



- 1 Impact Study of Increased Radio Occultation Observations during the
- 2 ROMEX Period Using JEDI and the GFS Atmospheric Model
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- 8 Abstract. The international collaborative Radio Occultation Modeling Experiment (ROMEX)
- 9 project marks the first time using a large volume of real data to assess the impact of increased
- 10 Global Navigation Satellite System (GNSS) radio occultation (RO) observations beyond
- 11 current operational levels, moving past previous theoretical simulation-based studies. The
- 12 ROMEX project enabled the use of approximately 35,000 RO profiles- nearly triple the
- 13 number typically available to operational centers, which is about 8,000 to 12,000 per day. This
- 14 study investigates the impact of increased RO profiles on numerical weather prediction (NWP)
- 15 with the Joint Effort for Data assimilation Integration (JEDI) and the global forecast system
- 16 (GFS), as part of the ROMEX effort. A series of experiments were conducted assimilating
- 17 varying amounts of RO data along with a common set of other key observations. The results
- 18 confirm that assimilating additional RO data further improves forecasts across all major
- 19 meteorological fields, including temperature, humidity, geopotential height, and wind speed,
- 20 for most of vertical levels. These improvements are significantly evident in verification against
- 21 both critical observations and the European Center for Medium-Range Weather Forecasts
- 22 (ECMWF) analyses, with beneficial impacts lasting up to five days. Conversely, withholding
- 23 RO data resulted in forecast degradations. The results also suggest that forecast improvements
- scale approximately logarithmically with the number of assimilated profiles, and no evidence
- 25 of saturation was observed. Biases in the forecast of temperature and geopotential height over
- 26 the lower stratosphere are discussed, and they are consistent with findings from other studies
- in the ROMEX community.





1 Introduction

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30 Radio occultation (RO) is an active remote sensing technique that measures the refraction of signals transmitted by Global Navigation Satellite System (GNSS) and received by 31 32 instruments aboard low-Earth orbit (LEO) satellites. The pioneering GPS/Meteorology 33 (GPS/MET) mission demonstrated that the RO technique can effectively probe the Earth's 34 atmosphere, providing profiles with high vertical resolution and accuracy (Kursinski et al., 35 1997). Since then, the number of RO profiles has increased with the expansion of GNSS 36 beyond GPS (e.g, GLONASS, Galileo, and BeiDou), along with the deployment of more RO 37 receivers aboard new LEO missions, such as the U.S./Taiwan FORMOSAT-3/Constellation Observing System for Meteorology, Ionosphere, and Climate (COSMIC-1; launched in 2006) 38 39 and its successor FORMOSAT-7/COSMIC-2 (COSMIC-2, launched in 2019), the European 40 Space Agency (ESA) MetOp series (MetOp-A, 2006; MetOp-B, 2012; MetOp-C, 2018), and 41 the ESA/US Sentinel-6 (launched in 2020). 42 RO data are considered as one of the most impactful observation types in terms of their 43 contribution to the forecast skills in numerical weather prediction (NWP). The positive impact 44 of RO observations on NWP analysis and forecast has been well-documented by numerous 45 NWP centers (Healy and Thépaut, 2006; Bowler 2020; Ruston and Healy 2021; Cucurull 2023). Unlike satellite radiance data, RO observations are inherently unaffected by clouds or 46 47 precipitation and therefore their assimilation in NWP requires no bias correction. Instead, RO 48 observations serve as a reference to anchor the bias correction of satellite radiance data (Healy 49 et al. 2008; Bauer et al. 2014). 50 Since launched in 2019, COSMIC-2 has been steadily providing approximately 6,000 RO profiles daily, primarily between 45°S and 45°N. Other government missions in polar orbits 51 52 contribute around 2,000 daily profiles globally. More recently, the emergence of commercial 53 RO providers, such as GeoOptics Inc. (Pasadena, CA, USA), PlanetiQ (Golden, CO, USA) and Spire Global Inc. (Boulder, CO, USA), has further expanded RO data availability. These 54 55 commercial sources supplement operational capabilities with additional profiles, depending on 56 purchase agreements. With this expanded data volume, NWP centers have recently been

enabled to explore the impact of assimilating slightly more than 10,000 RO profiles per day.





Several NWP centers have demonstrated that the relative impact of RO data in NWP has grown alongside the increasing availability of profiles. Bowler (2020) at the Met Office assessed the RO data produced by Spire and stated that the forecast impact of increasing the RO data volume is roughly proportional to the logarithm of the total amount of GNSS RO data assimilated. Ruston and Healy (2021) reported a novel finding that COSMIC-2 data improve the tropical tropospheric humidity forecasts in both the Navy Global Environmental Model (NAVGEM) and the European Center for Medium-Range Weather Forecasts (ECMWF) Integrated Forecasting System (IFS). The assimilation of COSMIC-2 and Spire observations was found beneficially in both the ECMWF and Met Office's NWP systems (Lonitz et al. 2021). Cucurrul (2023) demonstrated that COSMIC-2 observations have a significant impact in the forecast improvement of temperature and winds in the tropics.

While these studies demonstrated the valuable impact of increased RO profiles in operational NWP systems, the potential benefits of even larger data volumes were only explored through theoretical simulations. Harnisch et al. (2013) used an ensemble of data assimilations (EDA) approach to evaluate the change of RO data impact as a function of observation numbers. They demonstrated that saturation was not found with 128,000 simulated RO profiles per day. With a global observing system simulation experiment (OSSE) study, Privé et al. (2022) found that the assimilation of 100,000 daily RO profiles did not reach the impact saturation in the hybrid four-dimensional ensemble variational data assimilation system and Global Earth Observing System (GEOS) model.

Clearly, the number of RO observations currently utilized in real-time NWP operations remains significantly below the potential demonstrated in these simulated studies. Meanwhile, a large portion of RO observations from commercial providers is not purchased and remains unassimilated in operational systems, highlighting that the full impact of RO from both government and commercial providers has yet to be fully realized. Since May 2022, the International RO Working Group (IROWG; https://irowg.org), one of the scientific advisory working groups of the Coordination Group for Meteorological Satellites (CGMS), has led an international collaborative effort – the Radio Occultation Modeling Experiment (ROMEX; Anthes et al. 2024) – to explore the impact of RO observations by collecting as many profiles as available from both commercial and government providers during the testing period.





Specifically, ROMEX has collected nearly 35,000 daily profiles during the experimental period (September to November 2022), whereas there are about 12,000 daily profiles available in the real-time NWP operations at the U.S. National Oceanic and Atmospheric Administration (NOAA). ROMEX provides a unique opportunity for both NWP centers and the research community to evaluate impacts of increased RO numbers using large quantities of real RO observations for the first time.

The overarching objective of this study is to demonstrate forecast improvement through the assimilation of the increased RO data. We aim to quantitatively assess these data impacts with respect to operational implementations, while leveraging advanced features for enhanced performance. Specifically, this study utilizes the Joint Effort for Data assimilation Integration (JEDI; Trémolet and Auligné 2020) for data assimilation and the NOAA Global Forecast System (GFS) for forecasting. Given JEDI is the next generation data assimilation system for operations at NOAA, the National Aeronautics and Space Administration (NASA), the U.S. Naval Research Laboratory (NRL), and other NWP centers worldwide, this ROMEX study offers additional benefits by demonstrating JEDI's capabilities and providing insights for ongoing transitions to operations.

This manuscript is organized as follows: Section 2 summarizes ROMEX and the RO observations used for this study; Section 3 introduces the GFS forecast model, the JEDI data assimilation system, and the experimental design; Section 4 presents the evaluation of the ROMEX RO data impact using the JEDI-GFS system; Section 5 presents a summary of the work.

2 ROMEX and GNSS RO observations

ROMEX is an IROWG initiative designed to evaluate the impact of increasing radio occultation (RO) data volume using real observations from both government and commercial missions, extending beyond current operational capabilities. ROMEX involves approximately 30 international agencies and research institutions, including data providers, processing centers, NWP centers, universities, and research institutes. A complete list of ROMEX participants is available on the ROMEX website (https://irowg.org/ro-modeling-experiment-





romex/). The outcomes of ROMEX provide guidance to CGMS for formulating 116 117 recommendations to space agencies on RO mission planning and coordination. Additionally, 118 ROMEX results offer valuable insights for RO data providers, processing centers, and NWP 119 centers to enhance data retrieval techniques and improve the assimilation of RO data in weather 120 forecasting. 121 Through dedicated data agreements with commercial RO providers, the ROMEX effort 122 was able to access data not covered by existing global licenses held by NOAA, NASA, or 123 EUMETSAT from their respective commercial purchases. The ROMEX dataset includes 124 commercial RO data otherwise unavailable to the public, along with publicly available data 125 from sources such as the UCAR COSMIC Data Acquisition and Access Center (CDAAC), NOAA, NASA, and EUMETSAT. Due to the involvement of multiple processing centers and 126 127 data providers, different processing versions of the data were available to support validation 128 and processing studies. For this study, ROMEX data were processed either by UCAR, 129 EUMETSAT, or their original providers. All these data are distributed through the 130 EUMETSAT Radio Occultation Meteorology Satellite Application Facility (ROM SAF). 131 Our objective is to evaluate the impact of an increased volume of RO profiles on analyses 132 and forecasts, rather than comparing the performance or characteristics of various missions. Early data evaluation already shows the quality of these data is relatively comparable for NWP 133 134 applications (Marquardt 2024; Anthes et al. 2025). The available profiles are categorized into 135 two groups based on their sources: base missions (hereafter, BASE) which are governmentsupported missions, and the supplementary missions that are provided by commercial vendors. 136 137 The base missions include COSMIC-2, Metop-B, Metop-C, Kompsat-5, PAZ, Sentinel-6, 138 TerraSAR-X, and TanDEM-X, all of which are available in near-real-time for operational 139 NWP systems. The supplementary missions consist of GeoOptics, PlanetiQ (Kursinski 2025), 140 Spire (Nguyen 2025), Yunyao (Cheng 2025), and Tianmu (Tang 2025). On average, 141 approximately 35,000 profiles (8,000 from base missions and 27,000 from supplementary 142 missions) were available during the ROMEX period (hereafter, ROMEX). To further quantify 143 the impact of the increased profile volume, the EUMETSAT ROM SAF provided a sub-dataset 144 referred to as ROMEX20K, in which the average daily number of profiles is 20,000. In the





145 ROMEX20K sub-dataset, the supplementary profiles are reduced to approximately 12,000 per 146 147 Figure 1 presents the total number of RO profiles in each 5° × 5° latitude-longitude grid for September 2022, the testing period of this study. Specifically, Fig. 1a-c shows the number of 148 BASE profiles, supplementary mission profiles, and all available ROMEX profiles, 149 respectively. Fig. 1d-e displays the supplementary profiles used in the ROMEX20K and all 150 ROMEX20K profiles. Fig. 1f-g shows the ratio of supplementary profiles used in ROMEX20K 151 152 with all supplementary profiles, and the ROMEX20K profiles relative to the total number of 153 ROMEX profiles, respectively. 154 The total number of BASE profiles (Fig. 1a) peaks in the tropics and decreases poleward, 155 primarily due to the dominance of COSMIC-2. The supplementary profiles (Fig. 1b), however, 156 are more evenly distributed across the mid-to-high latitudes. Overall, the combination of all 157 available profiles (Fig. 1c) results in a relatively uniform global distribution geographically. The supplementary profiles used in the ROMEX20K sub-dataset kept a higher portion over the 158 northern hemisphere and southern mid-to-high latitudes than over the tropical regions (Fig. 1f). 159 160 Combined with the base profiles, ROMEX20K has a better coverage over the tropics than other 161 regions of the globe and the fewest profiles over the southern polar regions (Fig. 1g).



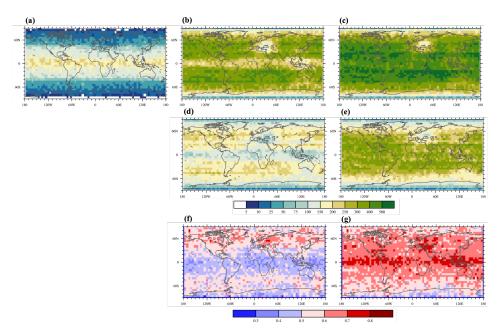


Figure 1: Number of RO profiles in September 2022, re-gridded to a $5^{\circ} \times 5^{\circ}$ latitude–longitude grid, for the following data sets: (a) base missions (BASE), (b) all supplementary missions, (c) all ROMEX missions (ROMEX), (d) supplementary missions in the ROMEX20K, (e) all missions used in the ROMEX20K configuration. Panels (f) and (g) show the ratio of total profile counts between (d) supplementary in ROMEX20K and (b) total supplementary missions, and between (e) ROMEX20K and (c) all ROMEX missions, respectively.

3 Forecast model, data assimilation, and experimental design

3.1 Forecast model

The Global Forecast System (GFS) is NOAA's medium-range operational global weather prediction model, developed and maintained by the National Centers for Environmental Prediction (NCEP). It is part of the Unified Forecast System (UFS), a community-based, coupled Earth modeling framework designed to integrate research and operational weather modeling for more consistent and advanced forecasts (Zhou et al. 2022). This study used the atmospheric forecast model component of UFS, and not the entire suite. Further, the next





planned release version 17 implementation¹ (GFSv17; GFS hereafter) was employed in this study. This latest version of GFS is continuing to use the Finite-Volume Cubed-Sphere (FV3; Lin 2004) dynamical core, and also incorporates significant upgrades in parameterizations for atmospheric processes such as cloud microphysics (Stefanova et al. 2022; Meixner et al. 2023), in comparison to the current operational implementation at NCEP. The global forecasts for this study are configured at a horizontal resolution of approximately 25 km with 127 vertical levels extending up to 80 km (C384L128). This is half of the operational resolution and is standard practice for pre-implementation testing at NCEP and by associated researchers.

3.2 Data assimilation

This study uses the Joint Effort for Data assimilation Integration (JEDI; Trémolet and Auligné 2020) to fulfill the data assimilation component. Led by the Joint Center for Satellite Data Assimilation (JCSDA), JEDI was initiated in 2017. As the project has grown partners now include NOAA, NASA, NRL, the U.S. Air Force, the NSF National Center for Atmospheric Research (NCAR), UK Met Office and developers from universities. JEDI has been interfaced to various models, including the GFS, through the JEDI-FV3 component (https://github.com/JCSDA/fv3-jedi/), allowing the partner agencies using FV3 core-based systems to implement JEDI in real-time applications.

The observation operators for JEDI are developed within an abstracted and generic coding layer known as the Unified Forward Operator (UFO). A generic design makes UFO model-agnostic and allows it to be used in a play-and-plug manner through configuration files. Currently, UFO includes six GNSS RO operators, four for bending angle and two for refractivity, contributed by different partners to replicate the implementation in their respective NWP systems. The four bending angle operators include one based on the operational NCEP Bending Angle Model (NBAM; Cucurull et al. 2013), the Met Office's bending angle operator (Burrows 2014, Burrows et al. 2014), and both one-dimensional (ROPP1D; Healy and Thepaut

¹GFSv17 has not been implemented in the operation as we started this work to the best of our knowledge. We checked out the branch prototype/hr3 in August 2024 from thttps://github.com/ufs-community/ufs-weather-model





2006) and two-dimensional (ROPP2D; Healy et al. 2007; Healy 2014) operators interfaced via the ROM-SAF Radio Occultation Processing Package (ROPP; https://rom-saf.eumetsat.int/ropp), which are used operationally by NRL and ECMWF, respectively. Additionally, two refractivity operators are included, following implementations from the Met Office and NCEP (Cucurull et al. 2007, Buontempo et al. 2008). Most NWP centers use 1D bending angle operators operationally, considering both its impacts and computational efficiency. While a detailed comparison of these operators is performed in a separate effort, we use the ROPP1D operator with the default JEDI configuration that was based on the current implementations by partner agencies. UFO also includes associated quality control (QC) procedures and observation error models, allowing creation of a consistent treatment to those used in operational applications at other centers.

3.2.1 GNSS RO forward operator

Assuming the atmosphere is horizontally homogeneous and spherically symmetric, ROPP1D computes the bending angle, α , by vertically integrating the refractive index from the model background, as shown in Equation 1 (Healy and Thepaut 2006). a is the observed impact parameter, n is the modelled refractive index, and x = nr is the product of the refractive index and the radius value r of a point on the ray path.

$$\alpha(a) = -2a \int_{a}^{\infty} \frac{1}{\sqrt{x^2 - a^2}} \frac{d \ln(n)}{dx} dx \tag{1}$$

The model background information is extracted for each observation point along the RO profile, valid at the horizontal location of its corresponding tangent point. Therefore, the vertical drift of tangent points is fully accounted for. Impact height – defined as the difference between the impact parameter and the local radius of curvature of the Earth – is referred to as the vertical coordinate when presenting RO space results. Note that the impact parameter is a geometric quantity representing the closest distance between the straight-line trajectory of a GNSS signal and the Earth's center; it is the actual coordinate in RO assimilation.

3.2.2 RO observation error and quality control





Observation error and quality control procedure are two crucial parameters in DA. The observation error accounts for measurement uncertainty, representativeness error, and forward operator error (Bormann 2015). Accurate modeling of observation errors is essential for appropriately weighting RO observations relative to other data types and the background error. Meanwhile, QC procedures are closely linked to both the forward operator and observation uncertainty, as they aim to remove observations with large departures that may result from forward operator limitations or various sources of measurement error. As such, observation error characterization, quality control, and the forward operator are tightly interconnected in the assimilation of RO observations.

We applied the observation error model used in the NRL designed system (Ruston and Healy 2021) that is run operationally at Fleet Numerical Meteorology and Oceanography Center (FNMOC). Figure 2 shows the observation errors (in percentage) averaged over a random day, as functions of latitude and impact height. In this scheme, observation errors are defined as a percentage of the observed values and decrease linearly with increasing impact height, reaching a minimum of 1.25 % at the "minimum error height". A damping factor is applied to account for latitudinal variation. In JEDI implementation, the error is specified as 20 % at the surface (impact height is 0) at 0° latitude and is reduced away from the equator following the cosine of latitude. The minimum error height also varies with latitude, decreasing from 12 km at the equator to 5.333 km at the poles – again modulated by the cosine of latitude.

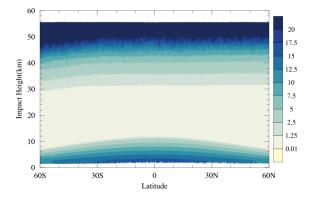


Figure 2: Percentage observation errors (%) for all RO observations on September 1, 2022, using the NRL error model.





3.3 Experimental design

Four sets of experiments—namely, noRO, BASE, ROMEX20K, and ROMEX—were conducted using the JEDI-GFS system over a one-month period in September 2022. All experiments assimilated a common set of conventional observations, cloud-motion vectors, and satellite radiances from the Global Data Assimilation System (GDAS) archive. Specifically, the observations including conventional data from radiosondes, aircraft, and surface stations, as well as scatterometer wind vectors and radiances from AMSU-A and ATMS measurements from multiple satellites available during the study period.

These four experiments differ only in the volume of RO profiles assimilated. The noRO experiment excludes RO data entirely. The BASE experiment assimilates only publicly available RO profiles, totaling approximately 8,000 per day. ROMEX20K and ROMEX assimilate approximately 20,000 and 35,000 daily RO profiles, respectively, based on the corresponding datasets. Differences in forecast skill among these experiments illustrate the impact of enhanced RO data volume available during the ROMEX period.

All experiments were initialized at 0000 UTC on September 1, 2022, using a 6-hour forecast as background from the operational GFS system at NCEP. The JEDI-GFS system was then cycled every 6 hours, with a 6-hour assimilation window and background fields provided by the previous forecast cycle. Data assimilation was performed using JEDI's hybrid three-dimensional variational (3DVar) method, with 40 ensemble members taken from the NCEP global ensemble forecast system. The data assimilation minimization was performed on a so-called dual-resolution grid: the background and forecasts used the C384 grid, while the minimization was carried out on the C192 grid.

4 Results and evaluation

This section compares the results of all experiments to assess the impact of RO observations on forecast skill. Short-to-medium range forecasts are evaluated against observations and model analyses. In observation space, common evaluation metrics include observation—minus—background (OMB) and observation—minus—analysis (OMA) statistics, whose mean values are





often referred to as background bias or analysis bias, respectively. In model space, forecast skill is assessed by comparing model forecasts and ECMWF analyses (FMA) at analysis grid points. Three basic metrics, root-mean-square error (RMSE), standard deviation (STDV), and mean bias, are calculated for OMB/OMA or FMA over the entire experimental period.

To further evaluate the impact of each experiment relative to the reference (noRO or BASE), two additional metrics are adopted to illustrate the impact of an experiment relative to the reference experiment. The first is the normalized difference of a given metric between the experiment and the chosen reference (Eq. 2), where a negative value indicates improvement and a positive value indicates degradation. The second is the mean absolute error reduction (MAER; Eq. 3), which compares the absolute biases between experiments. A negative MAER reflects a beneficial bias reduction relative to the reference experiment, while a positive value indicates a detrimental increase.

Normalized difference =
$$100\% \times \frac{M(Exp.) - M(Reference)}{M(Reference)}$$
 (2)

MAER = 100% x
$$\frac{|\text{Bias}(\text{Exp.})| - |\text{Bias}(\text{Reference})|}{|\text{Bias}(\text{Reference})|}$$
 (3)

4.1 Evaluation in observation space

Statistics in RO observation space are first calculated to evaluate the performance of the JEDI-GFS system in assimilating the large volume of real RO data from ROMEX. Because RO bending angle observation values span a few orders of magnitudes vertically, the OMB statistics are presented in a normalized format, i.e., OMB/O (B is a 6-h forecast from the previous cycle). Figure 3 shows the RMSE and bias from the ROMEX experiment. RMSE results (Fig. 3a) show that the assimilation produces a reasonable agreement between the bending angle observations and both the 6-h forecast and analysis, with lower RMSE in the analysis than in the background. Biases are also notably reduced after assimilation especially between 5 and 12 km impact heights, when comparing OMA to OMB (Fig. 3b), demonstrating the assimilation's effectiveness in correcting background errors in this key region.





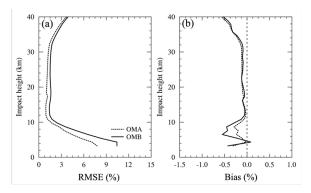


Figure 3: (a) RMSE and (b) Bias of OMB (RO bending angle observation minus 6-h model forecast) normalized by observations (OMB/O; solid) and OMA (RO bending angle observation minus analysis) similarly normalized (OMA/O; dashed), of the ROMEX experiment in September 2022.

Fig. 4 shows the normalized difference in STDV for the fit of temperature, specific humidity, and wind speed forecasts to radiosonde observations. All three experiments show smaller STDV than noRO across all variables and levels. The STDV reduction relative to noRO increases with height, reaching a maximum near 200 hPa for temperature and wind, and at around 700 hPa for humidity. Comparing all three RO experiments (ROMEX, ROMEX20K and BASE), the forecast improvements, or STDV reduction, increase with the growing volume of RO data. ROMEX generates the largest reduction among the three RO experiments. For example, the STDV reduction of temperature forecast at 200 hPa are 3.2%, 5.3%, and 6.8% for BASE, ROMEX20K, and ROMEX respectively. However, the difference between ROMEX and ROMEX20K is negligible in the lower troposphere (below 800 hPa) for temperature (Fig. 4a), and near the surface for wind speed (Fig. 4c). Note that RO data provide only information on the mass fields, wind forecasts are not directly impacted by the assimilation of RO data. Rather, they are impacted through the background error covariance between state variables in DA and the dynamic adjustment through the month-long cycles. Despite this indirect influence, the positive impact of RO data on wind forecasts seen here is very significant.

All RO experiments reduced noRO bias in temperature forecasts between 700 hPa and 100 hPa (Fig. 5). Experiments with additional RO data assimilation tend to cool nearly the entire troposphere, as indicated by the RO experiment curves lying to the right side of the NoRO





curve. In contrast, it warms the air above the jet-stream layer around 200 hPa, where the RO experiment curves shift to the left of the NoRO curve (Fig. 5a). However, the RO experiments produce larger humidity biases compared to noRO. Assimilating more RO data results in a drier atmosphere, while withholding RO data leads to a wetter one (Fig. 5b).

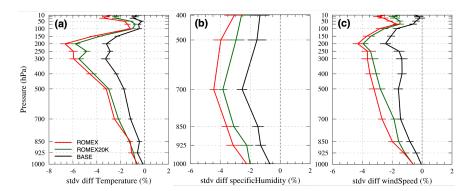


Figure 4: Normalized difference in STDV (%) of the RO experiments relative to noRO for the 6 h model forecasts verified against radiosonde observations of (a) temperature, (b) specific humidity, and (c) wind speed. Overlaid bars are the standard deviations of the normalized STDV difference.

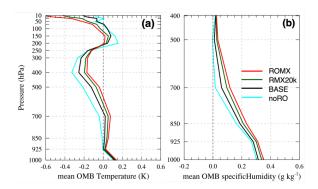


Figure 5: Bias of all experiments for the 6 h model forecasts verified against radiosonde observations (OMB) of (a) temperature (unit: K) and (b) specific humidity (unit: $g \ kg^{-1}$).

The impact of assimilating ROMEX RO data on other critical observing systems is also examined to understand whether the assimilation of RO data can indirectly enhance the assimilation of other observation types. Figure 6 presents the normalized difference in STDV,



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relative to noRO, for the fit to aircraft temperature and wind speed observations. Consistent forecast improvements are observed in both fields, aligning with the verification result from radiosonde data. Figure 7 presents the percentage difference in the number of assimilated observations relative to the BASE experiment for six observation types (radiosonde, aircraft, surface, scatterometer, satellite winds or atmospheric motion vectors, and GNSS RO observations), by comparing ROMEX20K, ROMEX, and noRO. BASE serves as the reference experiment to facilitate the inclusion of GNSS RO data in the comparison. Note also that the three RO experiments assimilate different RO datasets, which does not favor a direct comparison of the total number of assimilated RO observations across experiments. Therefore only the RO observations from the BASE dataset assimilated in all experiments are considered for the bars labeled GNSSRO in Fig. 7. The observations passing quality control in the OMB statistics are counted through the entire month for each type. Both ROMEX20K and ROMEX assimilate more data than BASE across all data types used in this study, whereas noRO assimilates clearly fewer. For example, ROMEX20K and ROMEX assimilate 0.88% and 0.59% more radiosonde observations than BASE, respectively, while noRO assimilates 0.73% less. This indicates that assimilating additional RO data brings the model analysis and shortrange forecast closer to these observations and enables the use of more observational data. This approach clearly shows that both ROMEX and ROMEX20K increased the assimilation of BASE RO data compared to the BASE experiment.

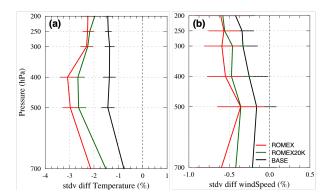


Figure 6: Fractional STDV difference (%) of experiments relative to noRO for the 6 h model forecasts verified against aircraft observations (OMB) of (a) temperature and (b) wind speed. Overlaid bars are the standard deviations of the fractional STDV difference.





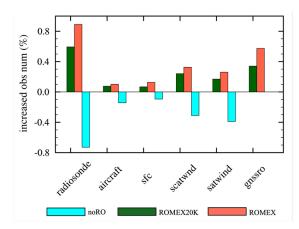


Figure 7: Fractional observation count difference (%) of experiments relative to BASE for key observing types (radiosonde, aircraft, surface, scatterometer, satellite winds or atmospheric motion vectors, and GNSS RO observations from base missions) assimilated in all experiments. Note experiment BASE is used for reference here to account for the GNSS RO observation. Only the base GNSS RO mission observations are counted for the GNSS RO calculation.





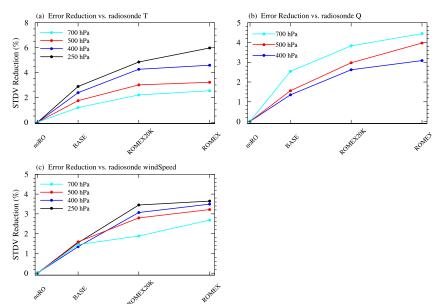


Figure 8: STDV reductions with the number of daily RO profiles for (a) temperature and (c) wind speed at 250-, 400-, 500-, and 700-hPa, and (b) specific humidity at 400-, 500-, and 700-hPa. X-axis is the experiments, and y-axis the fractional STDV difference relative to noRO. For illustration purposes, positive numbers are reductions for this figure.

To consolidate the STDV reductions for the forecast of temperature, specific humidity and wind at various levels, Figure 8 is presented to show the 6-h forecast STDV reductions for the various experiments that exhibit progressive increases in daily RO profiles at key pressure levels, following the NWP verification exchange guidance provided by ROMEX. Using noRO as the benchmark, the STDV reduction increases approximately logarithmically with the growing number of profiles, consistent with the findings of Lonitz (2025). Notably, there is no clear sign of saturation, as most levels continue to show improvement with increasing numbers of RO profiles. However, the degree of this non-saturation appears to depend on both the variable and vertical level and could be influenced by the specific data assimilation configuration.

4.2 Evaluation in model space



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This sub-section evaluates the impact of ROMEX data on short-to-medium range forecasts against the ECMWF analysis over the one-month experimental period. The first assessment focuses on the noRO experiment by comparing its results with that of BASE, which represents the consequences of losing or withholding all RO data. Figure 9 presents the zonal-mean of the STDV and bias of the 6-h temperature forecasts of BASE as a function of pressure levels. Also shown are the differences in STDV and bias between noRO and BASE, with statistical significance at the 95% confidence level. The BASE experiment exhibits large temperature STDVs in the tropical tropopause and lower stratosphere (above 50 hPa), as well as in the lower troposphere, especially in the Southern Hemisphere high latitudes (Fig. 9a). When RO data are withheld (noRO), significant forecast degradations (increasing STDV errors relative to the ECMWF analysis) occur above 850 hPa, with a maximum increase in STDV, about 0.4 K, centered near 50 hPa over the tropics. BASE's temperature bias exhibits multiple patterns, including prominent negative values in the tropical tropopause, positive values above and below the tropopause in the low latitudes, and negative values in the lower tropospheric at high latitudes (Fig. 9b). noRO amplifies BASE's existing biases: negative values become more negative, and positive values become more positive, particularly in the mid-to-upper tropical troposphere and near the mid-latitude tropopause. The negative impact of excluding GNSS RO observation as shown in noRO is also observed in the verification of other key parameters such as humidity and wind speed (not shown), therefore the following sections will focus on the presentation of the impact of the two ROMEX datasets relative to BASE.

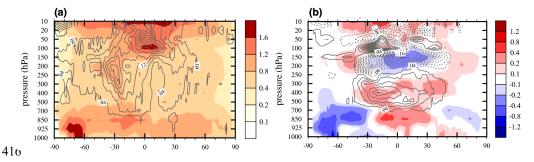


Figure 9: (a) Zonal-mean STDV (shaded), and (b) bias (shaded) of the 6-h temperature forecasts (unit: K) of BASE verified against ECMWF analysis. Overlaid contours are the differences between noRO and BASE (noRO–BASE; unit: K; interval: 0.04 K) in (a) STDV and (b) bias at 95% significance level. Solid/dashed curves represent positive/negative values respectively in both panels. In panel (a), positive

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values for contours indicate noRO forecast degradation relative to BASE, while negative values indicate improvement. In panel (b), opposite signs between the contours and shading indicate improvement in noRO relative to BASE, while matching signs indicate degradation.

Figure 10 compares the two ROMEX experiments, ROMEX and ROMEX20K, with respect to BASE in terms of STDV (Figs. 10a, c, and e) and MAER (Figs. 10b, d, and f) for 6h temperature forecasts, and also includes a direct comparison between ROMEX and ROMEX20K. As introduced earlier, negative MAER values indicate reductions in absolute bias relative to BASE (i.e., improvement), while positive values indicate increased bias (i.e., degradation). Hashed areas indicate regions where the results are statistically significant at the 95% confidence level. Overall, both ROMEX and ROMEX20K exhibit significant STDV reductions relative to BASE between 850 and 50 hPa. Substantial forecast improvements are observed in the Southern Hemisphere's mid-to-upper troposphere, the tropical tropopause region, and the middle troposphere and stratosphere of the Northern Hemisphere's high latitudes. One exception is that ROMEX produces increased STDVs above 50 hPa over the Southern Hemisphere middle latitudes (Figs. 10a and c), which is linked to the warming effect introduced by the additional ROMEX RO data as seen in the verification against radiosonde temperatures (Fig. 5a). The non-significant degradations over the southern hemisphere at around 950 hPa are likely caused by the terrain height mismatch between the forecasts and the ECMWF analyses. ROMEX outperforms ROMEX20K except for the regions of above 50 hPa over the Southern Hemisphere middle latitudes (Fig. 10e). This is consistent with the detrimental effect of increasing RO data as discussed earlier. Meanwhile, ROMEX and ROMEX20K demonstrate beneficial effects of assimilating additional RO data by showing overall reduced MAER values below 50 hPa, particularly in the tropical tropopause and the middle troposphere at low latitudes. On the other hand, both experiments exacerbate biases in the stratosphere above 50 hPa (Figs. 10b and 10d), which is again attributed to the warming effect introduced by the assimilation of ROMEX observations. Overall, assimilating additional ROMEX RO data improves short-range temperature forecasts in terms of both STDV and MAER when verified against ECMWF analyses, with the exception of the lower stratosphere.





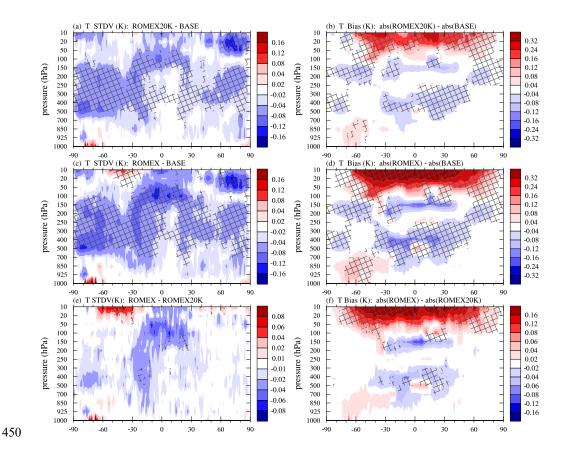


Figure 10: Differences in STDV (a) and (c) and MAER (b) and (d) of the two ROMEX experiments relative to BASE for the 6-h temperature forecasts (unit: K) verified against ECMWF analysis, and difference in STDV (e) and MAER (f) of ROMEX relative to ROMEX20K. Hashed areas overlaid indicate regions of 95% statistical significance level. Positive/negative values of the MAER differences indicate the experiment is farther/closer to the ECMWF analysis.

Figure 11 presents the STDV and bias of the 6-h specific humidity forecasts of the BASE experiment, verified against the ECMWF analysis (Figs. 11a–b), along with the differences in STDV and MAER between both the ROMEX and ROMEX20K experiments and BASE (Figs. 11c–f), and between ROMEX and ROMEX20K (Figs. 11g–h). The STDVs of BASE are largest in the tropical 850 hPa level and decrease gradually with height and toward the poles (Fig. 11a). Both ROMEX experiments show reduced STDV extending from the surface to 300

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hPa (Figs. 11c and e), indicating improved forecast performance. Notably, ROMEX yields additional STDV reductions compared to ROMEX20K in the low-to-mid troposphere over the tropics (Fig. 11g). In terms of bias, BASE exhibits a three-layer structure in the tropics, with positive bias near 700 hPa, negative bias around 900 hPa, and another positive bias near the surface (Fig. 11b). Both ROMEX experiments mitigate the positive bias above 700 hPa (Figs. 11d and f), while ROMEX achieves further bias reduction in the 700–400 hPa layer, particularly over tropical regions (Fig. 11h). Overall, both ROMEX and ROMEX20K outperform BASE in terms of humidity forecast skill, with improvements in both STDV and bias, and ROMEX demonstrating an additional advantage over ROMEX20K.





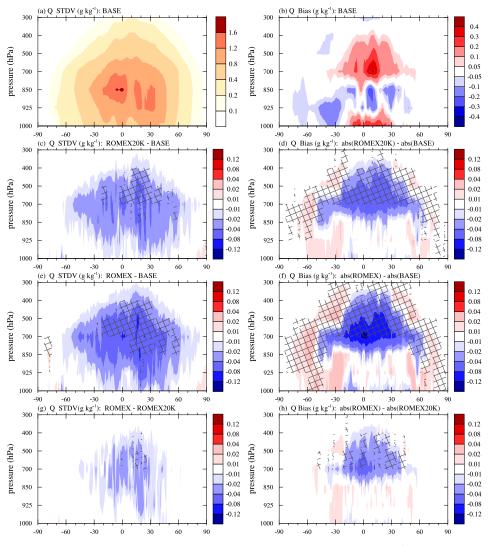


Figure 11: (a) STDV and (b) bias of BASE 6-h specific humidity forecasts (unit: g kg⁻¹) verified against ECMWF analysis, and differences in STDV (c) and (e) and MAER (d) and (f) of the two ROMEX experiments relative to BASE, and difference in STDV (g) and MAER (h) of ROMEX relative to ROMEX20K. Hashed areas overlaid indicate regions of 95% statistical significance level. Positive/negative values of such differences indicate the experiment is farther/closer to the ECMWF analysis.





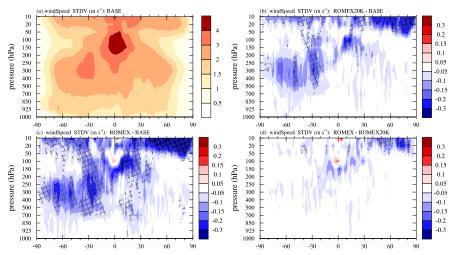


Figure 12: (a) STDV of BASE 6-h wind speed forecasts (unit: m s⁻¹) verified against ECMWF analysis. Differences in STDV for the two ROMEX experiments relative to BASE (b) and (c); and (d) ROMEX relative to ROMEX20K. Hashed areas overlaid indicate regions of 95% statistical significance level. Positive/negative values of such differences indicate the experiment is farther/closer to the ECMWF analysis.

The same diagnostics were also applied to wind speed. Figure 12 displays the 6-h wind speed forecast results: the STDV from BASE (Fig. 12a), the STDV differences between each ROMEX experiment and BASE (Figs. 12b–c), and between the two ROMEX experiments (Fig. 12d). The areas of the largest improvement of wind speed are primarily over the Northern Hemisphere lower stratosphere, the tropical tropopause, and the Southern Hemisphere middle troposphere. Systematic biases are not observed in the wind field and are therefore not presented.

The impact of ROMEX RO assimilation on medium-range forecasts of the JEDI-GFS system is further assessed. Five-day (120-h) forecasts, initiated at each 00Z cycle during September 2022, are examined at 24-h intervals for both experiments. Similar to the short-range evaluation, STDV is calculated against the ECMWF analysis and forecasts of various lead times. The STDV difference with the ECMWF verification is calculated as functions of forecast lead time. Evaluations are conducted in three regions, i.e., the Northern Hemisphere



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502 (NHX, 20°N–80°N), the Tropics (TRO, 20°S–20°N), and the Southern Hemisphere (SHX, 503 20°S–80°S).

504 Figures 13-15 illustrate the differences in STDV between ROMEX20K and BASE and between ROMEX and ROMEX20K, as a function of forecast lead time, for temperature, 505 506 specific humidity, wind speed and geopotential height, respectively. ROMEX20K exhibits 507 reduced STDV, or improved forecast skill, across all three regions throughout the atmosphere 508 above the surface for the temperature 0-5 day forecasts. These reductions persist through the 509 5-day forecast period, while decaying with lead time (Figs. 13a-c). The most beneficial impacts 510 in the TRO and NHX regions are around 150 hPa, with larger than 8 % improvement at the 511 initial forecast time, and are less than 1 % at day 5 (Figs. 13a-b). In the SHX region, larger 512 than 10 % improvement is observed in a broad layer between 200 and 500 hPa at the initial 513 time, which decays to 2-3% at day 5 (Fig. 13c). With additional data assimilated, ROMEX 514 leads to further improvement in temperature forecast over the lower-to-upper troposphere, 515 lasting up to 3 to 5 days (Figs. 13d-f). The detrimental impacts across the three regions are primarily limited to 20 or 50 hPa with relatively small positive STDV differences toward longer 516 517 forecast hours for TRO and NHX, and slighter larger values above 50 hPa for SHX than the 518 other two regions. It also shows that the near surface forecast in SHX does not gain benefit 519 from the additional RO assimilation, and the benefit gained above the surface only sustains in 520 the first 3–4 days (Fig. 13f).

ROMEX's degradation relative to ROMEX20K in the upper levels of SHX extends slightly downward with time. This is closely linked to a warming bias in the lower stratosphere over the mid-latitudes of the Southern Hemisphere, which is introduced by the assimilation of additional RO data (also shown in Figs. 10d, 10f, and 5a).





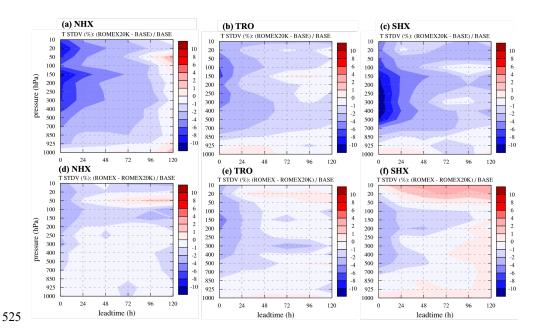


Figure 13: Difference in STDV for temperature forecast of (a-c) ROMEX20K relative to BASE, and (d-f) ROMEX relative to ROMEX20K, verified against ECMWF analysis as a function of forecast lead time for region of (a) and (d) NHX (20°N–80°N, (b) and (e) TRO (20°S–20°N), and (c) and (f) SHX (20°S–80°S).

Similar to the temperature forecasts, the positive impact of assimilating additional ROMEX data on specific humidity forecasts is sustained through five days, with the greatest improvement at the initial time that diminishes rapidly with lead time (Figs. 14a–c). For example, forecast improvements around 500 hPa at the 0-h lead time exceed 6% in all three regions, but decrease to less than 1% at 96-h in NHX and at 72-h in TRO. In SHX, forecast improvements are maintained throughout the troposphere over the 5-day period, with approximately a 2 % reduction in STDV at 120-h. The humidity forecast skill of the increased RO assimilation aligns with that of Prive et al. (2022), in which they stated that the dominant baroclinic process in the SHX winter may account for its longer time scale for improved forecast. As more RO data are assimilated, ROMEX's positive impact on top of ROMEX20K extends to 5 days in NHX and TRO, but only 2 days in SHX. The relative degradation starts at the surface and propagates upward from day 0 onward (Fig. 14f).



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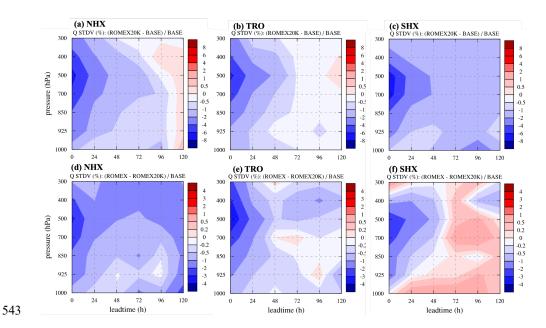


Figure 14: Same as Fig. 13, but for specific humidity forecast.

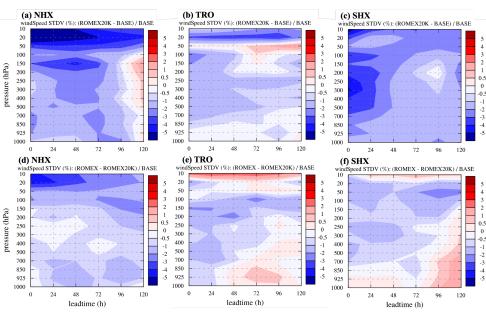


Figure 15: Same as Fig. 13, but for wind speed forecast.





The impacts of assimilating increased RO data on wind forecast are overall positive from ROMEX20K results for the three regions (Figs. 15a-c). Larger forecast improvements, such as more than 5% STDV reductions in the first three days for NHX, are seen in the lower stratosphere for all regions. Forecast degradations are only noticed after 96-h below 50 hPa (Fig. 15a). TRO are mostly positive as well, except slightly increasing negative impacts around 50 hPa beyond 48 h and forward (Fig. 15b). SHX shows all positive impacts from ROMEX20K. Unlike NHX and TRO with the largest positive impacts at upper levels, SHX wind improvements from ROMEX20K are presented for all levels, including 4% improvement between 600–1000hPa for the first 24 h. Also noted is that the impacts of RO data assimilation do not diminish as rapidly as was observed in the temperature and humidity forecasts. For example, the maximum improvement in the NHX troposphere is reached at the 48-h lead time (Fig. 15a).

5. Summary and discussion

This study investigates the impact of increased RO profiles as part of the international collaborative ROMEX project. The current RO profiles available to operational centers are about 8,000–12,000 daily depending on the volume purchased from commercial providers. Earlier studies demonstrated that saturation was not reached with even 128,000 daily profiles. For the first time, the ROMEX project enabled the use of approximately 35,000 RO profiles to explore this further with Observing System Experiments (OSEs).

As part of the ROMEX NWP efforts, this study contributes to building consensus on the impact of increased volumes of RO observations and to addressing the risks associated with potential loss of RO capabilities across NWP centers, specifically within the GFS framework – NOAA's operational forecasting system. At the same time, the study leverages advanced features of JEDI to enhance performance, serving as a valuable platform to evaluate JEDI as the next-generation data assimilation system.

Four sets of experiments were conducted over a one-month period in September 2022: noRO, BASE, ROMEX20K, and ROMEX. All experiments assimilated a common set of





conventional observations, cloud-motion vectors, and satellite radiances, differing only in the amount of RO data assimilated. The BASE experiment assimilated only the publicly available RO profiles (~8,000 per day), while noRO excluded RO entirely. ROMEX20K and ROMEX assimilated approximately 20,000 and 35,000 daily RO profiles, respectively. The actual number of RO profiles per day varies depending on quality control procedures.

The results show that assimilating additional RO profiles significantly improves forecast skill for all key meteorological fields, including temperature, humidity, geopotential height, and wind speed, for most of vertical levels. Forecast improvements were evident in verification against both the critical observations and ECMWF analyses, with impacts lasting up to 5 days (maximum forecast range in the experiments). For example, the STDV reduction of temperature 6h forecasts at 200 hPa, relative to noRO, was 5.3% for ROMEX20K and 6.8% for ROMEX when verified against radiosonde observations. Conversely, withholding RO data led to forecast degradations, with a maximum STDV increase of approximately 0.4 K near 50 hPa over the tropics. The results also suggest that forecast improvements scale approximately logarithmically with the number of assimilated profiles, and no evidence of saturation. These results were achieved without any additional tuning of the data assimilation system. All quality control procedures and observation error specifications for RO data used the default, generic configurations implemented in JEDI for testing purposes. The positive outcomes therefore underscore the consistency and robustness of the RO data quality, and demonstrate that assimilating a large volume of RO observations is both feasible and beneficial.

However, this effort also revealed areas requiring further investigation. In particular, the assimilation of additional RO increases biases in temperature within the lower stratosphere. The ROMEX20K and ROMEX experiments introduced a cooling effect throughout much of the troposphere and a warming effect above 200 hPa, leading to increased forecast biases relative to the ECMWF analysis. Consequently, Figure 16 illustrates the differences in geopotential height STDV between ROMEX20K and BASE, and between ROMEX and ROMEX20K, as a function of forecast lead time over the Northern Hemisphere. The degradation in height forecasts above 50 hPa due to additional RO data assimilation is clearly evident. Below this level, ROMEX shows forecast improvements lasting up to 5 days (Fig.





16a). The ROMEX versus ROMEX20K comparison (Fig. 16b) suggests that while additional RO data improve forecasts below 50 hPa, they may exacerbate degradation above that level.

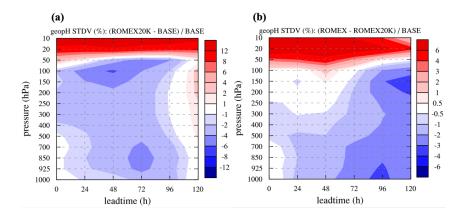


Figure 16: Difference in STDV for geopotential height forecast of (a) ROMEX20K relative to BASE, and (b) ROMEX relative to ROMEX20K, verified against ECMWF analysis as a function of forecast lead time for region of NHX (20°N–80°N).

The sources of these biases are still under investigation. Geopotential height forecast degradation has also been observed by other NWP centers, including the Met Office (Bowler and Lewis 2025) and ECMWF (Lonitz 2025). It is worth noting that the ECMWF analyses used as a reference were produced with the regular volume of RO data assimilated and therefore may not represent the best possible results achievable with the full ROMEX dataset. Additionally, the ECMWF analyses themselves may contain inherent biases, some of which are model-related. Further biases can arise from data processing procedures (e.g. Anthes et al., 2025), and assimilating large volumes of data may amplify such impacts. Moreover, this is the first instance in which the assimilated RO data volume has nearly tripled, and the interactions between RO data and other observations are not yet fully understood.

Ongoing efforts are underway to understand and mitigate these biases. We are actively investigating the role of quality control (QC) and observation errors in the assimilation of ROMEX profiles, especially focusing on whether QC procedures should be regionally adjusted in the upper troposphere and lower stratosphere.



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To conclude, the assimilation of ROMEX RO data has an overall significantly positive impact in the JEDI-based system. Although no saturation was observed even with the full ROMEX data, the 20K subset significantly improves forecast skill, consistent with recommendations from IROWG-10 (Shao et al. 2025) and the second ROMEX workshop. The combination of the ROPP1D forward operator, the NRL observation error model, and generic quality control, within the JEDI framework, not only enabled the successful assimilation of increased data volume in this study, but also lays the groundwork for future exploration and optimization. Author contribution. HZ and HS co-designed the experiments, built the end-to-end system, and jointly prepared the manuscript. HZ conducted the experiments and verifications and created the figures. HS and BR prepared the datasets and provided guidance throughout the project. All authors contributed to the interpretation of the results and the review of the manuscript. Competing interests. None of the authors has any competing interests. Acknowledgments. This work was sponsored by the NSF grant 2054356, NASA grant C22K0658, and the NOAA Science Collaboration Program - Agreement B under grant number NA23OAR4310383B. We thank Drs. Rick Anthes at UCAR and Xuanli Li for their suggestions and discussions which helped improve the work. Authors would like to acknowledge EUMETSAT/the Radio Occultation Meteorology Satellite Application Facility (ROM SAF) to facilitate ROMEX data sharing and processing. We would like to acknowledge high-performance computing support from the Derecho system (doi:10.5065/qx9a-pg09) provided by the NSF NCAR, sponsored by the National Science Foundation. Code and Data Availability Statement. The ROMEX RO profiles are provided through EUMETSAT ROM SAF. Other types of observations are publicly available at the NSF NCAR's Research Data Archive (RDA; https://rda.ucar.edu/datasets/), which contains a subset of the NCEP Global Data Assimilation System observations. The JEDI code of the presented work is available at https://github.com/JCSDA.





- 654 References
- 655 Anthes, R. A., C. Marquardt, B. Ruston, and H. Shao, 2024: Radio Occultation Modeling
- 656 Experiment (ROMEX): Determining the impact of radio occultation observations on
- 657 numerical weather prediction. Bull. Amer. Meteor. Soc., 105, 1552–1568.
- 658 <u>https://doi.org/10.1175/BAMS-D-23-0326.1</u>
- 659 Anthes, R., J. Sjoberg, J. Starr, and Z. Zeng, 2025: Evaluation of biases and uncertainties in
- 660 ROMEX radio occultation observations. EGUsphere [preprint],
- 661 <u>https://doi.org/10.5194/egusphere-2025-2089.</u>
- 662 Bauer, P., G. Radnóti, S. Healy, and C. Cardinali, 2014: GNSS Radio Occultation Constellation
- 663 Observing System Experiments. Mon. Wea. Rev., 142, 555-
- 572, https://doi.org/10.1175/MWR-D-13-00130.1.
- 665 Bormann, N., 2015: Observation errors. Available online at
- https://www.ecmwf.int/sites/default/files/ObsErrors_2015_v2.pdf
- 667 Bowler, N. E., 2020: An assessment of GNSS radio occultation data produced by Spire. Q. J.
- 668 Roy. Meteor. Soc., 146, 3772–3788, https://doi.org/10.1002/qj.3872.
- 669 Bowler, N. E., and O. Lewis: Understanding the impact of additional observations in the Met
- 670 Office system. The Second ROMEX workshop, EUMETSAT headquarter, Darmstadt,
- 671 Germany, 25–27 February 2025. https://cdn.eventsforce.net/files/ef-
- 672 xnn67yq56ylu/website/66/25c71745-2c8d-488b-be58-
- 673 02f274ecd1c0/7 20250225 neillbowler metoffice romex.pdf
- 674 Buontempo, C., A. Jupp, and M. Rennie, 2008: Operational NWP assimilation of GPS radio
- occultation data. *Atmos. Sci. Lett.*, **9**, 129–133. https://doi.org/10.1002/asl.173.
- Burrows, C., 2014: Accounting for the tangent point drift in the assimilation of gpsro data at
- the Met Office. Satellite applications technical memo, 14, Met Office.
- 678 Burrows, C., S. Healy, and I. Culverwell, 2014: Improving the bias characteristics of the ROPP
- 679 refractivity and bending angle operators. Atmospheric Measurement Techniques, 7, 3445–
- 680 3458. http://dx.doi.org/10.5194/amt-7-3445-2014





- 681 Cheng, Y., Improvement of Yunyao RO data quality and development of YunYao
- 682 constellations. The Second ROMEX workshop, EUMETSAT headquarter, Darmstadt,
- 683 Germany, 25–27 February 2025. https://cdn.eventsforce.net/files/ef-
- 684 xnn67yq56ylu/website/66/06d73675-ac8a-4c20-8c20-
- 685 ec976358ccd7/3_the_improvement_of_yunyao_ro_data_quality_and_development_of_yu
- 686 nyao_constellations-chengyan.pdf
- 687 Cucurull, L., and J. C. Derber, R. Treadon, and R. J. Purser, 2007: Assimilation of Global
- Positioning System radio occultation observations into NCEP's Global Data Assimilation
- 689 System. Mon. Wea. Rev., 135, 3174–3193.
- 690 Cucurull, L., J. C. Derber, and R. J. Purser, 2013: A bending angle forward operator for global
- 691 positioning system radio occultation measurements. J. Geophys. Res. Atmos., 118, 14–28.
- 692 Cucurull, L., 2023: Recent Impact of COSMIC-2 with Improved Radio Occultation Data
- Assimilation Algorithms. Wea. Forecast, 38, 1829–1847. https://doi.org/10.1175/WAF-D-
- 694 22-0186.1
- 695 Culverwell, I., H. Lewis, D. Offiler, C. Marquardt, and C. Burrows, 2015: The radio occultation
- 696 processing package, ROPP. Atmos. Meas. Tech., 8, 1887–1899,
- 697 https://doi.org/10.5194/amt-8-1887-2015.
- 698 Harnisch, F., S. Healy, P. Bauer, and S. English, 2013: Scaling of GNSS radio occultation
- 699 impact with observation number using an ensemble of data assimilations. Mon. Wea. Rev.,
- 700 **141**, 4395–4413, https://doi.org/10.1175/MWR-D-13-00098.1.
- 701 Healy, S., 2014: Implementation of the ROPP two-dimensional bending angle observation
- operator in an NWP system. ROM SAF Rep. 19, 33 pp.,
- 703 http://www.romsaf.org/general-documents/rsr/rsr 19.pdf.
- 704 Healy, S. B., Eyre, J. R., Hamrud, M., and Thépaut, J. N., 2007: Assimilating GPS radio
- 705 occultation measurements with two-dimensional bending angle observation operators, Q.
- 706 *J. Roy. Meteor. Soc.*, **133**, 1213–1227. https://doi.org/10.1002/qj.63
- 707 Healy, S. B., and J.-N. Thépaut, 2006: Assimilation experiments with CHAMP GPS radio
- occultation measurements. *Quart. J. Roy. Meteor. Soc.*, **132**, 605–623.





- Kursinski, E. R.: PlanetiQ status and plans. The Second ROMEX workshop, EUMETSAT headquarter, Darmstadt, Germany, 25–27 February 2025. https://cdn.eventsforce.net/files/ef-xnn67yq56ylu/website/66/44ea07c8-4191-4719-ad18-fb08f762f927/4 20250227 robkursinski planetiq romex.pdf
- 709 Kursinski, E. R., G. A. Hajj, J. T. Schofield, R. P. Linfield, and K. R. Hardy, 1997: Observing
- 710 Earth's atmosphere with radio occultation measurements using the global positioning
- 711 system. J. Geophys. Res., **102**, 23429–23465. https://doi.org/10.1029/97JD01569.
- 712 Lin, S., 2004: A "Vertically Lagrangian" Finite-Volume Dynamical Core for Global
- 713 Models. Mon. Wea. Rev., 132, 2293–2307, https://doi.org/10.1175/1520-
- 714 0493(2004)132<2293:AVLFDC>2.0.CO;2.
- 715 Lonitz, K., C. Marquardt, N. Bowler, and S. Healy, 2021: Final technical note of 'impact
- 716 assessment of commercial GNSS-RO data'. ECMWF Tech. Rep.
- 717 4000131086/20/NL/FF/a, 74 pp., https://doi.org/10.21957/wrh6voyyi
- 718 Lonitz, K., Updates on running ROMEX experiments at ECMWF. The Second ROMEX
- 719 workshop, EUMETSAT headquarter, Darmstadt, Germany, 25-27 February 2025.
- 720 https://cdn.eventsforce.net/files/ef-xnn67yq56ylu/website/66/9326ec50-ce3e-47b8-b714-
- 721 24bafb99d8d8/6 new 20250225 katrinlonitz ecmwf romex.pdf
- 722 Marquardt, C., ROMEX data processing. The First ROMEX workshop, EUMETSAT
- headquarter, Darmstadt, Germany, 17–19 April 2024. https://cdn.eventsforce.net/files/ef-
- 724 xnn67yq56ylu/website/61/565d7153-abac-414f-92dc-
- 725 5466867616fc/20240417_13_marquardt_et_al_eumetsat_romex.pdf
- 726 Meixner, J., and coauthors, 2023: Overview of the Next Global Forecast System GFSv17.
- 727 Unified Forecast System Community Workshop, Boulder, CO, July 24–28, 2023.
- https://epic.noaa.gov/wp-content/uploads/2023/08/UIFCW-2023-Tue-2.-
- 729 <u>Meixner_UIFCW-_GFSv17Overview-_202307.pdf</u>
- 730 Nguyen, V.: Spire RO data for ROMEX and Status Update. The Second ROMEX workshop,
- 731 EUMETSAT headquarter, Darmstadt, Germany, 25–27 February 2025.





- https://cdn.eventsforce.net/files/ef-xnn67yq56ylu/website/66/63efe996-5bac-4bf1-
- 733 9919-a982b4165bed/2_20250227_nguyen_vu_spire_romex.pdf
- 734 Privé, N. C., R. M. Errico, and A. E. Akkraoui, 2022: Investigation of the Potential
- 735 Saturation of Information from Global Navigation Satellite System Radio Occultation
- 736 Observations with an Observing System Simulation Experiment. *Mon. Wea. Rev.*, **150**,
- 737 1293–1316, https://doi.org/10.1175/MWR-D-21-0230.1.
- 738 Shao, H., Foelsche, U., Mannucci, A., Azeem, I., Bowler, N., Braun, J., Lonitz, K.,
- 739 Marquardt, C., Steiner, A., Ruston, B. and Vergados, P., 2025: Advances in GNSS-Based
- Remote Sensing for Weather, Climate, and Space Weather: Missions, Applications, and
- 741 Emerging Techniques. Bull. Amer. Meteor. Soc., BAMS-D-25-
- 742 0138.1, https://doi.org/10.1175/BAMS-D-25-0138.1, in press.
- 743 Stefanova, L., J. Meixner, J. Wang, S. Ray, A. Mehra, M. Barlage, L. Bengtsson, P.
- 744 Bhattacharjee, R. Bleck, A. Chawla, B. Green, J. Han, W. Li, X. Li, R. Montuoro, S.
- Moorthi, C. Stan, S. Sun, D. Worthen, F. Yang, W. Zheng, 2022: Description and Results
- 746 from UFS Coupled Prototypes for Future Global, Ensemble and Seasonal Forecasts at
- 747 NCEP, Office note (National Centers for Environmental Prediction (U.S.), 510pp,
- 748 DOI: <u>https://doi.org/10.25923/knxm-kz26</u>.
- 749 Ruston, B., and S. Healy, 2021: Forecast impact of FORMOSAT-7/COSMIC-2 GNSS radio
- 750 occultation measurements. Atmos. Sci. Lett., 22, e1019,
- 751 https://doi.org/10.1002/asl.1019.
- 752 Tang, Q.: Introduction to GNSS RO and GNSS-R products of Tianmu-1 Constellation. The
- 753 Second ROMEX workshop, EUMETSAT headquarter, Darmstadt, Germany, 25-27
- 754 February 2025. https://cdn.eventsforce.net/files/ef-xnn67yq56ylu/website/66/c42f0b73-
- de39-4f34-987c-01bd63625c0a/1 introduction to gnss-ro and gnss-
- r products of tianmu-1 constellation-for romex-20250227.pdf
- 757 Trémolet, Y. and T. Auligné, 2020: The Joint Effort for Data Assimilation Integration
- 758 (JEDI). *JCSDA Quarterly Newsletter*, **66**, 1–5, https://doi.org/10.25923/RB19-0Q26.

https://doi.org/10.5194/egusphere-2025-3235 Preprint. Discussion started: 17 July 2025 © Author(s) 2025. CC BY 4.0 License.





- 759 Zhou, X., Y. Zhu, D. Hou, B. Fu, W. Li, H. Guan, E. Sinsky, W. Kolczynski, X. Xue, Y.
- Luo, J. Peng, B. Yang, V. Tallapragada, and P. Pegion, 2022: The Development of the
- NCEP Global Ensemble Forecast System Version 12. Wea. Forecasting, 37, 1069–1084,
- 762 https://doi.org/10.1175/WAF-D-21-0112.1.