify the name of the model as a whole and each component including version numbers and how the component is initialized:

If available from your method.

a) Uncertainty/probability estimates:

Median

3.95

Lower error bound

3.22

Lower error bound

4.68

Standard Deviation

0.365

b) Brief explanation/assessment of basis for the uncertainty estimate (1-2 sentences).

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c) Brief description of any post-processing you have done (1-2 sentences).

estimated stochastic model