## Supplementary Information for GIR: a generalised impulse-response model for climate uncertainty and future scenario exploration

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## 1 RCP simulations

Since the RCPs themselves are *concentration* pathways, here we focus on the results of running GIR with concentrations from the RCP database, as was done in the GCMs in CMIP5. We include emissions-driven runs in the Supplementary Information, but these are not fully comparable due to the lack of integration between RCP concentrations and emissions as was planned in Moss et al. (2010). Figure 5 shows diagnosed emissions that are compatible with the RCP concentration series in GIR, run with default parameters plus uncertainties as described above. Here we see the large discrepancies between GIR compatible emissions and the RCP database emissions for  $CH_4$  and  $N_2O$ , while GIR diagnosed emissions agree well with bottom-up emission estimates to the present-day as expected, since GIR is tuned against a very similar historical concentration series; though here we run GIR with the  $CH_4$   $r_0$  parameter tuned to the Global Methane Budget (Saunois et al., 2019), hence the slight discrepancy between GIR and PRIMAP-histTP. Figure 6 shows the corresponding radiative forcings compared to those from the RCP database. The updated simple RF formulae described in Etminan et al. (2016) increase the RF of  $CH_4$  and  $CO_2$ , while marginally decreasing the  $N_2O$  RF. Figure 7 shows historical and future global mean surface temperature ranges (GMST) under the RCP scenarios diagnosed by GIR, alongside historical observed data and future projections from CMIP5. We see that the GIR diagnosed temperatures closely resemble the observed GMST series, but are lower than the CMIP5 ranges. This is due to the default climate response parameter selection in GIR as described above and is discussed in Richardson et al. (2016).

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**Table 1.** Units used in GIR when the default parameter set is used for each gas or aerosol species. Default forcing unit for all species is  $Wm^{-2}$ .

Variable	$CO_2$	CH <sub>4</sub>	$N_2O$	SOx	NOx	ВС	OC	NH3	VOC	All other WMGHGs
Emissions	GtC	MtCH <sub>4</sub>	MtN <sub>2</sub> O-N <sub>2</sub>	MtSO-2	MtN	MtC	MtC	Mt	Mt	Mt
Concentrations	ppm	ppb	ppb	-	-	-	-	-	-	ppb

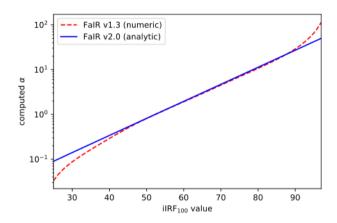


Figure 1. Numerical solution for  $\alpha$  in FaIR v1.0 and v1.3 versus analytic solution in GIR. This highlights the reason for the lowered  $r_0$  parameter in GIR compared to FaIR v1.3 (Smith et al., 2017) or v1.0 (Millar et al., 2017), as we see that for pre-industrial  $\alpha$  values (0.12 in v1.0 or 0.16 in v1.3), the analytic solution requires a lower iIRF<sub>100</sub> than the numeric. Despite the marginal difference in shape, we do not find GIR has reduced ability in reproducing historical concentration when compared with FaIR v1.0 or v1.3.

Code and data availability. The model and code used to produce the figures is publicly available at https://github.com/njleach/FaIR\_v2-0, and will be cleaned up and release ready prior to acceptance. All data used in this study is publicly available at the relevant cited sources.

20 Author contributions. NJL, SJ and MRA conceived the study. NJL and SJ wrote the model code, and BW helped tune model parameters. JT provided CMIP6 response parameters. JL and MC advised on model uses and tested the model. NJL, CJS, ZN, JL and MRA wrote the manuscript.

Competing interests. We declare that we have no competing interests.

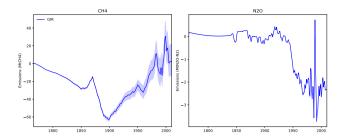
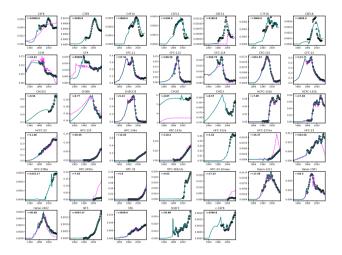
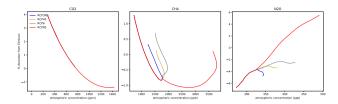


Figure 2. Differences between historical diagnosed emissions in GIR and the RCP database emissions for  $CH_4$  and  $N_2O$ . This displays high similarity to the imposed natural emissions in FaIR v1.3 (Figure 2 from Smith et al. (2017)), demonstrating that  $CH_4$  and  $N_2O$  cycles in GIR and FaIR v1.3 are not systematically different.

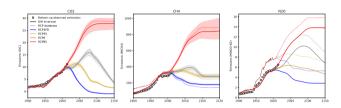


**Figure 3.** Best-estimate annual emissions from a more complex atmospheric model inversion Rigby et al. (2014), GIR inverse emissions and RCP database emissions. Open black circles show inverse emissions from a 12-box model Cunnold et al. (1994); solid green lines show inverse emissions from GIR with tuned parameters and solid pink lines show emissions from the RCP database. Inset text shows the tuned species lifetime.

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**Figure 4.** Corresponding deviations of FaIR v2.0 from Etminan et al. (2016) formulae for the RCPs. These are plotted as % deviations against gas concentration (so absolute deviations are considerably lower at low concentrations than high concentrations).



**Figure 5.** Diagnosed emissions corresponding to the RCPs. Solid lines show best-estimate GIR diagnosed inverse emissions, with associated 5-95% plumes. Dotted lines show emissions directly from the RCP database. Unfilled black circles show bottom-up emission estimates from GCP and PRIMAP-histTP (Ouéré et al., 2018; Gütschow et al., 2016), smoothed with a 5-yearly running mean.

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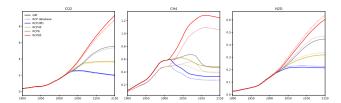
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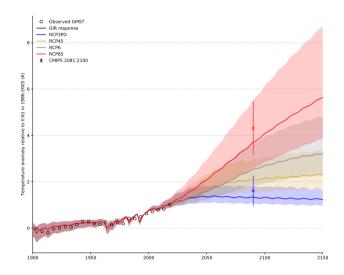
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**Figure 6.** Radiative forcing as computed from the RCP database concentrations using GIR, and corresponding forcings diagnosed by MAG-ICC from the database.



**Figure 7.** RCP temperature anomalies computed by GIR, and ranges from the CMIP5 ensemble. Solid lines and plumes show best-estimate and 5-95% range temperature anomaly simulated in GIR. Unfilled black circles show observed GMST as the mean of 4 temperature datasets (Vose et al., 2012; Cowtan and Way, 2014; Lenssen et al., 2019; Morice et al., 2011), smoothed with a 5-yearly running mean. Error bars in 2090 show CMIP5 projected 2081:2100 mean temperature anomalies (Collins et al., 2013).

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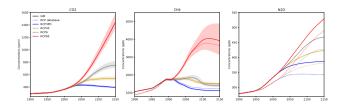


Figure 8. Simulated concentrations when RCP database emissions and other forcings drive GIR, compared to original RCPs.

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