

1 **Sensitivity of model estimates of CME propagation and**
2 **arrival time to inner boundary conditions**

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6 **Key Points:**

- 7 • Multiple fronts detected by STEREO's Heliospheric Imager can be used to describe
8 the longitudinal distortion of a CME.
9 • Close-Sun distortion can be simulated in a reduced physics model using back map-
10 ping techniques.
11 • Relating ensemble time-elongation profile error and arrival time error can be used
12 to estimate CME arrival to within 3 hours.

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Abstract

Accurately forecasting the arrival of coronal mass ejections (CMEs) at Earth is important to enabling mitigation of the associated space weather risks to society. This is only possible with accurate modelling of the event. To do so, we must understand the propagation of a CME through the heliosphere and quantify the performance of models through comparison with spacecraft observations. For the December 12th 2008 Earth-directed CME event, we compute ensembles using the HUXt solar wind model to analyse CME distortion with a structured solar wind and explore hindcast arrival time error (ATE). By highlighting the impact CME shape has on Root-Mean-Square-Error (RMSE) values, we show that time-elongation profiles of fronts captured by the Heliospheric Imager (HI) instruments onboard NASA's STEREO mission match those of the modelled CME nose and flank and can therefore be used to infer details of the longitudinal extent of the CME. We then show that accounting for CME distortion is important to enable accurate estimates of the CME arrival at Earth. This can be achieved by either using observations of multiple features in HI data to infer CME evolution or mapping the solar wind back to a lower inner boundary to allow CMEs to be distorted close to the Sun. For the event studied we show that these approaches resulted in reduced RMSEs of 0.726 and 0.638 with an ATE of one hour and three hours respectively.

Plain Language Summary

Coronal Mass Ejections (CMEs) are giant plasma "bubbles" that erupt from the Sun into space. Upon engulfing Earth, they interact with the near-Earth space environment and result in disruptions to modern electrical infrastructure. Therefore, accurately forecasting CME arrival time at Earth is vital in order to mitigate the risk of space weather. Here, for the December 12th 2008 Earth-directed event, we use a simple-physics solar wind model (HUXt) to explore the distortion that occurs to a CME throughout the journey from the Sun to Earth. Features of the leading edge are tracked from the viewpoint of two spacecraft (STEREO) that are positioned away from the Sun-Earth line and compared to position profiles of bright-light regions pictured by the mission's Heliospheric Imager camera. By running the HUXt model many times, we explore the relationship between the error of the tracked features and the arrival time error of the hindcast. We find that using multiple features to quantify model performance can improve the arrival time prediction, compared to tracking a single feature. Alternatively, we can account for changes to the CME geometry that occur close to the Sun by ejecting a CME into the model earlier.

1 Introduction

Coronal Mass Ejections (CME) are the major driver of severe space weather impacts (Cane & Richardson, 2000; Gosling, 1993). A CME is a cloud of magnetised plasma moving rapidly away from the Sun, at speeds varying from 300 km s^{-1} to 2500 km s^{-1} and containing $10^{12} - 10^{13} \text{ kg}$ of material (Kahler, 1987; Riley & Gosling, 1997). On arrival at Earth, a CME can interact with Earth's magnetosphere, causing short-term changes to the planet's magnetic field, especially if the CME has a southward interplanetary magnetic field component (Dungey, 1961; Chao & Chen, 2001; Arnoldy, 1971). Due to modern civilization's increasing reliance on electrical technologies, society is more vulnerable to the impacts of "space weather". A CME can cause disruption to satellite communications, ground-based power grids and the air travel industry (to name but a few of the detrimental impacts) (Cannon, 2013). This highlights the importance of forecast accuracy for Earth-directed CMEs (Oughton et al., 2019; Hapgood & Thomson, 2018; Möstl et al., 2011; Davies et al., 2012; Barnard et al., 2017).

Currently, operational space weather forecasts typically rely on three-dimensional magnetohydrodynamic (3D MHD) models of the solar wind, with CME initial conditions

63 being estimated by analysis of coronagraph data. Such models, like ENLIL (Odstrcil,
 64 2003), are used by NOAA’s Space Weather Prediction Center (SWPC) and the UK Met
 65 Office’s Space Weather Operation Centre (MOSWOC) to produce deterministic and en-
 66 semble forecasts of CME arrival time at Earth (Mays et al., 2015; Pizzo et al., 2011). Due
 67 to the computing resources needed by full-physics models to run once, the number of en-
 68 semble runs are limited by feasibility (Lee et al., 2013; Cash et al., 2015), with the MOSWOC
 69 computing 24 runs in their forecast (Henley & Pope, 2017). Analysis of these forecasts
 70 reveal a CME mean arrival time error of approximately 13 hours (Riley et al., 2018; Mays
 71 et al., 2015; Vršnak et al., 2014; Iwai et al., 2021). An alternative approach used in re-
 72 cent research (e.g. Barnard et al., 2020; Chi et al., 2021) is to use a computationally ef-
 73 ficient reduced-physics solar wind model (HUXt, Owens, Lang, Barnard, et al., 2020) to
 74 explore parameter space in ensembles with many hundreds to thousands of members. Such
 75 ensemble modelling methods can provide both robust uncertainty estimates and prob-
 76 abilistic forecasts (Barnard et al., 2020; Owens & Riley, 2017). It is anticipated that the
 77 results from these larger ensembles can ultimately be used to inform targeted ensemble
 78 runs using full 3D MHD solar wind models. We here present an analysis of the sensitiv-
 79 ity of the HUXt model results to inner boundary conditions. In particular, we determine
 80 the accuracy with which HUXt can reproduce multiple CME fronts seen in spacecraft
 81 data, aiding our interpretation of such complex observations.

82 NASA’s Solar TERrestrial RELations Observatory (STEREO) mission (Kaiser et
 83 al., 2008), launched in late 2006, was designed to further understand the initiation, struc-
 84 ture, and the propagation of CMEs, especially for Earth directed events. Consisting of
 85 two spacecraft in heliocentric orbits, STEREO-A (STA) travels ahead of Earth while STEREO-
 86 B (STB) lags behind in such a way that they separate from Earth by approximately 22°
 87 of ecliptic longitude annually. Both STEREO spacecraft have the same instrument pack-
 88 ages that take images of the Sun and solar wind, and in-situ measurements of solar wind
 89 plasma and magnetic fields. The remote sensing instrument package SECCHI (Sun–Earth
 90 Connection Coronal and Heliospheric Investigation) can provide information about the
 91 three-dimensional structural evolution of a CME. It includes white-light heliospheric im-
 92 agers (HIs), white-light coronagraphs (COR) and an extreme ultraviolet imager (EUVI).
 93 Together, they observe space from the lower solar atmosphere to beyond 1AU. COR1
 94 and COR2 cover $1.4 - 4$ solar radii (R_\odot) and $2 R_\odot - 15 R_\odot$, respectively, while HI has
 95 a field of view from $15 R_\odot$ to beyond Earth orbit at $215 R_\odot$ (R. A. Howard et al., 2008).
 96 The heliospheric imager is composed of two cameras; HI-1 and HI-2, each with differ-
 97 ent fields of views that allow a CME to be tracked up to 88.7° elongation from the Sun
 98 (R. A. Howard et al., 2008; Eyles et al., 2009; Davis et al., 2009). This mission was the
 99 first to observe CME events from multiple viewpoints outside the Sun-Earth line.

100 The heliospheric imagers rely on Thomson scattering - the process of sunlight be-
 101 ing scattered off free electrons (Billings, 1966; T. A. Howard & Deforest, 2012; Defor-
 102 est et al., 2013; T. A. Howard et al., 2013) to generate their images. Each pixel contains
 103 an integration of the light scattered by plasma along that line of sight. As with conven-
 104 tional photography, an HI image is a two dimensional representation of three dimensional
 105 structure.

106 The first major Earth-directed CME was observed by STEREO on the December
 107 12th 2008, and has since been the subject of a number of studies (Davis et al., 2009; Byrne
 108 et al., 2010; Lugaz et al., 2010; Liu et al., 2010; Deforest et al., 2011, 2012; Manchester
 109 et al., 2017; Zhang et al., 2019; T. A. Howard & Deforest, 2012). At this time, STA and
 110 STB were positioned approximately 42° and -44° from Earth respectively. For STB, this
 111 is a similar viewing geometry to the proposed ESA *Vigil* mission, formerly known as the
 112 *Lagrange* mission (Vourlidas, 2015; Thomas et al., 2018; Akioka et al., 2005), therefore,
 113 this event represents a useful study of CME observations from such an experiment. One
 114 previous study of this event (Davis et al., 2009) assumed the density enhancements seen
 115 in HI were discrete substructures within the CME and disturbed solar wind. By using

116 the forecast front arrival times at L1, the behaviour of the magnetic and plasma mea-
 117 surements were shown to correlate to the properties of the leading edge, magnetic flux
 118 rope, and the core of the CME. Later, by assuming both STEREO spacecraft observed
 119 the same feature, a geometric triangulation technique out to 1AU with prescribed and
 120 fixed spherical (Lugaz et al., 2010) and non-spherical fronts (Liu et al., 2010) showed that
 121 the direction and distance of the CME could be tracked though the HI field of view to
 122 match the L1 in-situ observations. Mapping two-dimensional views seen in HI images
 123 to the complex, three-dimensional structures of CMEs is not straight forward. Barnard
 124 et al. (2017) demonstrated that assuming a CME retained a simple geometric shape as
 125 it propagated through the heliosphere was not consistent with the known uncertainties
 126 of the observations. More recently, Scott et al. (2019) suggested that a single three-dimensional
 127 leading edge could produce multiple regions (i.e. a nose and a flank) of enhanced Thom-
 128 son scatter leading to multiple structures within a two dimensional HI image. For the
 129 December 12th 2008 event it was demonstrated that multiple features seen in HI images
 130 evolved in a way that was consistent with a Kinematically Evolving Flux Rope (Owens
 131 et al., 2006) expanding into a structured solar wind.

132 Here, we use the HUXt model, described below, to make multiple simulations of
 133 the December 12th 2008 CME using varying initial to produce elongation profiles and
 134 hindcasts. Through comparison with the multiple CME fronts observed by HI, we de-
 135 termine which experiment best reproduces the observations and whether good model per-
 136 formance leads to an improved estimate of the CME arrival time at Earth. We test to
 137 see how the inclusion of data assimilated solar wind at the inner boundary, the distance
 138 of the inner boundary and the assumed uncertainty in the initial parameters impact model
 139 output. These will be discussed in more detail in the following section.

140 2 Techniques

141 2.1 Solar Wind Data Assimilation

142 Near-Sun solar wind conditions which serve as the inner boundary conditions to
 143 heliospheric models are typically provided by coronal models constrained by the observed
 144 photospheric magnetic field (Linker et al., 1999; Arge et al., 2004; Holst et al., 2014, e.g.
 145 MAS, WAS, AWSOME). Here, we start with output from the MAS (Magnetohydrody-
 146 namic Algorithm outside a Sphere) model (Linker et al., 1999), but additionally assim-
 147ilate the available in situ solar wind observations to provide a more accurate reconstruc-
 148tion of the ambient solar wind conditions.

149 The Burger Radial Variational Data Assimilation (BRaVDA) solar wind scheme
 150 (Lang & Owens, 2019; Lang et al., 2020) calculates an optimal inner boundary condi-
 151 tion for a steady state simulation of the equatorial solar wind, through the assimilation
 152 of in-situ solar wind observations. A perturbation of the MAS model solution is defined
 153 as the prior state at the scheme’s inner boundary and allowed to propagate out into the
 154 simulated heliosphere using the HUX model. Error between model and observation is
 155 computed with measurements from 1 AU spacecraft. By seeking the minimum cost func-
 156 tion (sum of prior and observation squared errors, weighted by their relative uncertain-
 157 ties) in the scheme, an optimum inner boundary array is defined as the posterior state.
 158 The output array from BRaVDA can then be used to define a solar wind structure at
 159 $30 R_{\odot}$ in a heliospheric models, such as HUXt.

160 In this study, NASA’s Advanced Composition Explorer (ACE) is used to provide
 161 solar wind speed observations. Whilst multi-spacecraft observations are generally desir-
 162 able to improve accuracy, STEREO A and B observations were not used here due to dif-
 163 ferences in solar wind structure that are likely the result of spacecraft heliographic lat-
 164 itude. A fast-stream was measured by STA but not in STB, suggesting large differences
 165 in solar wind structure over relatively small latitudinal ranges about the helio equator,

166 an issue common during solar minimum (Owens, Lang, Riley, et al., 2020). Instead, in-
 167 situ data from ACE between 360° and 90° Carrington longitude (i.e., a full Carrington
 168 rotation excluding the time at which the CME interacted with the spacecraft) was used
 169 in the assimilation. Figure 1 shows time-series of solar wind speed at L1 produced by
 170 HUXt initialised with the non-data assimilated MAS solar wind conditions and the data
 171 assimilated solar wind conditions (BRaVDA), compared to real observations from the
 172 ACE spacecraft. Both model outputs produce peaks at the same periods seen in the ob-
 173 servation, however, BRaVDA is shown to be better at not overestimating the speed or
 174 creating extra peaks. This is most noticeable on December 3rd 2008.

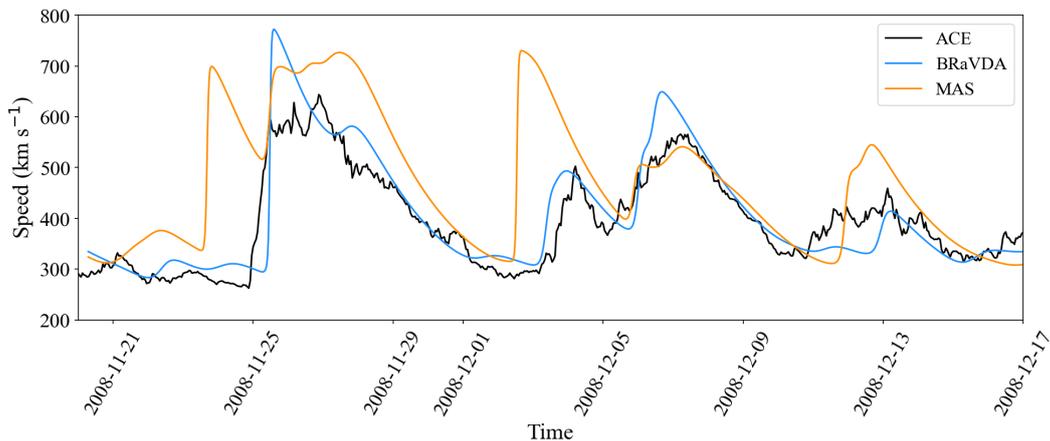


Figure 1. Time series of solar wind speed at L1 for real ACE data (black) and HUXt model output initialised at $30 R_\odot$ with a non-data assimilated solar wind scheme (MAS; orange) and a data assimilated solar wind scheme (BRaVDA; blue). Dates ticks correspond to 00:00 UTC.

175 2.2 HUXt

176 The Heliospheric Upwind eXtrapolation model with time-dependency (HUXt; Owens,
 177 Lang, Barnard, et al., 2020) is a reduced physics numerical model used to simulate he-
 178 liospheric conditions and CME propagation. To do this, the complex magnetic equations
 179 that are found in full-physics 3D models, like Enlil (Odstrcil, 2003), are simplified greatly,
 180 to the assumption that the plasma is a purely radial and behaves as an incompressible
 181 and inviscid hydrodynamic flow. In the steady-state approximation, this approach was
 182 shown to match 3D MHD output (Riley & Lionello, 2011) and has recently been adapted
 183 to allow for time-dependent solar wind structure (Owens, Lockwood, & Barnard, 2020).
 184 Even with the high level of physical approximation, HUXt can replicate the outwards
 185 movement of plasma to beyond 1AU within 5% accuracy of the full-physics models (Owens,
 186 Lockwood, & Barnard, 2020). Due to these simplifications, the time for a single run of
 187 HUXt takes a fraction of a second on your average desktop computer, significantly re-
 188 duced compared with the full-physics models. The computational efficiency of this model
 189 enables the computation of many-member ensembles, as has been done by Barnard et
 190 al. (2020).

191 Within HUXt, CMEs are introduced as a velocity pulse at the inner boundary. The
 192 properties of this pulse are determined with information about the source latitude and
 193 longitude, width, speed, and thickness derived from cone-model fits to coronagraph im-
 194 ages (Millward et al., 2013). Through the thickness parameter required for HUXt setup,
 195 which initialises the radial extent of the inserted CME structure, the subsequent decel-
 196 eration rate of a CME can effectively be changed, essentially by altering the momentum

197 of the CME (Owens, Lockwood, & Barnard, 2020). A CME of zero thickness is a spher-
 198 ical perturbation, whilst larger values produce more of a "sausage" shape. In the ver-
 199 sion of HUXt used in this study (Version 1.0.0), the CME shape is limited to resolution
 200 of the grid cells within the model; longitudinally 2.8° and radially $1.5 R_\odot$. In order to
 201 track the CME disturbance through the model, the edge of a CME is defined as where
 202 the velocity difference per a grid-cell is greater than 20km^{-1} than compared to the
 203 ambient (i.e. no CME) solar wind solution.

204 2.3 Ghost Fronts

205 To bring out dynamic features in HI images, it is common practice to take a running-
 206 difference of the images, wherein the previous image is subtracted from the current one
 207 to reveal finer detail. In doing so, static and slowly varying features are removed and any
 208 changes between HI images appear as bright (increased intensity) and dark (decreased
 209 intensity) features. A bright feature that forms a coherent shape is classified as a front.
 210 Two nearly identical leading CME fronts are frequently observed in such HI images that
 211 are separated by a few degrees of elongation (Scott et al., 2019). Such a feature is also
 212 seen for the December 12th 2008 CME, as shown in figure 2.

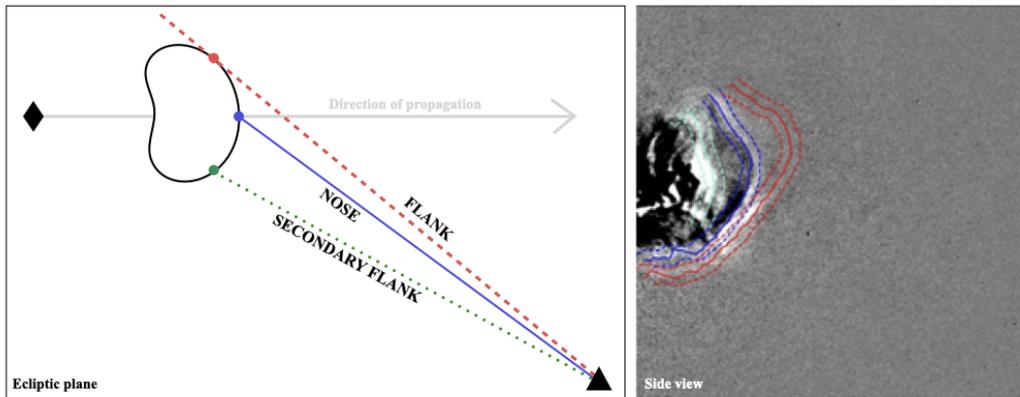


Figure 2. a) Schematic of the CME leading edge features that are detectable in Heliospheric Imager data using the ghost-front theory of Scott et al. (2019). Red dashed line shows the flank (i.e. tangent) of the CME edge, blue solid line shows the nose of the leading edge and pale green dotted line shows the position of a secondary flank. The position of a spacecraft, such as STEREO, is represented by the triangle and the Sun is positioned at the diamond. b) An example of a running-differenced Heliospheric Imager image taken from STEREO-B on December 13th 2008. The Sun is position on the left of the image and the Earth is positioned on the right. The fronts have been highlighted, where solid line is the average and dashed line is the uncertainty. The colour of the front corresponds to the feature colour in the schematic. Red is the outer most front, followed by blue and then pale green closest to the Sun.

213 It has been shown that there can be a correspondence between the position of the
 214 outer front and the pile up of plasma ahead of the CME (Pant et al., 2016). Shock fronts
 215 are produced when the speed of a CME eruption is significantly greater than the am-
 216 bient solar wind, however, there are occasions where a ghost-front is observable but there
 217 is no in-situ evidence of a leading shock. The CME of 12 December 2008 (Scott et al.,
 218 2019) is one such example. It is also possible that multiple CME fronts observed in HI
 219 data represent different structural features of the same CME, for example the upstream
 220 disturbance, the leading edge and the flux rope (Möstl et al., 2011; Davis et al., 2009).

221 More recently Scott et al. (2019) demonstrated that for the December 12th CME, where
 222 the two fronts contained similar latitudinal structure in HI images, the fronts were con-
 223 sistent with the location of the leading edge (or 'nose') and flank of a single CME front.

224 Thomson scattering is vital in being able to understand how CMEs are seen in HI
 225 images. The intensity of scattered light in each pixel of an HI image is proportional to
 226 the density of free electrons integrated along that line of sight as a function of distance
 227 from the observer. Features can therefore appear bright in an HI image for a number of
 228 reasons. For example, the nose of the CME, where the CME front lies along the line of
 229 propagation, is likely to appear bright because of the localised increase in plasma den-
 230 sity. The tangent to the leading edge (i.e. the flank of the CME) can also appear bright
 231 due to the cumulative plasma density integrated along the extended line of sight. Thus
 232 two bright fronts can often be seen in HI images, each resulting from light scattered from
 233 different regions of the same front. This 'ghost front' model has been applied in the work
 234 of Chi et al. (2021) to simulate bright features within CME images. It was shown to be
 235 consistent with observations over a range of latitudes. This suggests that measuring the
 236 elongation difference between ghost fronts can help us determine the longitudinal shape
 237 and/or width of a CME from a single spacecraft. Further, there has been a suggestion
 238 that using both the nose and the flank to optimise HI-1 fitting could improve forecast-
 239 ing abilities of CMEs at L1/ Earth (Barnard et al., 2020; Hinterreiter et al., 2021).

240 3 Case study: December 12th 2008 CME

241 In COR1, the CME was first observed at a radial distance of $7.9 R_{\odot}$ at 10:37 UTC
 242 December 12th 2008. Later the event was also captured by HI-1, first at a radial distance
 243 of $34 R_{\odot}$ at 20:49 UTC and was tracked throughout the instrument's field of view. Many
 244 techniques have been developed to reconstruct CMEs using coronagraph data, such as
 245 the CME Analysis Tool (CAT) used by SWPC (Millward et al., 2013), geometric local-
 246 isation (Koning et al., 2009), and equal masses (Colaninno & Vourlidas, 2009). By av-
 247 eraging these methods, Scott et al. (2019) concluded that this CME propagated along
 248 a solar longitude of $10 \pm 4^{\circ}$ and latitude $9 \pm 2^{\circ}$ (HEEQ coordinates) with a speed of
 249 380 km s^{-1} whilst close to the Sun. Further, CME reconstruction using HI-1 data sug-
 250 gested an increased radial speed of $497 \pm 63 \text{ km s}^{-1}$ throughout the distances observed.

251 This CME was tracked though the HI-1 field of view using the same interface as
 252 used with Solar Storm Watch (Barnard et al., 2017). By identifying bright fronts within
 253 an image and taking the radial distance of these features where they cross the ecliptic
 254 plane, the elongation angle of each feature was measured. After repeating this for all HI
 255 images, a time-elongation profile was produced. We can see the time-elongation profiles
 256 for the features of the December 12th 2008 CME in figure 5a, where two fronts were iden-
 257 tified in STEREO-A data and three-fronts from STEREO-B. These profiles were also
 258 included in the work by Scott et al. (2019), however only two features were identified within
 259 STEREO-B's HI images due to a limitation on the number of features that could be tracked
 260 at that time. Now, we find the "inner front" (shown in blue) to have a less-noisy time-
 261 elongation profile, similar to the profiles seen by STEREO-A. A third feature (shown in
 262 green) has not been used or identified in other ghost-front studies, but could plausibly
 263 be explained by one of the other multiple front theories mentioned earlier. Hence, here
 264 in this work we do not use the time-elongation profile of the third front in our analysis.
 265 However, we do investigate if an extension of the ghost-front theory can explain this fea-
 266 ture.

267 From ACE data, seen in figure 3, we analyse the CME's arrival at Earth. We in-
 268 terpret the enhancement in ion density at 07:00 UTC ± 1 hour on the December 16th
 269 2008 as the arrival of the compressed solar wind ahead of the CME, just under four days
 270 from the first COR observation. Upon the arrival of the CME the solar wind velocity
 271 increases, seeing the value ranging from approximately 330 km s^{-1} at arrival to 370 km s^{-1}

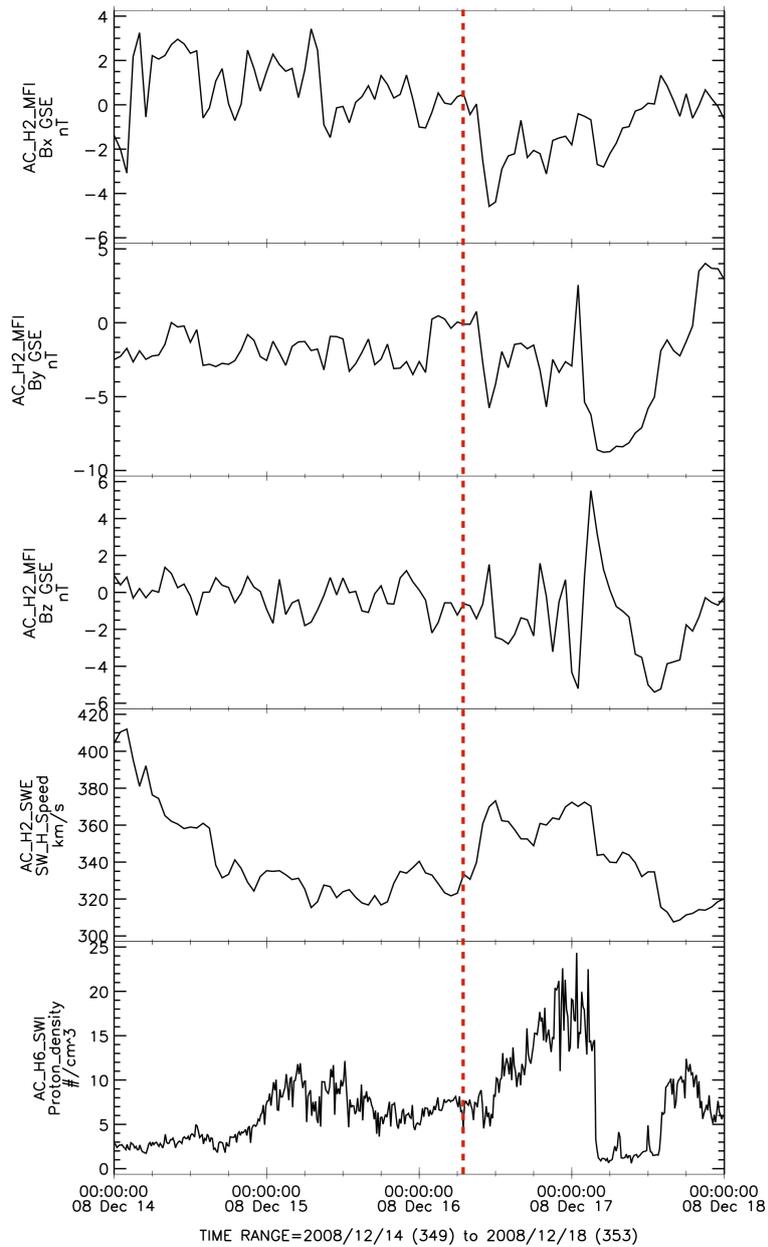


Figure 3. Solar wind data from NASA’s ACE spacecraft. Top three panels show the heliospheric magnetic field (in GSE coordinates) and the lower two panels show the solar wind radial speed and proton concentration. The red dashed vertical line highlights the CME arrival time, as estimated from an enhancement in ion density. (Source: <https://cdaweb.gsfc.nasa.gov/cgi-bin/eval1.cgi>.)

272 several hours later, along with an associated enhancement in ion density. It becomes clear
 273 that these features are indeed associated with a CME when, a few hours later, the mag-
 274 netic field turns southwards indicating the arrival of the magnetic structure. This ob-
 275 served arrival time is consistent with other studies (Davis et al., 2009; Deforest et al.,
 276 2013; Zhang et al., 2019).

277 4 Method

278 4.1 CME propagation ensemble

279 For the CME of December 12th 2008, a deterministic run of the HUXt model ini-
 280 tialised with MAS solar wind data for CR2077 and CME initial conditions (42° width,
 281 497 km s^{-1} speed, 10° longitude and 9° latitude), determined from STEREO COR ob-
 282 servations, produces time-elongation profiles for the nose and the flank of the CME that
 283 are in reasonable agreement with the observations and result in an arrival time at Earth
 284 that is within 5.5 hours of the in-situ spacecraft observations. While this could be con-
 285 sidered to be a successful result, it does not explore the sensitivity of the model to the
 286 uncertainty and accuracy of the initial solar wind and CME input data.

287 In this study, we investigate whether alterations can be made to the HUXt model
 288 to better reproduce the HI observations and whether such improvements in turn improves
 289 the model's ability to reproduce the observed CME arrival time at L1. To do so, 201-
 290 member (200 random and 1 deterministic member) ensembles of the HUXt model were
 291 made for a range of initial conditions that either saw alterations to the solar wind con-
 292 ditions or the CME parameterisation, each one called an experiment. For each exper-
 293 iment, the initial conditions of the modelled CME were perturbed using a random, uni-
 294 form distribution corresponding to the uncertainty in observations. The modelled time-
 295 elongation profiles of the CME nose and flank were extracted from each model run by
 296 identifying where the CME front crossed the radial line of propagation (the nose) and
 297 the maximum observed elongation of the front from the position of each spacecraft (the
 298 flank). The combined nose and flank profiles as seen from both spacecraft simultaneously
 299 were used to determine the efficacy of each model run by computing the Root-Mean-Square-
 300 Error (RMSE) between model and observations. In this way, the model runs within each
 301 experiment can be ranked according to RMSE. By determining the relation between RMSE
 302 and arrival time error (ATE), we can determine if the HI observations contain any in-
 303 formation that could be used to improve the estimated arrival time at Earth. We repeat
 304 the analysis using the flank only, similarly to the work of Barnard et al. (2020), to as-
 305 sess the benefit from using ghost fronts to evaluate model performance.

306 Three alterations to the HUXt model setup were considered as variable that could
 307 impact the model's outcome; the use of data assimilation to the MAS solar wind scheme;
 308 the location of the model inner boundary ($30 R_\odot$ compared with $8 R_\odot$); and the mag-
 309 nitude of the assumed uncertainties in the initial CME observations. By taking combi-
 310 nations of these model alterations, a total of five experiments are presented here:

- 311 1. CME initialised with the parameter uncertainty estimated from COR observation,
 312 interacting with BRaVDA solar wind solution from $30 R_\odot$.
- 313 2. CME initialised with twice the parameter uncertainty estimated from COR ob-
 314 servation, interacting with BRaVDA solar wind solution from $30 R_\odot$.
- 315 3. CME initialised with the parameter uncertainty estimated from COR observation,
 316 interacting with BRaVDA solar wind solution from $8 R_\odot$.
- 317 4. CME initialised with twice the parameter uncertainty estimated from COR ob-
 318 servation, interacting with BRaVDA solar wind solution from $8 R_\odot$.
- 319 5. CME initialised with the parameter uncertainty estimated from COR observation,
 320 interacting with MAS solar wind solution from $30 R_\odot$.

321 The initial condition values, associated uncertainty, and solar wind scheme used in each
 322 experiment can be found in table 1.

Table 1. Overview of initial conditions variables used for each experiment. Value and uncertainty (if applicable) of the inner boundary radial distance, CME speed, CME source location, CME full width, CME thickness, Earth’s Carrington longitude at the start of the experiment and the solar wind scheme are displayed.

Model Run Name	Inner Boundary (R_{\odot})	Speed (kms^{-1})	Longitude ($^{\circ}$)	Latitude ($^{\circ}$)	Width ($^{\circ}$)	Thickness (R_{\odot})	Earth’s Carrington Longitude	Solar Wind solution
experiment 1	30	497 ± 63	10 ± 4	9 ± 2	42 ± 6	5 ± 2	67.966	BRaVDA
experiment 2	30	497 ± 126	10 ± 8	9 ± 4	42 ± 12	5 ± 4	67.966	BRaVDA
experiment 3	8	600 ± 150	10 ± 4	9 ± 2	42 ± 6	16.9 ± 2	63.363	BRaVDA
experiment 4	8	600 ± 300	10 ± 8	9 ± 4	42 ± 12	16.9 ± 4	63.363	BRaVDA
experiment 5	30	497 ± 63	10 ± 4	9 ± 2	42 ± 6	5 ± 2	67.966	MAS

323 In the following section, the reasoning behind the choice of each of these experi-
 324 ments will be considered in more detail.

325 **4.2 Reducing the inner boundary of the model**

326 CMEs have been shown to be accelerating in the initial stages of the propagation
 327 up to distances greater than $30 R_{\odot}$ (Manchester et al., 2017). The acceleration is not
 328 primarily due to the interaction with the ambient solar wind, but, the CME’s magnetic
 329 energy within the flux-rope driving it forward up to $20 R_{\odot}$ (Subramanian & Vourlidis,
 330 2007). The HUXt model does not account for such CME acceleration, but as the inner
 331 boundary is typically taken to be $30R_{\odot}$, as is the case with the MAS model, this is not
 332 a major issue. However, injecting a CME with a geometrically simple shape into the model
 333 at $30 R_{\odot}$ does not account for any distortion the CME may have undergone while prop-
 334 agating out from the initial coronagraph observations (at $8 R_{\odot}$ where the CME appears
 335 to have a more regular structure). With the lack of magnetic field in HUXt CMEs (or
 336 other representations of cone model CMEs), it is important to ensure the kinematics of
 337 a CME in the earlier stages of propagation are represented by other means when using
 338 a lower inner boundary.

339 Firstly, in order to initialise the ambient solar wind at $8 R_{\odot}$, the solar wind must
 340 be back-mapped to this location from the $30 R_{\odot}$. We must consider changes in solar wind
 341 acceleration and solar longitude, resulting from solar rotation during the transit time,
 342 T , taken for this radial movement to occur between the two boundaries. Riley and Li-
 343 onello (2011) stated a solar wind acceleration term based on MHD simulations as

$$accV(r) = \alpha V_0 \left[1 - exp \left(\frac{-(r - r_0)}{r_H} \right) \right], \tag{1}$$

344 where an acceleration term $\alpha = 0.15$ and the scale height $r_H = 50 R_{\odot}$ produced results
 345 in agreement with the HelioMAS model. This equation is used to compute $V(8 R_{\odot})$ from
 346 $V(30 R_{\odot})$. T is then given by

$$T = \int_{8R_{\odot}}^{30R_{\odot}} \frac{1}{V} dr, \tag{2}$$

347 where:

$$V(r) = V_0 + accV(r). \tag{3}$$

Using the information derived from these above equations, the change of solar longitude, $\Delta\phi$, can be calculated by

$$\Delta\phi = 2\pi \frac{T}{T_{syn}}, \quad (4)$$

where T_{syn} is the sidereal rotation period of the Sun. Since a structured solar wind is used, there are a range of different transit times and hence longitudinal changes between $30 R_{\odot}$ and $8 R_{\odot}$. This back-mapping method ignores any stream interaction that take place. The final step in the process interpolates the output on the HUXt's longitudinal grid spacing.

A cone CME will also alter in size significantly between these radial distances, whereby the CME radius will increase as the event expands throughout its propagation. As a result of the CME radius being smaller at a lower inner boundary, we must conserve the momentum of the event which can be done through the thickness parameter. Ultimately, we aim to simulate a CME from $8 R_{\odot}$ that will obtain similar parameters at $30 R_{\odot}$ to those we used to initialise the model at the same radial distance. To ensure the total radial extent of the CME is kept constant then the following equation must be true

$$2r_{30} + T_{30} = 2r_8 + T_8 \quad (5)$$

where r is the CME radius of the spherical ends and T is the thickness, together making the initial "sausage" shape of the CME, and the subscript tells us the initial height of the CME. With a reduced CME radius at $8 R_{\odot}$, we require a greater initial thickness than seen at $30 R_{\odot}$ to compensate. The radius of the spherical ends can be calculated using

$$r_{cme} = R_{cme} \tan(\theta) \quad (6)$$

where R is the initial height of the CME and θ is the half-width angle of the CME. By substitution, we can find the equivalent thickness of the CME at the lower inner boundary using

$$T_8 = 2R_{30} \tan(\theta) + T_{30} - 2R_8 \tan(\theta). \quad (7)$$

For the December 12th 2008 event, we calculate that for a CME launching from $8 R_{\odot}$ with half-width of 21° a thickness of $16.9 R_{\odot}$ is required.

Since we cannot model the initial CME acceleration with HUXt, we simulate this CME acceleration by initialing the model with a speed greater than observed at $30 R_{\odot}$ when using a lower inner boundary. Using CME parameters estimated from coronagraph observations, a small set of runs was carried out in which only the initial CME speed is changed between runs. After exploring a large range of initial speeds ($300 - 1200 \text{ km s}^{-1}$), a local minimum in RMSE for HI-1 measurements is found to occur at approximately $600 \pm 100 \text{ km s}^{-1}$ for all features tracked. Hence, we use this value to define the CME initial speed at the $8 R_{\odot}$ boundary. The model runs continue to use COR observations made at the radial distance of this lower boundary to describe the source location and longitudinal width of the CME.

Finally, we want to test if the quoted uncertainties of the coronagraph fit were adequate for capturing all potential outcomes. We allow the ensemble members to perturb are the given uncertainty and twice the given uncertainty, still with a random, uniform distribution.

5 Results and Discussion

5.1 Modelling the solar wind: BRaVDA or HelioMAS?

We explore the differences of using HUXt with and without data assimilation of the in-situ solar wind observations by comparing the deterministic outcome of the model in each case. Earlier, we described the differences between the two solar wind schemes

391 from the near-Earth solar wind speed time-series, however no CME was present in this
 392 case. Here, we initialise the CME using the estimated coronagraph observations for a
 393 $30 R_{\odot}$ inner boundary. In figure 4 we can compare the impact of using data assimilated
 394 solar wind scheme using the geometric shape of the CME two-days into the propagation
 395 and the time-elongation profiles for the nose and flank.

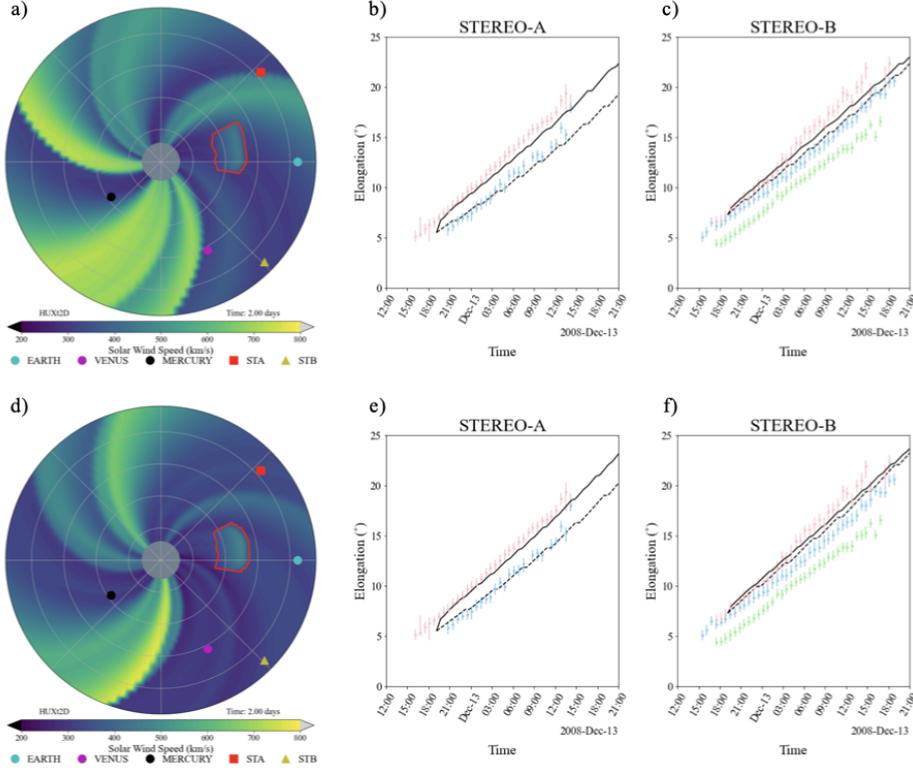


Figure 4. Comparison of HUXt output when ran with different solar wind schemes. The top panel shows the results from the MAS non-data assimilated solar wind scheme. The bottom panel shows the results from the BRaVDA data assimilated solar wind scheme. Ecliptic plane plot are displayed in a) and d), showing a CME’s interaction with the solar wind structure two-day into simulation. Structured solar wind is shown by the background colour, whilst the CME edge is identified by the red-outlined shape. Markers for Earth, STEREO, Venus, and Mercury positions are also included. Time-elongation plots measured from the position of STEREO-A [b) and e)] and STEREO-B [c) and f)] show the evolution of the leading edge nose (black dashed) and flank (black solid) throughout the Heliospheric-Image 1 field of view. Coloured error bars show the HI-1 data.

396 When the MAS (non-data assimilated) solar wind scheme is used, we find that the
 397 time-elongation profile of the nose and flank throughout the HI-1 field of view agree well
 398 with STEREO-A observation, obtaining a combined RMSE of 0.913° , and STEREO-B,
 399 obtaining a combined RMSE of 0.809° . From both spacecraft, the RMSE value is 0.856° .
 400 This CME hindcast has an L1 arrival of $12:30 \text{ UTC} \pm 34 \text{ minutes}$ on December 16th 2008
 401 (5.5 hours after observed arrival) with a speed of 328 kms^{-1} .

402 Alternatively, using BRaVDA to assimilate in-situ observations we find that the
 403 CME hindcast has an estimated arrival time at Earth of $07:53 \text{ UTC} \pm 34 \text{ minutes}$
 404 on the same date. This hindcast provides an estimated arrival time that is 4.62 hours ear-

405 lier than MAS, and is within the observed arrival time uncertainty (07:00 \pm 1.0 hour on
 406 the December 16th 2008). Further, we find that the CME arrives with a speed of 353
 407 kms^{-1} , which falls within the observed range of speeds seen in figure 3. Computing the
 408 RMSE of the time-elongation profiles, we find that by using BRaVDA we recreate the
 409 observed time-elongation profiles better from the viewpoint of STEREO-A (0.684°) but
 410 not from STEREO-B (0.929°). Due to the broader, faster stream seen in the HeliMAS
 411 solution at 12:00 UTC on the December 13th 2008 in figure 1, the CME evolution dif-
 412 fers significantly enough whereby the westward flank (observed by STEREO-B) is dragged
 413 out with time causing a flatter leading edge to the CME, as seen in figure 4b. This pro-
 414 duces an flank time-elongation profile with larger angles which the RMSE values sug-
 415 gest is more representative of the HI-1 observation. Despite this, when combining the
 416 viewpoint of both spacecraft, the BRaVDA solution still reduced the RMSE of the time-
 417 elongation profiles (0.830°), although marginal.

418 In this case study, neither solar wind scheme produced a solution with a significantly
 419 smaller RMSE than the other. But it is highlighted that understanding CME distortion
 420 cause by interaction with solar wind streams plays a vital role in accurately reproduc-
 421 ing HI-1 observations. But since the BRaVDA scheme reduced the error in arrival time
 422 then we continue to use this solar wind scheme going forward with the study.

423 5.2 Constraining models using Ghost-Front features

424 In figure 5 we show a development of simulating time-elongation profiles that rea-
 425 sonably reproduce the profiles of tracked fronts within HI-1 imagery. In the top row we
 426 show the observed time-elongation profiles seen by STEREO's Heliospheric Imager-1 (more
 427 details of these plots was included in section 2.3), and in the middle row we show all HUXt
 428 simulated profiles from the 201-member ensemble of experiment 1 (modelling CME from
 429 $30 R_\odot$ with parameters varying within coronagraph observed estimates; table 1) over-
 430 laid with the HI-1 observations. We show the feature identified as the flank in red and
 431 the nose in blue, directly relating to the colours seen in the top row. The bold line in-
 432 dicates the deterministic run. From this ensemble, we show the best-fit time-elongation
 433 profile for the nose and flank as seen from both STEREO spacecraft - i.e., the model run
 434 that produced the lowest RMSE - in the bottom row. From the perspective of STA, the
 435 modelled profiles agree with the front elongations fairly well (0.598° RMSE). The flank
 436 appears to fit very well, with the simulated profile agreeing within the error of the HI-
 437 1 observed fronts, meanwhile, the nose feature doesn't perform quite as well. However
 438 the gradient of these features is in good visual agreement. On the other hand, from the
 439 perspective of STB, the separation between the nose and flank is narrower than seen in
 440 the observations. Here we see that the gradient of the flank's time-elongation profile is
 441 less than observed, with many of the observations differing from the simulations by more
 442 than can be explained by uncertainties. The model nose elongation agrees with the points
 443 better and therefore we find a RMSE of 0.752° . Collectively, this model run agrees with
 444 the observations with a RMSE of 0.688° .

445 Here, we can also introduce the third feature seen by STEREO-B (shown in green
 446 in figure 5). This appears to correspond to the location of the CME flank observed from
 447 STEREO-A, as seen from STEREO-B. This may be due to the proximity of this flank
 448 to STEREO-B resulting in a bright feature. Using this interpretation, the observations
 449 match the modelled location of this feature (0.556° RMSE) highlighting a good fit. Whilst
 450 this is interesting to note, and may help identify the cause of this feature in the images,
 451 it is not used further in the analysis presented in this paper, since it is unclear how fre-
 452 quently such features would appear in other case studies.

453 The narrower difference between nose and flank time-elongation profiles produced
 454 by the model suggests that the model is not capturing the CME distortion accurately
 455 enough. More precisely, the radius of curvature of the CME front is too small. One ex-

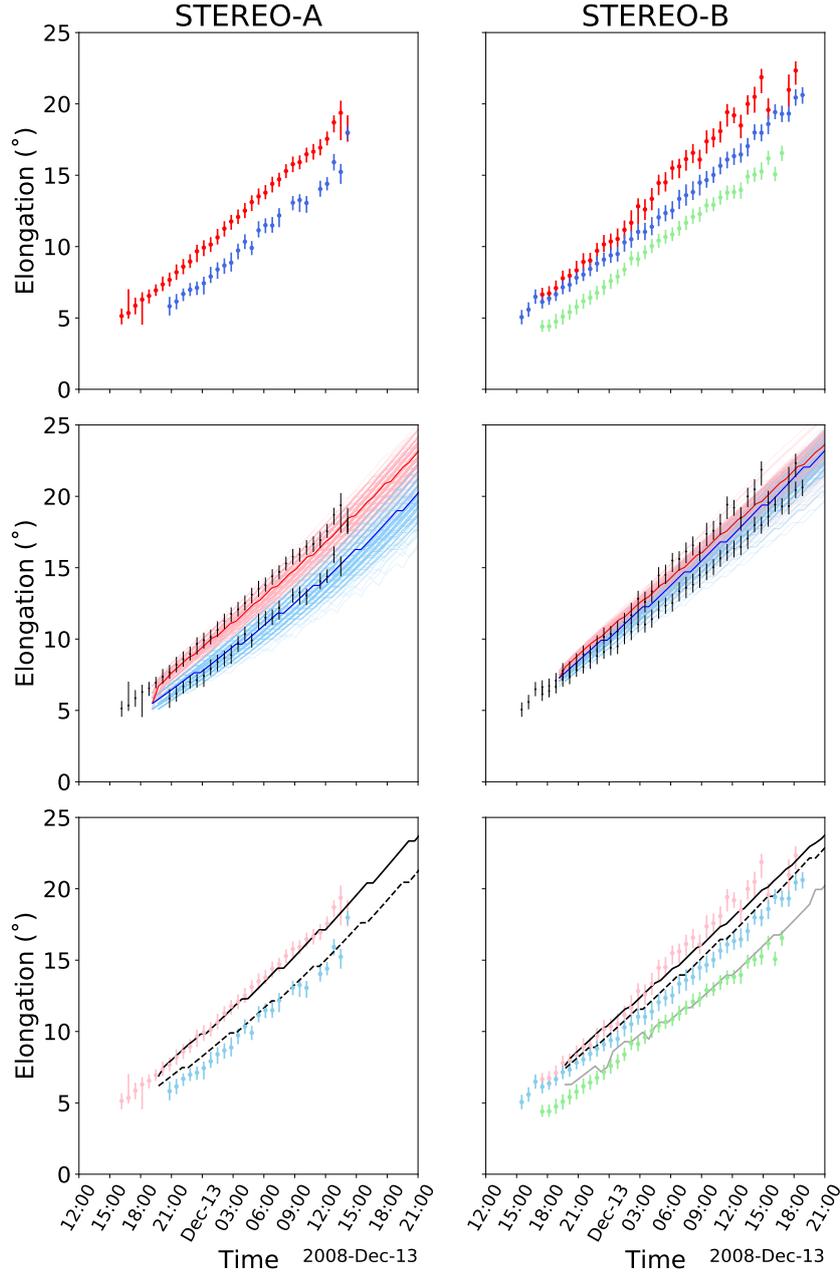


Figure 5. Time-elongation profiles for experiment 1, with HUXt modelling CME propagation from $30 R_{\odot}$ inner boundary, measured from the position of STEREO-A (left column) and STEREO-B (right column). Top row: error bar profiles of the dense-plasma fronts seen in Heliospheric Imager data, measured using the Solar StormWatch technique. Middle row: Modelled data showing the profiles of the nose (blue) and flank (red) from the 201 member ensemble, with the deterministic run shown in bold. Bottom row: profiles of the best-fit run, i.e. the minimised RMSE error of the nose and flank from both spacecraft. Here, we also show the profile of the secondary flank in solid grey in STEREO-B only.

456 planation could be that we have used an insufficient range over which the initial parameters of the CME were allowed to vary in the ensemble. We later test this by allowing
 457

458 the initial CME conditions to vary by up to twice the quoted parameter uncertainties.
 459 Another reason could be that significant distortion of the CME occurs before $30 R_{\odot}$. De-
 460 spite this "best-fit" model run agreeing with the HI observation better than the deter-
 461 ministic run, we find that we obtain an arrival time that is 2 hours 18 minutes earlier
 462 and falls outside the observed uncertainty. Such result demonstrates the limitation of
 463 using HI time-elongation profiles to constrain a CME's evolution as they are essentially
 464 degenerate for many combinations of initial CME parameters. So, as you find here, the
 465 "best fit" time-elongation profile might correspond to a simulation that poorly reflects
 466 the arrival time at Earth.

467 Further, due to the random nature of the sampling, identifying the experiment mem-
 468 ber that produces the lowest RMSE may not representative of the entire ensemble, and
 469 as such it is not sufficient to use this to draw conclusions on the benefit of using ghost
 470 fronts to constrain propagation modelling. Hence we now explore the correlation between
 471 the HI-1 elongation RMSE and arrival time error, as shown in figure 6 for the experi-
 472 ments modelled from a $30 R_{\odot}$ inner boundary. We note that there are two data clusters
 473 for experiment 1 (fig 6a) of which both appear to form a local minimum at similar ar-
 474 rival time error, though one has much higher RMSE values. The cause of the two clus-
 475 ters is yet to be determined, but a plausible explanation is that there is a common sen-
 476 sitivity to one of the initial values between these model runs. For example a wider or nar-
 477 rower CME can produce time-elongation profiles that are off-set from the observations
 478 by a few degrees yet the propagation of the nose is not impacted and hence produces an
 479 arrival time in agreement with other members. By finding a subset of ensemble mem-
 480 bers which identify the members with lowest RMSE value, we can ensure that we only
 481 consider those ensemble members that best match with the HI-1 data.

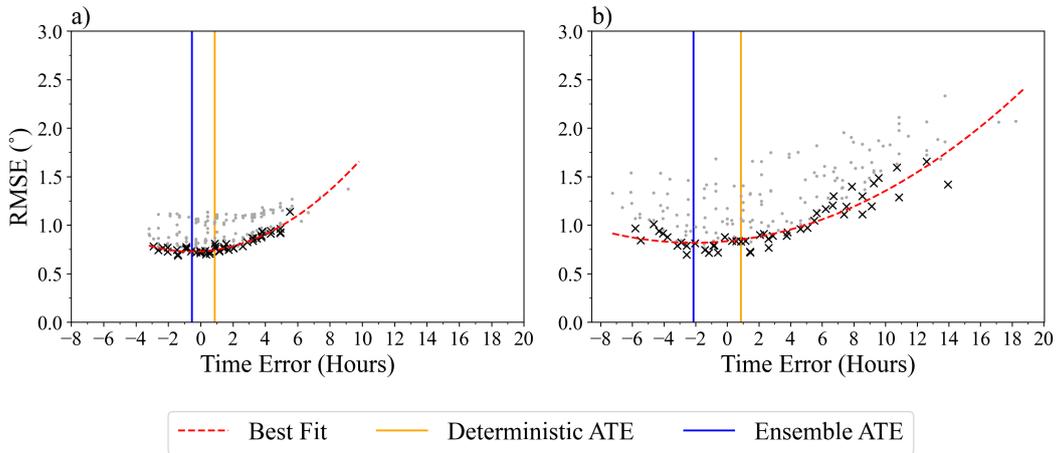


Figure 6. Scatter plots of time-elongation profile RMSE values against arrival time error for each model member. In these experiments, HUXt is configured with an inner boundary of $30R_S$ with initial conditions allowed to vary with (a) the coronagraph estimated uncertainty and (b) twice the estimated uncertainty. A quadratic line of best fit (red, dashed) is fit to the lowest 25% of data per a bin (black dots).

482 To do this, we bin the members by their ATE, with 34 minute resolution (HUXt's
 483 output time-step in the configuration of this study), to ensure the ensemble distribution
 484 continues to be represented. The lowest 25% of data points per each bin are then found
 485 and used to create the data subset in which the regression line is best fit to. By taking
 486 the subset of data like this, we can explore the variation of best-fit RMSE as a function
 487 of arrival time while accounting for the noise generated by random sampling of the CME

488 initial conditions. Hence, upon calculating the minimum value of the curve we can anal-
 489 yse the relationship between the quality of fit and the arrival time at L1. The uncertainty
 490 of the fit and is measured by where the line of best fit is significantly different from the
 491 minimum value, based on 1 standard deviation error. In such way, sharp curves that bet-
 492 ter define the minimum will have a low uncertainty whilst less well defined curves have
 493 greater uncertainty.

494 For experiment 1, the scatter produces a curve minimising at -0.5 ± 0.2 hours which
 495 equates to an arrival time before the observed arrival. This is within the observed ar-
 496 rival uncertainty of ± 1.0 hours. The result is 1.5 hours before the deterministic hind-
 497 cast, but since this is only a difference of three model resolution time-step we cannot say
 498 that this is a strong improvement on the deterministic hindcast by. The RMSE value
 499 associated with the curve minimum in 0.726° . We carry out the same analysis for ex-
 500 periment 2, where the ensemble of CME parameters are generated by randomly sampling
 501 from uniform distributions with twice the spread of experiment 1. We find that the RMSE
 502 variation minimises at -2.1 ± 0.5 hours. This result produces a less accurate arrival time
 503 than both the deterministic and experiment 1 model runs, which is suggested by an in-
 504 significant change to the minimising RMSE value of 0.818° , highlighting no benefit for
 505 doubling the range over which the CME initial conditions are allowed to vary. From in-
 506 creasing the initial parameter range and maintaining the ensemble size, we note that there
 507 is a reduced resolution within the parameter space which may increase the variation in
 508 RMSE values between ensemble members.

509 If only the flank time-elongation profile (i.e. the outermost CME front detected in
 510 HI) is considered, as was done in the work by Barnard et al. (2020), we find that the quadratic
 511 curve minimises at -4.4 ± 1.1 hours with RMSE of 0.900° for experiment 1 and $-11.1 \pm$
 512 4.9 hours with RMSE of 0.697° for experiment 2. However, we note these ATE values fall
 513 outside of the range of values obtained by the ensemble members (minimum ATE is 03:48
 514 UTC December 16th 2008 and 23:44 UTC December 15th 2008 respectively) highlight-
 515 ing that the outcomes are likely to be unreliable estimates. Thus far in the case study,
 516 an early CME arrival is estimated when launched from an initial height of $30 R_\odot$, but
 517 the accuracy of the ATE estimate is improved when including CME nose tracking in en-
 518 semble modelling.

Table 2. Overview of the arrival time error results of the experiments. Deterministic value is computed using a single fun of HUXt initialise with the COR-1 parameter estimates. Ensemble (Ens.) ATE is the minimum of the quadratic relation between the HI-1 elongation RMSE and arrival time error for the lowest 25% of members, as seen in figure 6 and 7. Where the number is in bold font, we note that the value falls outside of the ensemble range of ATE values.

Model Run Name	Deterministic	Nose and Flank	Flank Ens.
	ATE (hours)	Ens. ATE (hours)	ATE (hours)
Experiment 1	0.9 ± 0.6	-0.5 ± 0.2	-4.4 ± 1.1
Experiment 2	0.9 ± 0.6	-2.1 ± 0.5	-11.1 ± 4.9
Experiment 3	0.4 ± 0.6	2.8 ± 0.5	0.7 ± 0.8
Experiment 4	0.4 ± 0.6	2.9 ± 0.6	1.6 ± 1.0
Experiment 5	5.5 ± 0.6	n/a	n/a

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5.3 Lowering the model inner boundary to $8 R_{\odot}$

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By reducing the inner boundary to the radius at which the parameters were estimated, the deterministic hindcast produces an L1 arrival at 07:24 UTC + 34 minutes on the December 16th 2008 with a speed of 333 kms^{-1} . Compared to $30 R_{\odot}$ deterministic, the difference in arrival time estimate is smaller than the model resolution, therefore, we can say the model output performs similarly. When comparing the time-elongation profiles of the nose and flank to the fronts observed in HI-1, the RMSE values for STA, STB, and both spacecraft are 0.611° , 0.741° , and 0.688° respectively. This equates to an improvement in the RMSE compared with the $30 r_{\odot}$ deterministic run by at least 10%.

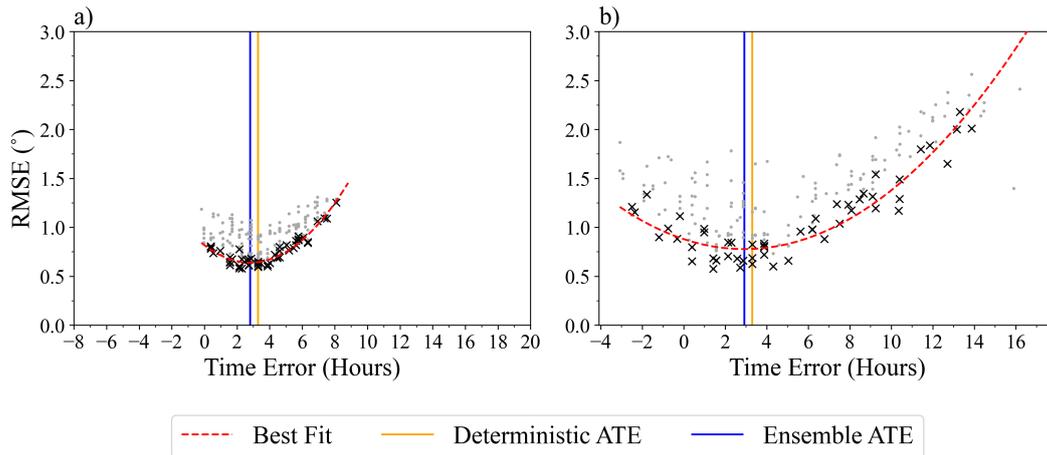


Figure 7. Scatter plots of the time-elongation profile RMSE values and the arrival time error for each model member in (a) experiment 3 and (b) experiment 4. A quadratic line of best fit (red, dashed) is fit to the lowest 25% of data per a bin (black dots).

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We analyse the model outputs for experiment 3 and 4 and display the relationship between the RMSE and ATE in figure 7. By fitting the curve to the lowest 25% of binned data, we find that the curve minimises at 2.8 ± 0.5 hours and 2.9 ± 0.5 hours for experiment 3 and 4 respectively. There is no improvement to the arrival time accuracy produced here by both experiments than compared to those already discussed, and we find that the estimates now predict an arrival later than observed. But it is interesting to see that despite doubling the range over which the initial parameters can vary in experiment 4, the fit to the data produces a consistent ATE result. This was not observed in the earlier experiment suggesting that by simulating early distortion to the CME then the time-elongation profiles throughout the HI-1 field of view is more unique to the model member allowing RMSE values to better indicate the performance of the initial parameters. We see this demonstrated in the figure by a sharper best-fit curve fitting to the data than compared to their counterpart experiment for $30 R_{\odot}$.

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The $8 R_{\odot}$ ensemble estimated ATE associates with a lower RMSE value than its $30 R_{\odot}$ counterpart. 0.726° drops to 0.638° when using the estimated coronagraph parameter uncertainty (experiment 3), and 0.818° drops to 0.776° when using twice the estimated uncertainties (experiment 4). These results highlight again that by allowing the CME to distort before reaching the field of view covered by HI-1 then the model can simulate the real data with better accuracy. Whilst this is expected of a real CME propagation through the heliosphere, it is important to note that the simplified HUXt model is replicating the kinematics.

549 Again, we look at the results for a situation where only the flank RMSE is consid-
 550 ered. It is found that the curve minimises at 0.7 ± 0.8 hours for experiment 3 and 1.6
 551 ± 1.0 hours for experiment 4, therefore producing a better estimate of the arrival time
 552 than using both the nose and flank features. Whilst previous results highlight that the
 553 elongation of multiple fronts detected in HI-1 data can be replicated using the nose and
 554 flank of a CME leading edge, these results here suggest that tracking a single feature may
 555 also be sufficient in producing arrival predictions as long as we account for early CME
 556 distortion by lowering the inner boundary of the model. The accuracy with which the
 557 flank’s time-elongation profile evolves from the lower inner boundary may be unique to
 558 this case study in which HUXt accurately captures the distortion of a relatively slow-
 559 moving event.

560 6 Conclusions

561 We have modelled the December 12th 2008 CME using the HUXt solar wind model
 562 with BRaVDA’s solar wind solution to produce multiple ensemble hindcasts. The data
 563 assimilated solar wind scheme was found to improve the arrival time error of the deter-
 564 ministic run by 4.62 hours, and produced an estimated arrival time that coincided with
 565 the observed CME arrival at L1 on December 16th 2008, 07:00 UTC ± 1.0 hour. Each
 566 experiment investigates the sensitivity of model inner boundary and the parameter un-
 567 certainty using a 201-member ensemble, and a collective L1 arrival time error is estimated
 568 based upon the relationship between the RMSE of HI-1 observation of the nose and flank
 569 and arrival time error of the individual members.

570 From the presented results, it is clear that accounting for CME distortion, at alti-
 571 tudes below the typical lower boundary of solar wind models, is vital for accurate mod-
 572 elling. We have shown here that this can be achieved by two methods. Firstly, we can
 573 use the multiple fronts seen in HI-1 images to inform us of the longitudinal distortion.
 574 By tracking the nose and flank of the CME we can evaluate the model’s performance against
 575 the time-elongation profiles of the ghost-front features observed by HI images. Allow-
 576 ing the ensemble parameters to vary within the uncertainty of chronograph parameter
 577 estimates, we were able to obtain a quadratic function that estimates an L1 arrival time
 578 within one-hour before the observed arrival. If only the flank was used then the same
 579 experiment the result was unrealistic, producing an ATE which fell outside the ensem-
 580 ble’s range. Alternatively, we can lower the model inner boundary to allow distortion
 581 to the CME shape before reaching the HI-1 field of view. In doing so, RMSE values of
 582 the ensemble members are more varied such that we obtain lower RMSE values via this
 583 method, suggesting the time-elongation profiles better represent the model’s sensitivity
 584 to the initial conditions. The hindcast estimate is less accurate than method one (ap-
 585 proximately 3 hours after observed arrival), however the ATE is consistent despite al-
 586 lowing ensemble parameters to vary within 1x and 2x the uncertainty of chronograph
 587 parameter estimates. On this occasion, evaluating model performance based on the CME
 588 flank profile not only worked but also estimated a better L1 arrival time error, produc-
 589 ing an arrival error of no more than 1.6 ± 1.0 hour. But this was associated to largest
 590 RMSE values of all the work presented. Upon investigating the efficacy of the uncertainty
 591 estimates associated with the CME characterisation from the data, there was no improve-
 592 ment to the hindcast (in this example) suggesting that the uncertainties quoted for the CME
 593 initial conditions are indeed adequate. Since current magnetohydrodynamic forecast mod-
 594 els use a heliospheric inner boundary of approximately $30 R_{\odot}$, above which most CME
 595 acceleration is considered to have already occurred, adopting the tracking ghost-front
 596 features as a way of constraining forecast models would seem to be the most convenient
 597 way of accounting for low altitude distortion of the CME.

598 Through this research, we have highlighted the efficacy of the HUXt model, show-
 599 ing the ability to account for the missing physics (i.e., plasma is assumed to be purely
 600 radial to eliminate the need of complex magnetic equations to describe its motion), es-

601 pecially at a lowered inner boundary, to produce near-accurate estimated arrivals at L1
602 from the flank alone.

603 Of course, we have only looked at a single event in this study, which is of a slow-
604 moving CME. Further investigations following this should explore whether these con-
605 clusions are consistent across a range of CME initial conditions and viewing geometries.
606 For fast-moving events, which are more geoeffective, it is also necessary to differentiate
607 between ghost fronts and shock-fronts propagating ahead of the CME. Here we analysed
608 results based upon observations from the twin spacecraft mission STEREO, whereas fu-
609 ture missions like ESA's Vigil L5 Langrangian point (Vourlidas, 2015; Akioka et al., 2005)
610 will only provide a HI viewpoint from one position outside the Sun-Earth line. In the
611 current event, our interpretation of the data suggests that a single spacecraft is able to
612 detect three features of the leading edge (the CME nose and both flanks), suggesting that
613 significant benefits to forecasting accuracy could be obtained using data from a single
614 spacecraft using ghost front observations.

615 Using HI time-elongation profiles to assess the model performance of CME distor-
616 tion can be limiting as they can be degenerate for many combinations of initial CME pa-
617 rameters. Therefore, estimating the arrival time using a function which describes the ar-
618 rival time error and RMSE is more representative of the ensemble, removes noise pro-
619 duced from random sampling and allows us to analyse and the impact ghost-fronts. This
620 would not be possible if we only seek the single member that produced the best-fit to
621 HI-1 observations. Adopting a comprehensive systematic sampling approach (where model
622 runs are created for all combinations of input parameters within their uncertainty in-
623 tervals) would be computationally more expensive but may enable a consistent estimate
624 of the best-fit to the ghost-front data. This will be considered in future work.

625 7 Data Availability

626 The HUXt model (version 1.0.0) used in the research can be accessed at [https://github.com/University-](https://github.com/University-of-Reading-Space-Science/HUXt)
627 [of-Reading-Space-Science/HUXt](https://github.com/University-of-Reading-Space-Science/HUXt). Data assimilated solar wind scheme BRaVDA can be
628 access from <https://github.com/University-of-Reading-Space-Science/BRaVDA>, using
629 MAS inputs from <http://www.predsci.com/mhdweb/home.php>. ACE data is accessible
630 via <https://cdaweb.gsfc.nasa.gov/cgi-bin/eval1.cgi>. Heliospheric Imager data was accessed
631 from <http://www.ukssdc.rl.ac.uk/solar/stereo/data.html> and analysed using the Solar
632 Stormwatch front tracking interface.

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