Preliminary Quantification of Uncertainty in Parameters of the **Glaciers and Small Ice Caps Melt Model** Mercedes Gainor^{1,2}, Kelsey Ruckert Student Mentor^{1,3}, Chris Forest Faculty Co-mentor^{1,4}, and Klaus Keller Faculty Mentor^{1,3,5}

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ABSTRACT

Glaciers and Small Ice Caps (GSICs) contribute to sea-level rise. Previous studies have used semi-empirical models to derive projections of GSIC melt. Wigley and Raper(2005) modified a glacier melt model Intergovernmental Panel on Climate Change Third Assessment Report which projects future sea level-rise. Here we present preliminary results to quantify the sensitivity of this model to its parameters and to quantify the uncertainty surrounding the parameter estimates. This approach has the potential to yield a better understanding of what drives sea-level rise due to GSIC melt, to improve projections, and to inform the design of climate risk management strategies.

INTRODUCTION

- Semi-empirical glacier models have been produced which project future sea level from the melting of GSICs.
- Model projections of the contribution of sea level rise from GSIC ranges from 0.08 to 0.39 m sea level equivalent in 2100.^[1]
- We recreate a semi-empirical model from Wigley and Raper which is an extended model of IPPC TAR that uses the melt parameters: β_0 , V₀, n, and the intial contribution of GSICs to project future sea level rise.^[2] Many times model input/output relationships are poorly understood which
- leads to a need for sensitivity analysis.
- Model parameters are sources of uncertainty which can limit how confident scientists or decision makers are with the response of the model.
- This study uses statistical methods to answer the following questions:
 - How sensitive is the model to its parameters?
 - What values should these parameters be set to?
 - What is the range of uncertainty about the parameters that make up the glacier melt model?
 - How different are the projections with the range of uncertainties?

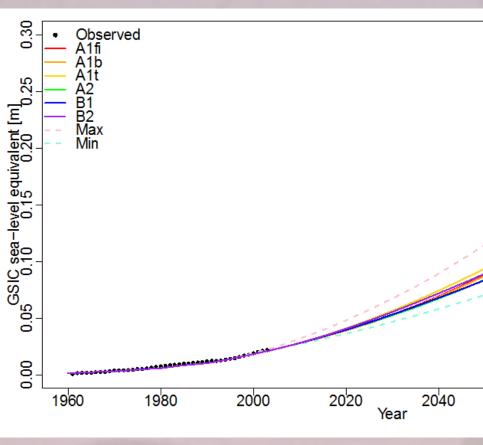


Figure 1: Comparison of Best Guess hindcasts and projections for each IPCC future emission scenario. The max scenario has a temperature change of 5.80°C and the min has a temperature change of 1.37°C.

GSIC MELT ALGORITHM

- The surface mass balance algorithm is based on the work of Gregory and Oerlemans [1998] and Van der Wal and Wild [2001] in which the total melt in sea level equivalent is calculated assuming the GSIC volume will stay constant and corrected so that the mass balance sensitivity will decrease as the volume of GSIC decreases when melting occurs. ^[3,4]
- The original algorithm works to 2100; however, Wigley and Raper modified the formula to extend it to 2400 by using global mass balance sensitivity as a function of the GSIC area directly:

$Dg_{s}/dt = \beta_{0}(0.15 + T(t))(1 - g_{s}/V_{0})^{n}$	
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Symbol	Meaning mass balance	Units	Bounds	Best Estimate
β ₀	sensitivity	cm/yr/C	0-1	0.142
V ₀	initial volume of all GSIC in 1961	cm	30-50	30
n Initial Value	exponent related to the size of GSIC GSIC sea-level rise	dimensionless	0.65-1.5	1.5
(InV)	contribution in 1961	cm	unconstrained	0.132

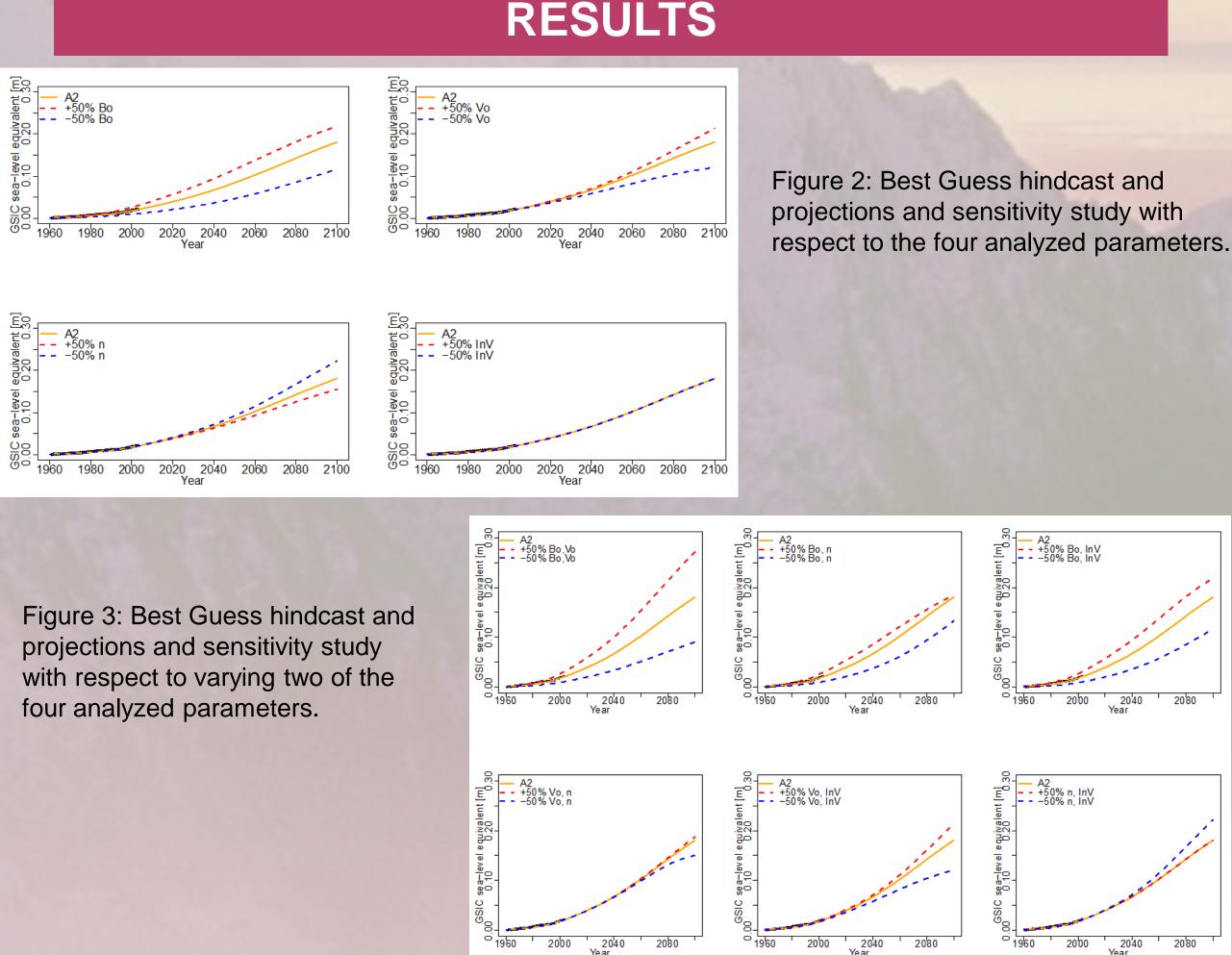
¹The Network for Sustainable Climate Risk Management (SCRiM), The Pennsylvania Sate University ²SCRiM Undergraduate Research Scholar

METHODS

- 2080

90% C.I. 0.121-0.168 31.0-48.9 0.696-1.45 -0.015-0.204

- R programming language
- Estimate the best fit by minimizing the root mean squared error using Differential Evolution^[5]
- Use a bootstrap resampling method to estimate the uncertainty surrounding the parameter estimates. We assumed independent and identically distributed errors.^[6]
- Time series and temperatures dataset from National Oceanic and Atmospheric Administration^[7]
- GSIC Data from National Snow & Ice Data Center^[8]
- A2 future emission scenario which has a cumulative greenhouse gas emission of ~1350-1850 GtC from 1990-2100 and a 3.79C temperature change in 2100 ^[9]



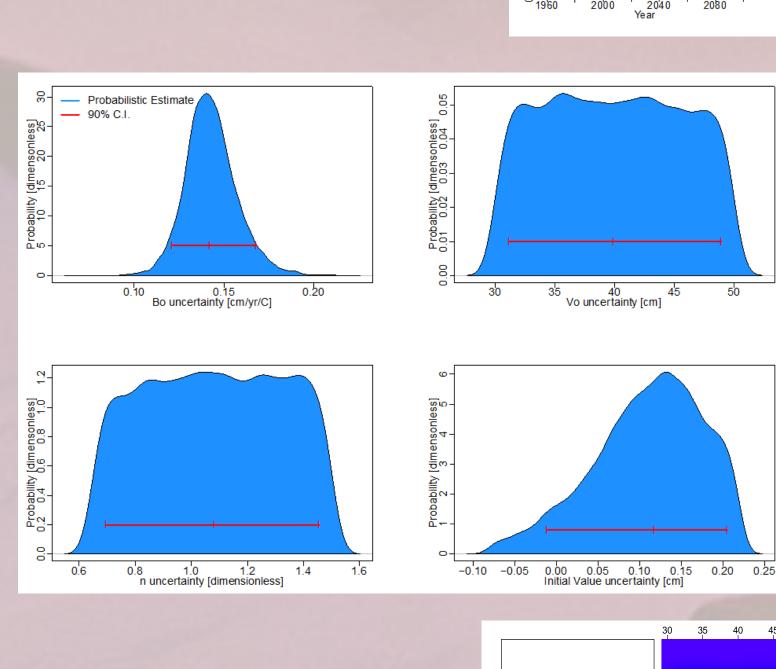


Figure 5: Density image that shows the correlation between the parameter estimates derived from the bootstrap.

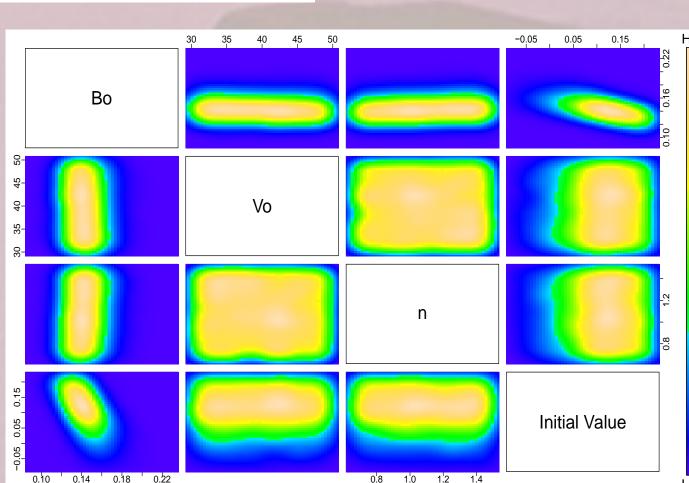
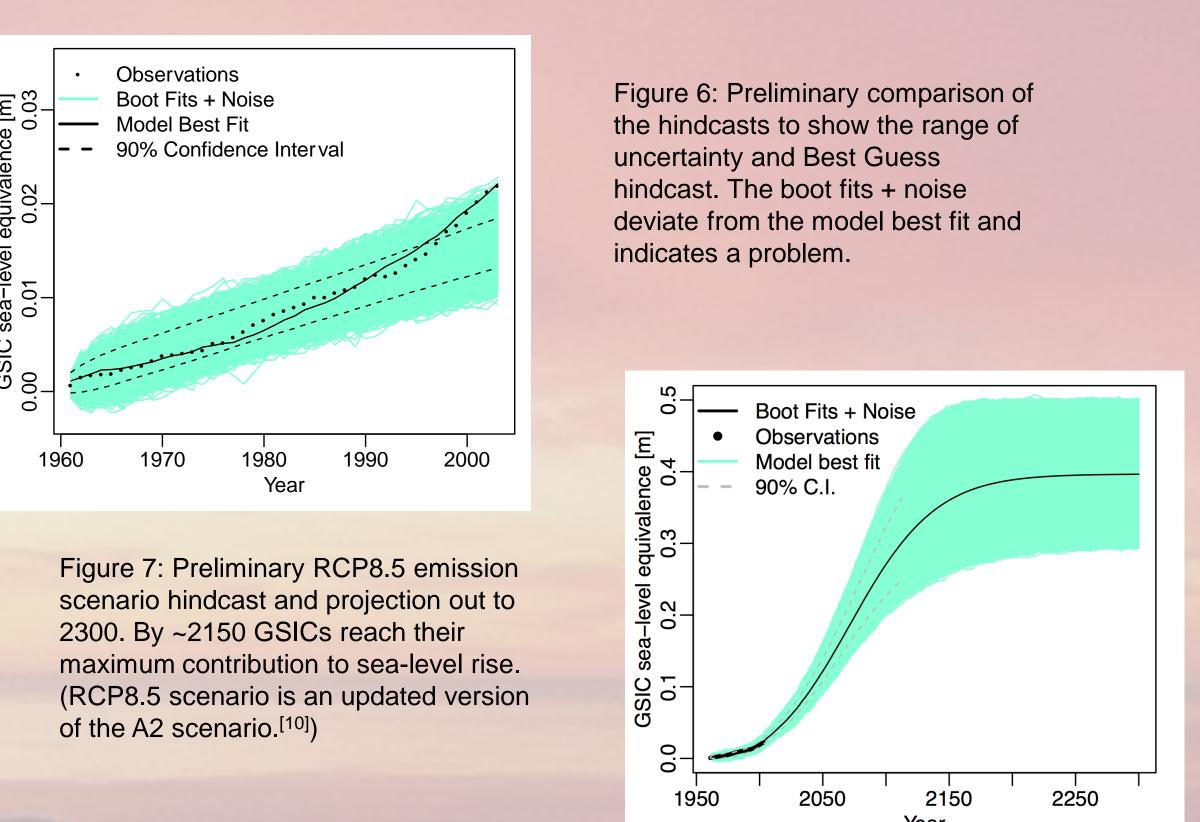


Figure 4: Probabilistic estimate of each parameter by resampling the Differential Evolution best fit parameters.



- an increase in sea level rise an it decreases.
- code.
- United Kingdom and New York, NY, USA.
- Geophysical Research Letters. Vol 32. L05704
- and seasonally resolved temperature changes. Nature. 391. 474-476.
- Continuous Spaces, Journal of Global Optimization, 11:4, 341-359.
- http://www.ncdc.noaa.gov/cag/time-series/global
- http://nsidc.org/forms/G10002_or.html?major_version=1
- greenhouse gas emissions. Climatic Change 109:33-57
- [11] Sunset Over Franz Josef Glacier New Zealand [Wallpaper].2009.

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RESULTS

DISCUSSION/CONCLUSIONS

The GSIC melt model is mainly sensitive to changes in B_0 and V_0 which cause an increase in sea level rise as they increase and n which causes

Bootstrapping the estimation allows us to assess not just the best estimate of the parameter values, but also the associate d uncertainties. These results are preliminary and still need to be carefully tested. For example, we see evidence that the bootstrap method, as implemented, has a problem as the hindcast is biased and the uncertainties seem too wide. This is a work in progress. We hypothesize that this is a bug in our

REFERENCES

[1] Church, J.A., P.U. Clark, A. Cazenave, J.M. Gregory, S. Jevrejeva, A. Levermann, M.A. Merrifield, G.A. Milne, R.S. Nerem, P.D. Nunn, A.J. Payne, W.T. Pfeffer, D. Stammer and A.S. Unnikrishnan, 2013: Sea Level Change. In: Climate Change 2013: The Physical Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Stocker, T.F., D. Qin, G.-K. Plattner, M. Tignor, S.K.

Allen, J. Boschung, A. Nauels, Y. Xia, V. Bex and P.M. Midgley (eds.)]. Cambridge University Press, Cambridge,

[2] Wigley, T. M. L., and S. C. B. Raper. 2005. Extended Scenarios for glacier melt due to anthropogenic forcing.

[3] Gregory, J. M., and J. Oerlemans .1998. Simulated future sea level rise due to glacier melt based on regionally

[4] Van der Wal, R. S. W., and M. Wild. 2001. Modelling the response of glaciers to climate change by applying volume area scaling in combination with a high-resolution GCM. Clim. Dyn. 18. 359–366.

[5] Storn, R. and Price, K. 1997 Differential Evolution – A Simple and Efficient Heuristic for Global Optimization over

[6] Efron, B. and Tibshirani, R. 1993. An Introduction to the Bootstrap. Chapman & Hall.

[7] Climate at a Glance. National Climatic Data Center National Oceanic and Atmospheric Administration.

[8] Occasional Paper No. 58. 2005. Institute of Arctic and Alpine Research University of Colorado.

[9] Intergovernmental Panel on Climate Change. 2000. Summary for Policy Makers Emissions Scenarios. [10] Riahi, K., S. Rao, V. Krey, C. Cho, V. Chirkov, ... P.Rafaj. 2011. RCP 8.5- A scenario of comparatively high

http://www.hdwallpapers.in/sunset_over_franz_josef_glacier_new_zealand-wallpapers.html

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