

Distances to molecular clouds at high galactic latitudes based on *Gaia* DR2

Qing-Zeng Yan^{1,2}, Bo Zhang², Ye Xu¹, Sufen Guo^{2,3}, Jean-Pierre Macquart⁴,
Zheng-Hong Tang^{2,3}, and Andrew John Walsh⁵

¹ Purple Mountain Observatory, Nanjing 210008, PR China
e-mail: qzyan@pmo.ac.cn

² Shanghai Astronomical Observatory, Chinese Academy of Sciences, Shanghai 200030, PR China

³ School of Astronomy and Space Science, University of Chinese Academy of Sciences, 19A Yuquanlu, Beijing 100049, PR China

⁴ International Centre for Radio Astronomy Research, Curtin University, GPO Box U1987, Perth WA 6845, Australia

⁵ Department of Physics and Astronomy and MQ Research Centre in Astronomy, Astrophysics and Astrophotonics, Macquarie University, NSW 2109, Australia

Received 28 September 2018 / Accepted 8 February 2019

ABSTRACT

We report the distances of molecular clouds at high Galactic latitudes ($|b| > 10^\circ$) derived from parallax and G -band extinction (A_G) measurements in the second *Gaia* data release, *Gaia* DR2. Aided by Bayesian analyses, we determined distances by identifying the breakpoint in the extinction A_G toward molecular clouds and using the extinction A_G of *Gaia* stars around molecular clouds to confirm the breakpoint. We used nearby star-forming regions, such as Orion, Taurus, Cepheus, and Perseus, whose distances are well known to examine the reliability of our method. By comparing with previous results, we found that the molecular cloud distances derived from this method are reliable. The systematic error in the distances is approximately 5%. In total, 52 molecular clouds have well-determined distances, most of which are at high Galactic latitudes, and we provide reliable distances for 13 molecular clouds for the first time.

Key words. dust, extinction – ISM: clouds

1. Introduction

Determining the distances of molecular clouds, the birthplaces of stars, is usually difficult. Many distance measurement methods, such as the photometric parallax method and the period-luminosity relation, are not applicable to molecular clouds. However, the distance relates directly to many important properties of molecular clouds, especially the mass and size, making their physical states and relationships with Galactic spiral arms largely uncertain.

There are a few approaches to derive molecular cloud distances, but they usually either yield large uncertainties or are not applicable to diffuse and translucent molecular clouds at high Galactic latitudes. For instance, maser astrometry with multiple-epoch Very Long Baseline Interferometry (VLBI) measurements (Xu et al. 2009; Zhang et al. 2013; Ortiz-León et al. 2017a) cannot be performed toward molecular clouds that lack masers. Furthermore, kinematic distance estimates based on modeled rotation curves (Roman-Duval et al. 2009; Rice et al. 2016) suffer from large uncertainties and ambiguities.

Optical extinction provides another approach to determining the cloud distances. However, the optical extinction derived from counting stars toward molecular clouds (Bok 1937; Magnani & de Vries 1986) involves large uncertainties. A more straightforward but sophisticated way is to examine the variation in optical extinction with respect to distance along the line of sight (see Knude & Hog 1998). For example, using this method, Schlafly et al. (2014) provided a large distance catalog of 18 well-known star-forming regions, and 108 molecular clouds with high Galactic latitude were cataloged by Magnani et al. (1985, hereafter

denoted MBM). They derived the distance and extinction for each star simultaneously from Pan-STARRS1 (Kaiser et al. 2010; Green et al. 2014) photometry and subsequently estimated the molecular cloud distances according to the breakpoint of the extinction. However, as Schlafly et al. (2014) have pointed out, their distances may have a $\sim 10\%$ systematic uncertainty.

The release of the second *Gaia* data (*Gaia* Collaboration 2016; *Gaia* Collaboration 2018), *Gaia* DR2, providing parallaxes for about 1.3 billion stars, a large proportion of which have G -band extinction (A_G) measurements, has advanced this approach substantially. The correspondence between the extinction A_G map (Arenou et al. 2018) and the distribution of molecular clouds (Dame et al. 2001) suggests that through A_G , we may be capable of detecting molecular clouds. Although, as cautioned by *Gaia* Collaboration (2018), A_G has large uncertainties, it is found to be reliable on average (see Andrae et al. 2018).

Gaia DR2 permits a more precise analysis of the extinction caused by molecular clouds than was previously possible and offers an independent means of examining the distance estimates of Schlafly et al. (2014). The rest of this paper is organized as follows. In Sects. 2 and 3, we introduce the method of reducing *Gaia* data and the process of deriving molecular cloud distances. We present the distance catalog in Sect. 4 and compare the results with previous studies in Sect. 5. The conclusions are summarized in Sect. 6.

2. *Gaia* DR2 and Planck 857 GHz data

We primarily investigated diffuse molecular clouds at $|b| > 10^\circ$, which are less likely to be contaminated by other molecular

clouds that are either adjacent or overlap along the line of sight. Despite being located at high Galactic latitudes, they are usually at a distance closer than 1 kpc from the Sun and are still located in the Galactic plane (Magnani et al. 1996; Schlafly et al. 2014). They tend to occupy large areas on the sky, making them able to optically extinct a large number of stars. Consequently, the extinction A_G imposed on stars behind molecular clouds can be investigated statistically.

We selected *Gaia* DR2 stars according to their parallaxes and extinction A_G . First, we removed stars with parallaxes <0.5 mas, corresponding to an upper distance threshold of 2 kpc. Molecular clouds at high Galactic latitudes are usually near the Sun (<1 kpc), and a 2 kpc cutoff suffices for our study. If the relative errors of stellar parallaxes were to exceed 20%, the corresponding distances would differ from the reciprocal of their parallaxes by a large amount (Bailer-Jones 2015), and consequently, we only kept stars whose relative parallax errors are smaller than 20%. For A_G , we required $A_G > 0$, which rejected 13 stars. In total, 30 259 242 *Gaia* stars meet these criteria.

The extinction A_G error of a single star, ΔA_G , is estimated with

$$\Delta A_G = \frac{1}{2} (84\text{th_percentile} - 16\text{th_percentile}), \quad (1)$$

where 84th_percentile and 16th_percentile are provided in *Gaia* DR2. To derive the distance and its standard deviation, we drew 10 000 samples from the Gaussian distribution $\mathcal{N}(\varpi, \Delta\varpi)$, where ϖ and $\Delta\varpi$ are the parallax and its error in *Gaia* DR2, and the mean and standard deviation of the distance corresponds to the mean and standard deviation of the reciprocal of the 10 000 samples, respectively.

We used Planck 857 GHz images (Planck Collaboration I 2014) to trace the molecular clouds rather than IRAS 100 μm and CO spectral lines. Although the IRAS 100 μm images (Schlegel et al. 1998; Miville-Deschênes & Lagache 2005) have a similar spatial resolution ($5'$) as the Planck 857 GHz survey, the sensitivity of the Planck 857 GHz survey is higher, and consequently, the Planck 857 GHz survey classifies *Gaia* stars more accurately and produces slightly better results, details of which are discussed in Sect. 4.1. Planck 857 GHz data are more complete than CO survey data (Dame et al. 2001) at high Galactic latitudes, and although Planck 857 GHz emission cannot distinguish molecular cloud components that overlap each other along the line of sight, this situation is not severe at high Galactic latitudes, and usually only one nearby molecular cloud component is present in one direction. The Planck 857 GHz images were only used to classify *Gaia* DR2 stars, not to build dust models.

3. Method

In this section we describe the method of deriving molecular cloud distances with the *Gaia* DR2 and Planck 857 GHz data. Principally, molecular clouds increase the extinction A_G of all stars behind them, thus producing breakpoints in A_G along the line of sight. Consequently, identifying the breakpoints in A_G is the essential point in our method.

First, we collected two classes of *Gaia* stars, on- and off-cloud stars, based on Planck 857 GHz emission. On-cloud stars, showing breakpoints in A_G , are the *Gaia* stars toward molecular clouds. We define as “off-cloud stars” *Gaia* stars around molecular clouds, and because they are not affected by molecular clouds, off-cloud stars have no breakpoints in

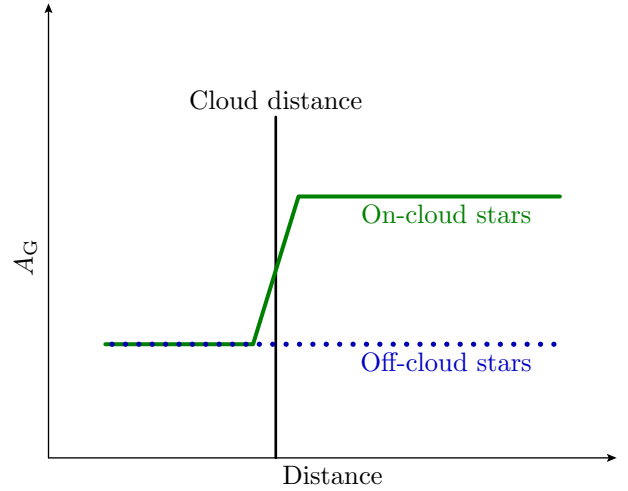


Fig. 1. Extinction A_G of on- (green) and off-cloud (blue) stars. The vertical black line marks the distance of the molecular cloud.

A_G . Practically, the breakpoints were determined with on-cloud stars alone, while off-cloud stars were used to confirm the breakpoints by eye. The schematic diagram of Fig. 1 depicts the extinction A_G feature of on- (green) and off-cloud (blue) stars.

In the second step, the distances were determined using Bayesian inference with on-cloud stars and were subsequently confirmed with off-cloud stars by eye. We use two molecular clouds, Taurus (Pineda et al. 2010; Ward-Thompson et al. 2016) and Gemini (Li et al. 2015), to illustrate the process of determining distances.

3.1. On- and off-cloud stars

The on- and off-cloud stars were classified according to Planck 857 GHz intensity thresholds. The on-cloud stars are those whose Planck 857 GHz emission is stronger than an intensity threshold (the signal level), while toward off-cloud stars, the Planck 857 GHz emission is fainter than a lower threshold (the noise level). We determined the two thresholds through fitting a mixed distribution combining two functions, Gaussian and exponential, as described below.

First, we manually drew a box region for each molecular cloud, and only *Gaia* stars in this box region were considered. This region contained at least part of the molecular cloud and incorporated an additional nearby region. This additional region, where the Planck 857 GHz emission was significantly lower than the molecular cloud region, contained off-cloud stars. These box regions are not necessarily the same as the traditional boundaries of molecular clouds, that is, they may be larger or smaller than the entire molecular clouds as long as these regions contain sufficient *Gaia* stars.

Second, we assigned each *Gaia* star a Planck 857 GHz emission value according to the Planck 857 GHz data. As the histograms demonstrate in Fig. 2, the Planck 857 GHz emission roughly contains two components: (1) a background noise that has an approximately Gaussian distribution in intensity, and (2) the molecular cloud emission that resembles an exponential distribution.

We used four parameters to model this mixed distribution: the mean (μ) and the standard deviation (σ) of the Gaussian part, a switch point (SP), and the rate (λ) of the exponential distribution. The cumulative distribution function (CDF) of the mixed

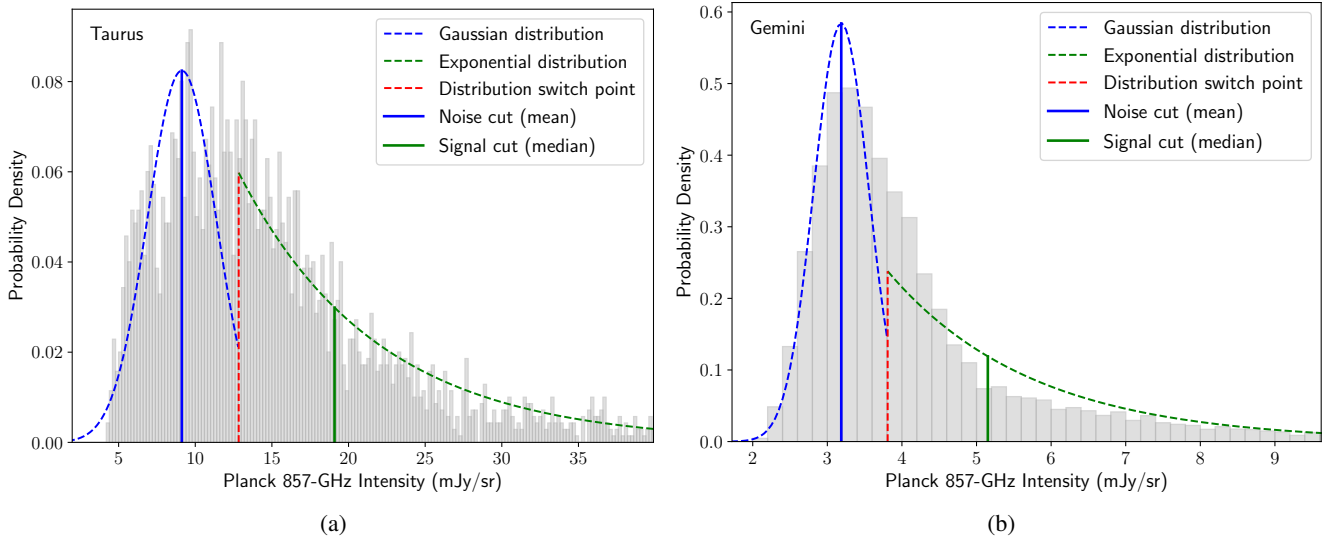


Fig. 2. Determining the noise and signal levels to classify on- and off-cloud stars for the Taurus (*panel a*) and Gemini (*panel b*) molecular clouds. The blue and green dashed lines represent the Gaussian and exponential distribution, respectively, and the dashed vertical red lines mark the switch point between the two distributions. We use the Gaussian mean (the solid vertical blue lines) for the noise cutoff level and the exponential median (the solid vertical green lines) for the signal cutoff level.

distribution (the likelihood) is

$$\int_{-\infty}^x p(I|\mu, \sigma, SP, \lambda) dI = \begin{cases} \int_{-\infty}^x \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(I-\mu)^2}{2\sigma^2}\right) dI, & x \leq SP, \\ \int_{-\infty}^{SP} \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(I-\mu)^2}{2\sigma^2}\right) dI \\ + \int_{SP}^x \lambda \exp(-\lambda(I-SP)) dI, & x > SP, \end{cases} \quad (2)$$

where I is the observed Planck 857 GHz intensity.

The four parameters were estimated by maximizing this likelihood, and the noise and signal level were subsequently determined according to the two distributions. As demonstrated in Fig. 2, we used a noise level of Gaussian mean μ , and a signal level of exponential median $(SP + \ln(2)/\lambda)$.

There is a trade-off in the choice of the signal level. Lower signal levels keep more stars, which is good for a statistical analysis, but involve many stars with low extinction A_G , which smears the breakpoints. Higher signal levels make the breakpoints evident, but fewer stars remain. The signal level, the median value of the exponential component, is a compromise choice. However, in Sect. 4.2, we show that as long as the breakpoints are detected, the derived distances are insensitive to the choice of signal levels.

The signal and noise levels classify on- and off-cloud stars, respectively. Although only on-cloud stars were employed to calculate distances, the off-cloud stars are useful as a reference to confirm the breakpoints.

3.2. Estimating the distance

With the extinction A_G and the corresponding distances of on- and off-cloud stars, we are now in the position to calculate the distance of molecular clouds. We built a Bayesian model to estimate the distance with on-cloud stars and solved for the parameters in the model with Markov chain Monte Carlo (MCMC) sampling. We emphasize that off-cloud stars were not involved in the model but were only used to confirm the breakpoint by eye.

The extinction A_G of on-cloud stars includes three components: (1) the foreground A_G , (2) the background A_G , and (3) the transition values. When passing through a molecular cloud along the line of sight, the extinction A_G gradually increases from the foreground A_G (which is usually 0 mag) level to the background A_G level. Here, we ignored the third component and focused on the average distance in our model. Because the minimum (A_G^{\min}) and maximum (A_G^{\max}) values of the extinction A_G are 0 and 3.609 mag, respectively, we followed the procedure of Andrae et al. (2018) using a truncated Gaussian distribution to model the on-cloud star extinction A_G .

Our model involves four parameters, the cloud distance (D), the extinction A_G dispersion (σ_1) of foreground stars, and the extinction A_G (μ_2) and its dispersion (σ_2) of background stars. We found that the extinction A_G of foreground stars (μ_1) is precisely zero, but for only four molecular clouds, therefore we used $\mu_1 = 0$ in most cases. These four molecular clouds have additional small molecular cloud components in front of them, therefore they treated them separately. We used a lower distance cutoff to remove the foreground stars of the small molecular cloud components, and their μ_1 were also modeled, which means that we used five parameters in total for them.

With these four parameters and a list of distances d_i with standard deviations Δd_i , which were calculated with the reciprocal of 10 000 parallax samples (see Sect. 2), and extinction A_{G_i} (with standard deviations ΔA_{G_i}), we derive an appropriate likelihood below. This approach requires no binning, and the binned extinction A_G and distances were only used for visual confirmation.

We calculated the likelihood of each star on the condition that it was in front of or behind the molecular cloud. Denoting the CDF of the standard normal distribution as

$$\phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-t^2/2} dt \quad (3)$$

and given D , the probability for a star to be in front of the molecular cloud is

$$f_i = \phi\left(\frac{D - d_i}{\Delta d_i}\right), \quad (4)$$

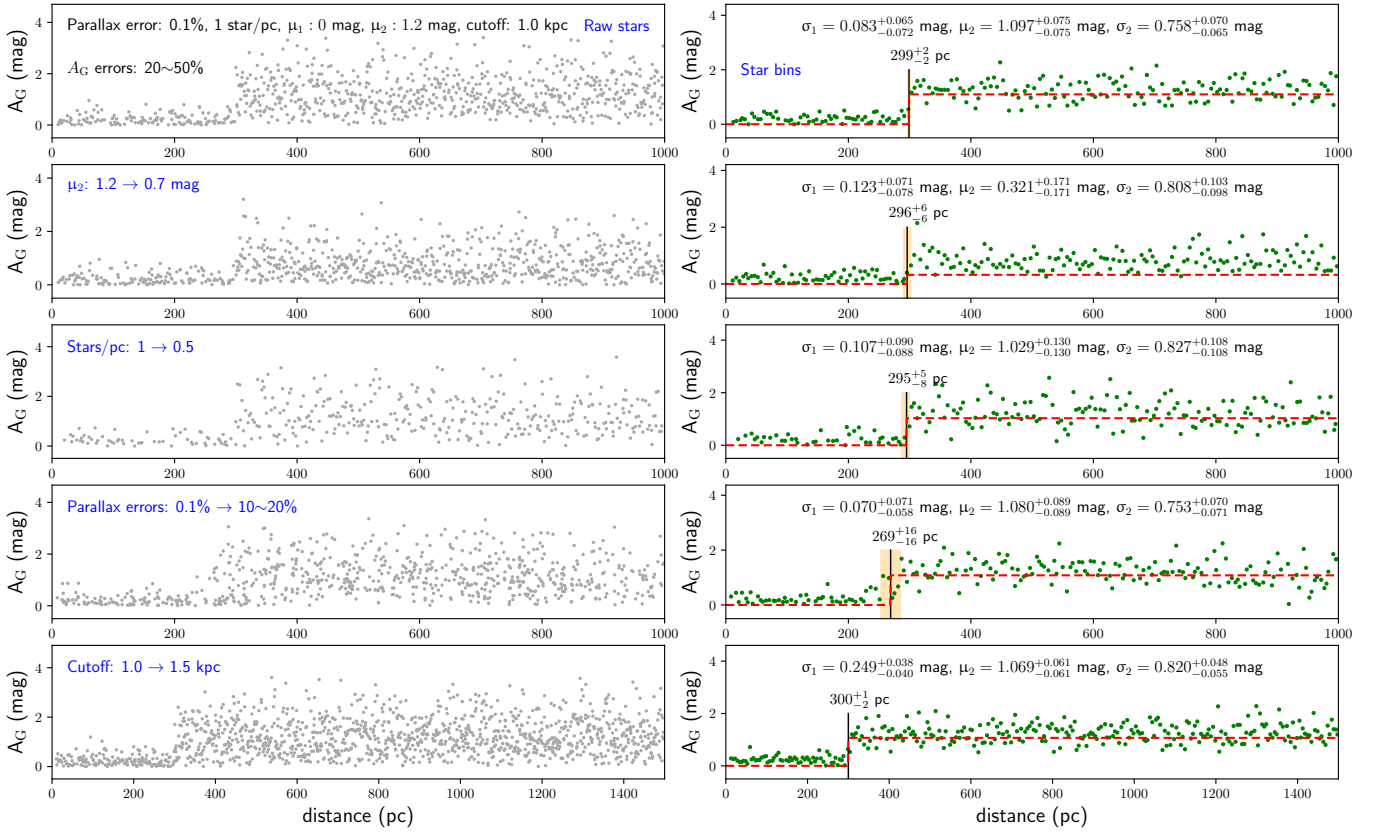


Fig. 3. Test of the model for calculating molecular cloud distances with simulated on-cloud stars. The simulated molecular cloud is assumed to be located at a distance of 300 pc. The reference values of parameters are displayed in the *top left panel*, and we change one parameter at any one time to see the variation in distance. *Left column*: simulated raw data. *Right column*: corresponding binned data and derived distance results with the model. The raw extinction A_G and distances are averaged every 5 pc, weighted by their errors. Only raw data are used in the distance estimation, while the binned data are only used to confirm the results by eye.

while the probability to be located behind the molecular cloud is $1 - f_i$.

The form of a truncated Gaussian distribution (Andrae et al. 2018), with mean μ and standard deviation σ , is

$$p(A_{Gi}|\mu, \sigma) = \begin{cases} \frac{\frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{1}{2}\left(\frac{A_{Gi}-\mu}{\sigma}\right)^2\right)}{\frac{1}{2}\left(\operatorname{erf}\left(\frac{A_G^{\max}-\mu}{\sqrt{2}\sigma}\right)+\operatorname{erf}\left(\frac{\mu-A_G^{\min}}{\sqrt{2}\sigma}\right)\right)}, & A_G^{\min} \leq A_{Gi} \leq A_G^{\max}, \\ 0, & \text{otherwise,} \end{cases} \quad (5)$$

where

$$\operatorname{erf}(z) = \frac{2}{\sqrt{\pi}} \int_0^z e^{-t^2} dt. \quad (6)$$

In order to consider the measure error of the extinction A_G , we convolved the standard deviation of the extinction A_G , ΔA_{Gi} , with σ_1 and σ_2 . Consequently, the likelihood of foreground stars is

$$PF_i = p\left(A_{Gi}|\mu_1, \sqrt{\sigma_1^2 + \Delta A_{Gi}^2}\right). \quad (7)$$

Similarly, the likelihood of background stars is

$$PB_i = p\left(A_{Gi}|\mu_2, \sqrt{\sigma_2^2 + \Delta A_{Gi}^2}\right). \quad (8)$$

Consequently, the likelihood of a star is

$$p(A_{Gi}|\mu_1, \sigma_1, \mu_2, \sigma_2, D) = f_i PF_i + (1 - f_i) PB_i. \quad (9)$$

The total likelihood is the product of the likelihoods over all the stars, and we solved this model with MCMC sampling. In order to obtain a high sampling rate, we may need to set smart priors. The prior distribution of D was assumed to be uniform, so that the model can uniformly search the switch point, that is, the breakpoint of A_G , along the line of sight. Denoting the minimum and maximum of d_i as D_{\min} and D_{\max} , the prior of D is $\mathcal{U}(D_{\min}, D_{\max} - 50 \text{ pc})$, where \mathcal{U} represents the uniform distribution and the 50 pc was set to avoid touching the edge. In practice, instead of sampling σ_1 and σ_2 , we sampled their reciprocal, denoted as $I\sigma_1$ and $I\sigma_2$. The prior distributions of $I\sigma_1$, μ_2 , and $I\sigma_2$ were assumed to be exponential, and here, we used a form of $\mathcal{E}(\beta)$ for the exponential distribution, where β is the mean. We summarize the priors as

$$\begin{cases} D \sim \mathcal{U}(D_{\min}, D_{\max} - 50 \text{ pc}), \\ I\sigma_1 \sim \mathcal{E}(2), \\ \mu_2 \sim \mathcal{E}(\mu_{50}), \\ I\sigma_2 \sim \mathcal{E}(I\sigma_{50}), \end{cases} \quad (10)$$

where \mathcal{U} and \mathcal{E} represent the uniform and exponential distributions, respectively, μ_{50} and $I\sigma_{50}$ are the mean and reciprocal standard deviation of A_G of stars with distances $>(D_{\max} - 50 \text{ pc})$, and the initial guess of 2 mag^{-1} for $I\sigma_1$ was derived from the reciprocal of 0.45 mag, which is the typical extinction A_G standard deviation of clustering stars derived by Andrae et al. (2018).

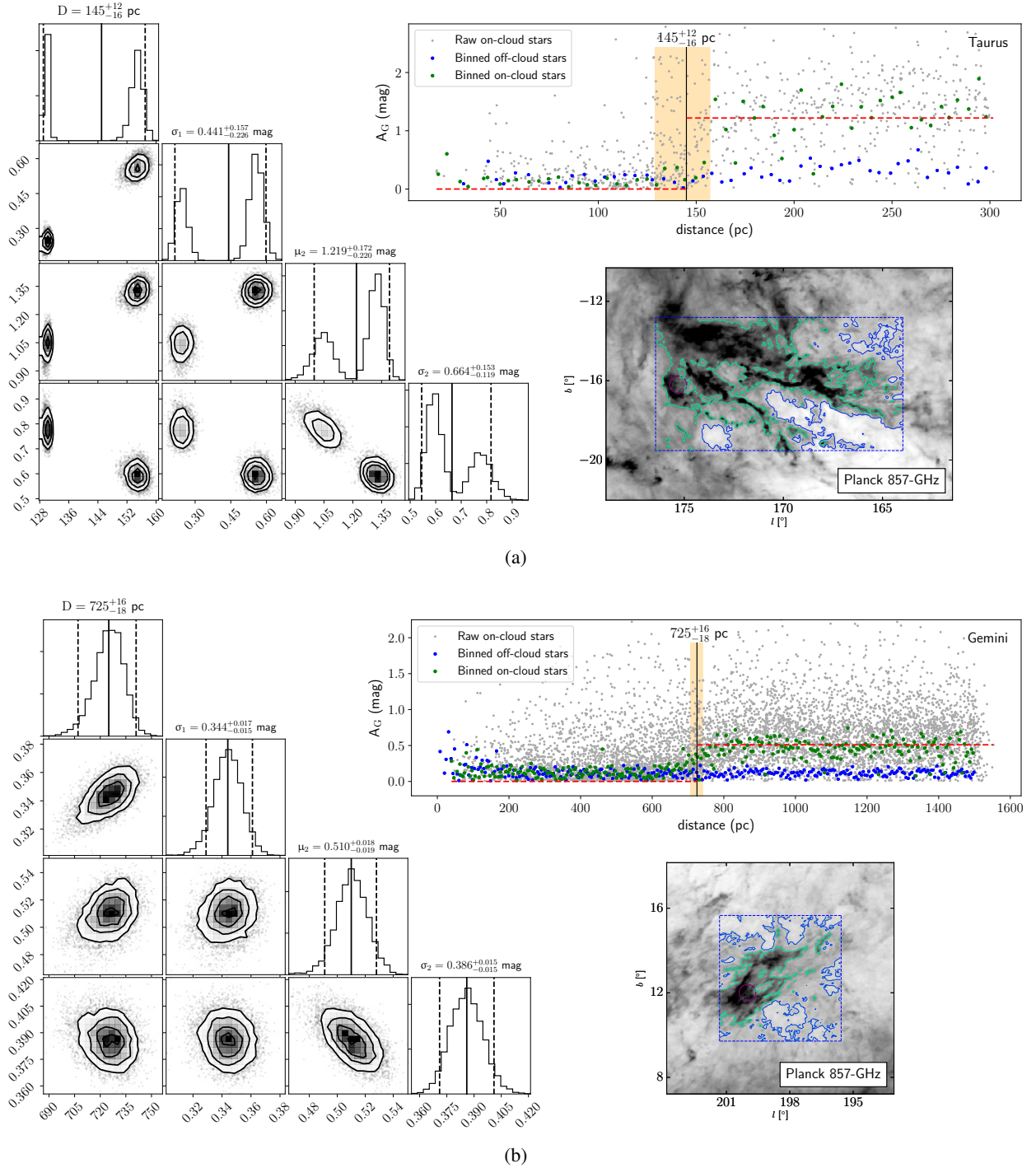


Fig. 4. Distance of the Taurus (*panel a*) and Gemini (*panel b*) molecular clouds. In the panels at the bottom right, showing the Planck 857 GHz images, purple circles mark the position of molecular clouds, while the blue and green contours correspond to the noise and signal thresholds in Fig. 2, respectively. *Top right panel:* green and blue points present on- and off-cloud stars (binned every 5 pc), respectively. The dashed red lines are the modeled extinction A_G . The distances were derived with raw on-cloud *Gaia* DR2 stars, which are represented with gray points. The black vertical lines indicate the distance (D) estimated with Bayesian analyses and MCMC sampling, and the shadow areas depict the 95% HPD distance range. The corner plots of the MCMC samples are displayed on the left. The mean and 95% HPD of the samples are shown with solid and dashed vertical lines, respectively, and the systematic uncertainty is not included. The Taurus molecular cloud may contain two components.

In order to decrease the autocorrelation time of the MCMC samples, we thinned the samples. Because the acceptance rate is about 25%, we thinned the samples by a factor of 15, considering only every 15th step in each chain, and the autocorrelation time is short (about 4) after thinning. We calculated eight independent chains, and each chain had 1000 thinned samples (with an

additional 50 for burn-in, which means that the first 50 thinned samples were removed), that is, 8000 thinned posterior samples for each parameter. In order to increase the sampling speed, we used the Gibbs sampler (Geman & Geman 1984): we changed one parameter at a time and used the Metropolis–Hastings algorithm (Metropolis et al. 1953; Hastings 1970) for each time until

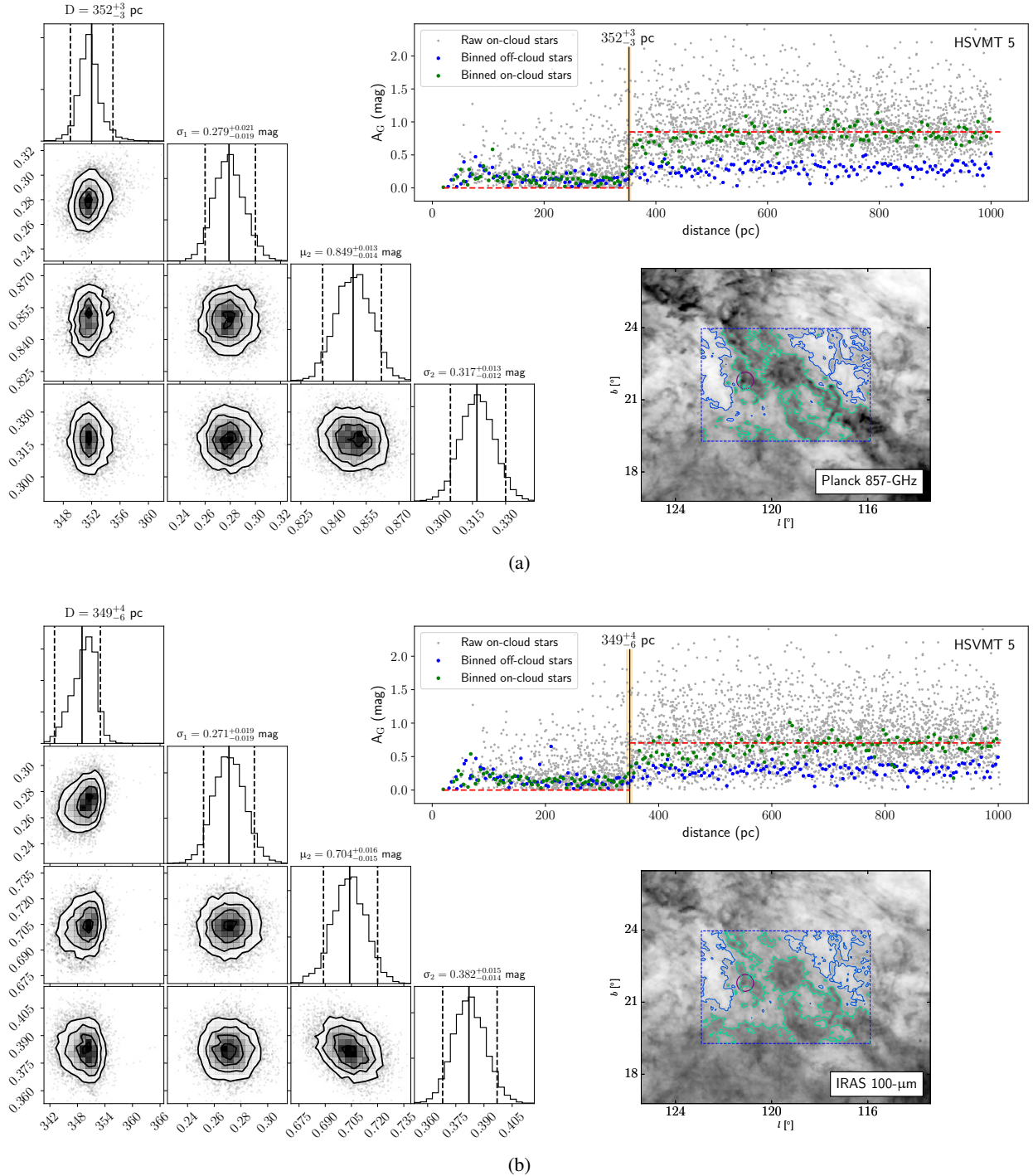


Fig. 5. Distances of molecular cloud HSVMT 5 produced with Planck 857 GHz (*panel a*) and IRAS 100 μ m (*panel b*). See the caption of Fig. 4 for other details.

all four parameters had obtained the next values. The state transition functions of the parameters are all Gaussian, whose standard deviations are 50 pc for D , 0.3 mag for μ_2 , and 0.3 mag^{-1} for $I\sigma_1$ and $I\sigma_2$.

We investigated the convergence of each chain with the Gelman–Rubin diagnostic (Gelman & Rubin 1992). The potential scale reduction factor (PSRF), \hat{R} , is smaller than 1.01 for all molecular clouds except for Taurus (whose \hat{R} is about 6), which has two close components. Consequently, the MCMC chains have converged except for Taurus.

In order to verify the MCMC results, we compared the Gibbs sampler with the affine-invariant ensemble sampler (Goodman & Weare 2010) implemented by the Python MCMC module EMCEE (Foreman-Mackey et al. 2013). The EMCEE package usually produces many outliers, but when it occasionally generates less outliers, it gave the same results with the Gibbs sampler.

In order to balance the star numbers of the two truncated Gaussian components in Eq. (5) and to avoid the interference of farther molecular clouds along the line of sight, we set a distance

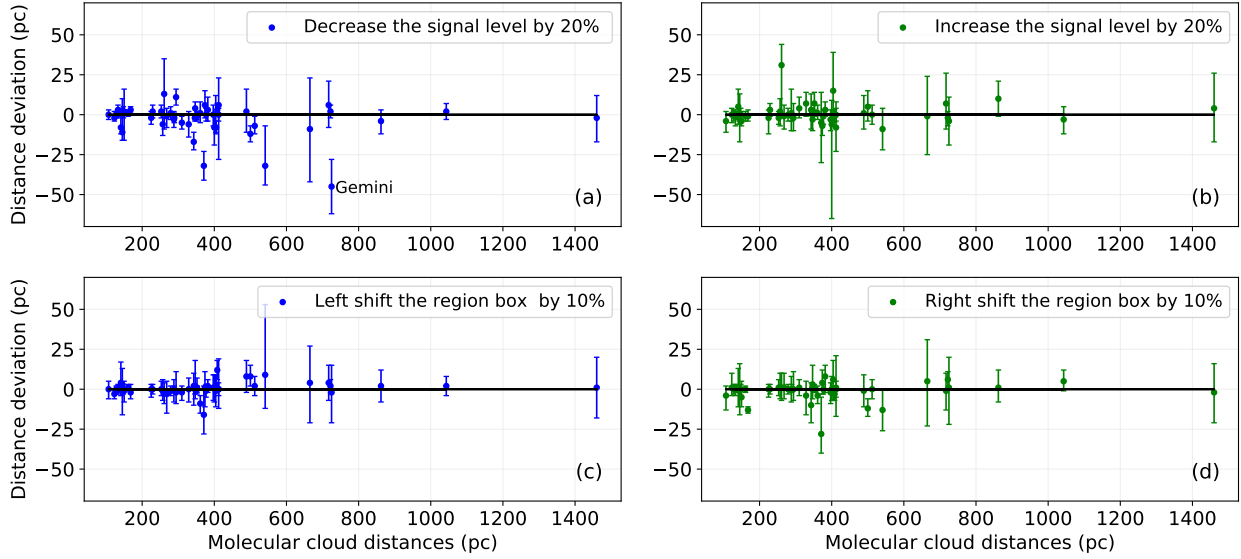


Fig. 6. Distance discrepancies and errors with altered parameters. The effect of molecular cloud region boxes is shown in *panels a and b*, and the signal levels are plotted in *panels c and d*. Only one molecular cloud, Gemini, shows large discrepancies (with magnitudes >40 pc), which is in case of panel a.

cutoff for each molecular cloud. Because we need to know the distance before setting the distance cutoff, this is a recursive process. Usually, a distance cutoff of 1000 pc is sufficient, and we adjusted this cutoff for far or near molecular clouds when necessary. Close molecular clouds usually have too few foreground stars compared to background stars, and a small fluctuation of A_G in the background stars would cause their distances to be incorrectly recognized.

A lower cutoff for foreground stars was not necessary except for four molecular clouds, which are the far component of Cepheus, Mon R2, Polaris, and Rosette. In order to remove the effect of foreground components, we removed on-cloud stars that were nearer than 400, 500, 325, and 1000 pc to them, and the corresponding μ_1 were estimated to be 0.683, 0.499, 0.295, and 0.839 mag, respectively. Because we had five parameters to model, we calculated ten chains for these four molecular clouds, that is, 10 000 thinned samples for each of them.

3.3. Testing the model

We tested this model of calculating molecular cloud distances with simulated data before we applied it to *Gaia* DR2 stars. The simulated molecular cloud is at a distance of $D = 300$ pc, and the extinction of its foreground star is 0 mag, that is, $\mu_1 = 0$. The dispersions of the foreground and background stars A_G are 0.3 and 0.6 mag, respectively, that is, $\sigma_1 = 0.3$ mag and $\sigma_2 = 0.6$ mag. We added scatter A_G errors of 20–50% on the extinction A_G , and stars whose extinction $A_G < 0$ or > 3.609 mag were removed.

We changed four parameters, the extinction of the background star μ_2 , the number of stars per pc, the relative parallax errors, and the distance cutoff, to determine the variation in the resulting distances. As demonstrated in Fig. 3, our model detected the distances successfully in all cases. Unsurprisingly, smaller jumps in A_G , fewer *Gaia* DR2 star samples, and larger parallax errors would cause larger distance errors, while the distance cutoff has no effect in the distance determination. Remarkably, although the distance errors became larger, the distance was still consistently recognized.

In *Gaia* DR2, the errors on distances are usually better than 10% for stars nearer than 1 kpc. Therefore, the two most impor-

tant factors that may affect our results are the number of on-cloud stars and the magnitude of A_G caused by molecular clouds.

3.4. Distance of the Taurus and Gemini molecular clouds

We applied our model to the Taurus and Gemini molecular clouds. The corner plots of the four parameters, D , σ_1 , μ_2 , and σ_2 , are displayed in Fig. 4, together with their means and 95% highest posterior densities (HPDs). In the bottom right panel, the dashed blue boxes of (a) and (b) are the manually chosen cloud regions (see Sect. 3.1), which contain molecular clouds and additional noise regions. The blue and green contours represent the noise and signal thresholds, respectively, which were determined in Sect. 3.1 (see Fig. 2). In the top right panel, we display the distances and on- and off-clouds stars with green and blue points, respectively. The derived distances to the Taurus and Gemini molecular clouds are 145_{-16}^{+12} and 725_{-18}^{+16} pc, respectively.

Interestingly, the Taurus molecular cloud seems to have two components. Farther away than 130 pc, many stars still have low extinction A_G until about 150 pc. The Taurus distance given by Torres et al. (2009) is 161.2 ± 0.9 pc, which may be corresponding to this component.

4. Molecular cloud distances

4.1. Planck 857 GHz versus IRAS 100 μm data

We demonstrate that Planck 857 GHz data trace dust better than IRAS 100 μm , or at least that these data better suited to our method. As an example, we show the distances of the molecular cloud HSVMT 5 produced with Planck 857 GHz and IRAS 100 μm in Fig. 5.

The distance estimated with Planck 857 GHz is 352_{-3}^{+4} pc, and 349_{-6}^{+4} pc for IRAS 100 μm . As indicated in Fig. 5, the average extinction of on-cloud stars classified with Planck 857 GHz is higher than that with IRAS 100 μm , suggesting that the on-cloud stars identified with Planck 857 GHz are less strongly contaminated by other *Gaia* stars that have low extinction A_G . Consequently, the extinction values, A_G , of the background

Table 1. Comparison with VLBI distance measurements.

Cloud	<i>Gaia</i> DR2 (pc)	VLBI (pc)	References
L1641	408 ⁺⁴ ₋₄	428±10	1
NGC 2068	412 ⁺⁴ ₋₄	388±10	1
Mon R2	862 ⁺¹⁰ ₋₁₀	893 ⁺⁴⁴ ₋₄₀	2
Ophiuchus	131 ⁺² ₋₂	137.3±1.2	3
Perseus	310 ⁺⁴ ₋₄	293±22 or 321±10	4

Notes. In Perseus, [Ortiz-León et al. \(2018\)](#) obtained a distance of 321±10 pc (IC 348) with VLBI, while the distance of NGC 1333 is 293±22 pc, as suggested for young stars in *Gaia* DR2.

References. (1) [Kounkel et al. \(2017\)](#); (2) [Dzib et al. \(2016\)](#); (3) [Ortiz-León et al. \(2017b\)](#); (4) [Ortiz-León et al. \(2018\)](#).

stars are better separated from the foreground stars with Planck 857 GHz emission.

4.2. Effect of the parameter choice

In this subsection, we examine the reliability of our method and its dependence on parameters. In our method, we primarily involve two subjective parameters: the manually chosen molecular cloud region, and the signal level cutoff in the Planck 857 GHz intensity. In total, we found that 52 molecular clouds have well-determined distances with the chosen molecular cloud regions and signal levels.

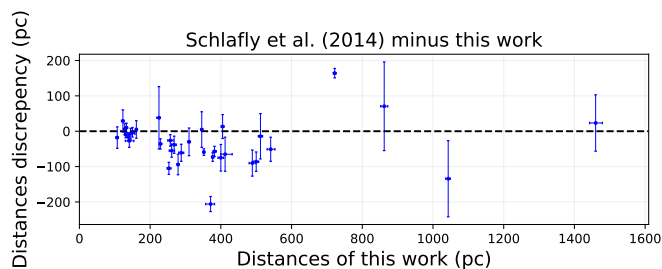
In order to determine the influence of the two parameters, we shifted them one at a time and compared the distance variations. In Fig. 6 we display the distance discrepancies and uncertainties after altering parameters. To reveal the effect of the molecular cloud region box, we shifted the molecular cloud region box along l to the left and right by 10% of the box size in l , while the signal levels were altered by ±20%. The distances show slight fluctuations in the molecular clouds, while their uncertainties widely agree with the distances calculated with the chosen parameters. Clearly, large distance discrepancies usually yield large uncertainties, and no evident systematic deviations are present. Only Gemini shows large distance deviations (with magnitudes >40 pc) in case (a), confirming our claim that correct recolonization of breakpoints guarantees comparable distances.

Consequently, our distance estimation is robust and only weakly dependent on the choice of region boxes and signal levels in a reasonable range. In addition, in order to ensure that the breakpoints are genuine, it is necessary to confirm the breakpoints with off-cloud stars.

4.3. Distance catalog

With the method described in Sect. 3, we calculated distances for many molecular clouds at high Galactic latitude, most of which have been cataloged by [Magnani et al. \(1985, 1996\)](#) and [Schlafly et al. \(2014\)](#). In these three catalogs, many molecular clouds belong to the same molecular cloud complexes or filaments, and we only give one distance for each complex or filament because of the large dispersion in A_G and because *Gaia* stars are not suitable for small subregions.

We have reproduced the distances of many star-forming regions, whose distances are well determined with VLBI measurements, such as Orion ([Kounkel et al. 2017](#)), Mon R2 ([Dzib et al. 2016](#)), Ophiuchus ([Ortiz-León et al. 2017b](#)), and Perseus

**Fig. 7.** Comparison of distances in our work with those from [Schlafly et al. \(2014\)](#).

([Ortiz-León et al. 2018](#)). In Table 1 we compare our results with VLBI-measured distances. Considering the distance errors and dispersions, the molecular cloud distance given by *Gaia* DR2 agrees quite well with VLBI measurements. We suggest that VLBI and *Gaia* DR2 may see different components of the molecular clouds, and the slight discrepancies may be due to the structure of the molecular clouds.

We summarize the distances in Table A.1, listing 52 molecular clouds whose distances are well determined, that is, whose breakpoints are evident. The distances for many molecular clouds cannot be determined because (1) they are not well defined in Planck 857 GHz, (2) their optical depths are too low, or (3) their covering areas are too small. In the penultimate column, we indicate the molecular clouds whose distances are not provided by [Schlafly et al. \(2014\)](#) or were measured using VLBI with “Y”. There are 13 such clouds in total.

The distance figures of all molecular clouds are provided on Harvard Dataverse¹.

4.4. Systematic error

In this section, we estimate the systematic error in the distances. First, as suggested by the test results (see Fig. 3), the distances show a systematic error of 1–10%. In the testing data, we simulated parallax errors of 10–20% and within 1 kpc from the Sun, but the uncertainties of *Gaia* DR2 parallaxes are smaller than 10%. Therefore, the true systematic error is likely to be smaller.

Second, as shown in Fig. 6, the root mean squares (RMS) of the relative distance deviations in the four cases are 3% (a), 2% (b), 1% (c), and 2% (d), respectively. This suggests a systematic error of about 3% that is caused by subjective choices.

Third, if the VLBI results (see Table 1) are the true distances, our distances show a systematic error of ≤6%. However, VLBI usually measures the parallax of one single star, which may deviate from the main part of the molecular clouds.

Fourth, [Lindgren et al. \(2018\)](#) reported that the *Gaia* DR2 parallaxes have a systematic error of approximately 0.1 mas. Considering a molecular cloud at a distance of 500 pc (2 mas), this systematic parallax error causes a systematic error of about 5% in the distance.

Consequently, we estimate that the systematic distance error is about 5%. It might be slightly larger for distant molecular clouds.

5. Discussion

5.1. Comparison with previous results

In Fig. 7 we draw the distances derived from *Gaia* DR2 against those provided by [Schlafly et al. \(2014\)](#). We averaged the dis-

¹ <https://doi.org/10.7910/DVN/C6Y04T>

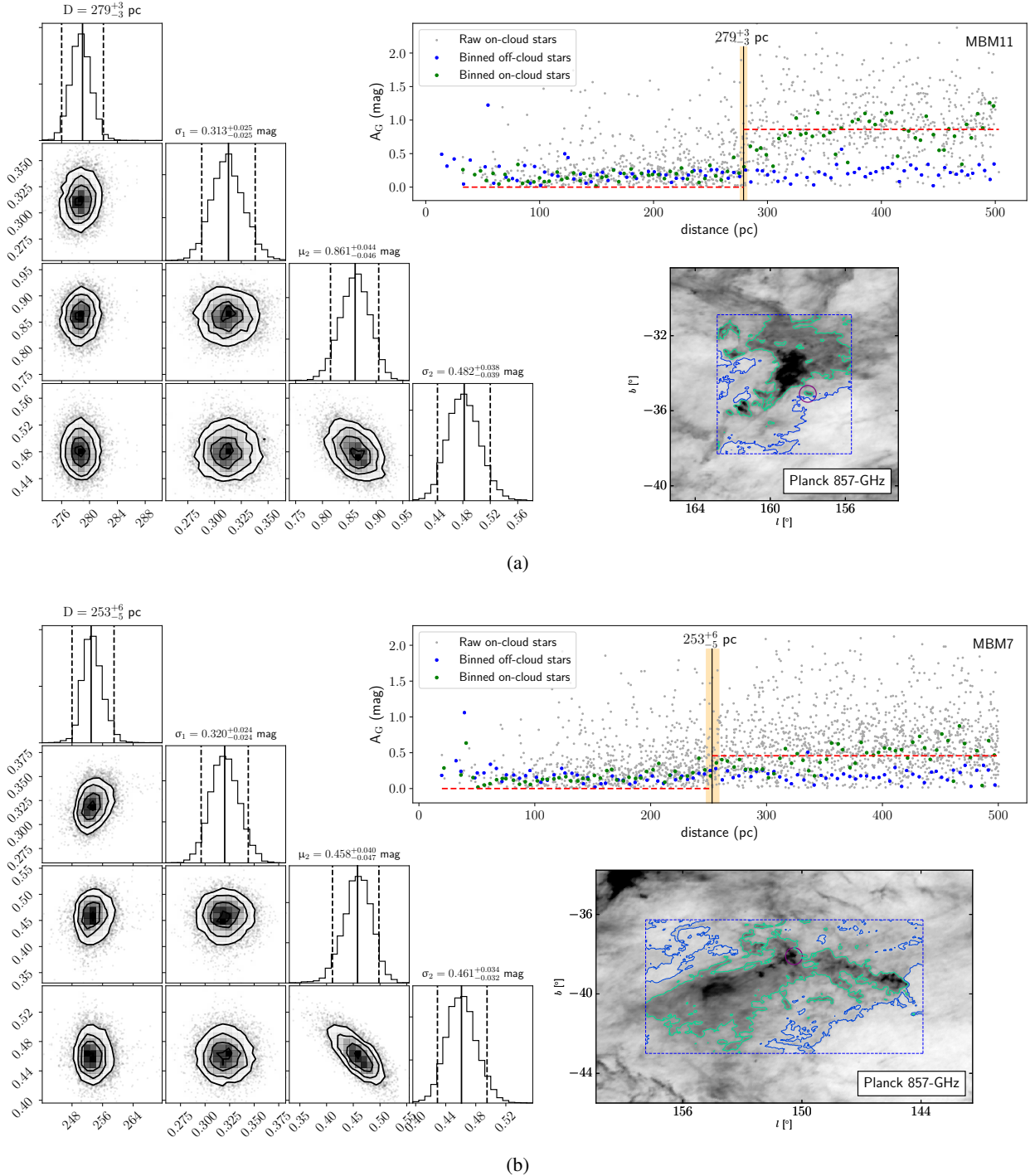


Fig. 8. Distance of MBM11 (*panel a*) and MBM7 (*panel b*). These two molecular clouds show large distance discrepancies with that provided by [Schlafly et al. \(2014\)](#). See the caption of Fig. 4 for other details.

tances weighted by the square inverse of their errors when multiple distances (corresponding to the same molecular cloud) are provided by [Schlafly et al. \(2014\)](#). The distances provided by [Schlafly et al. \(2014\)](#) display a slight systematic shift, ~ 24 pc, toward the farther distances. In terms of relative errors, the systematic shift is, after removing one outliers ($>70\%$), about 13%.

This systematic shift was predicted by [Schlafly et al. \(2014\)](#). They attributed the systematic errors to stellar models, dust models, and the reddening law. However, in addition to these, we

found that the molecular regions used by [Schlafly et al. \(2014\)](#) are much smaller than the regions in our work, which means that their results may only represent the distances of molecular clouds in these small regions and are less affected by the thickness of the molecular clouds along the line of sight. Alternatively, nearer clouds are brighter and easily chosen, which could cause the distances given by [Schlafly et al. \(2014\)](#) to be systematically smaller.

However, the *Gaia* DR2 data may also contribute to this systematic error. The ~ 0.1 mas systematic parallax error in *Gaia*

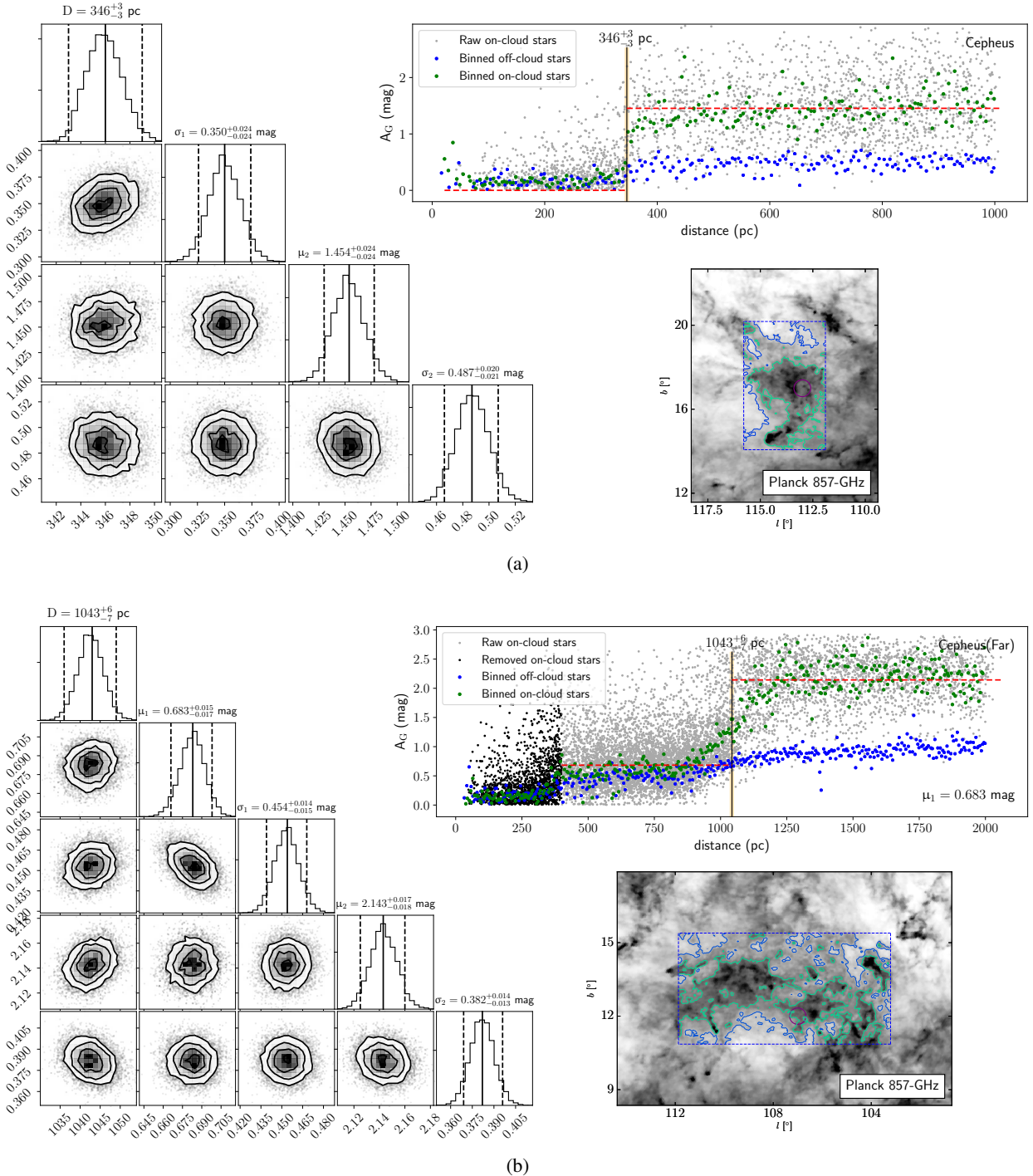


Fig. 9. Distances of near (*panel a*) and far (*panel b*) components in the Cepheus molecular cloud. See the caption of Fig. 4 for other details.

DR2 (Lindegren et al. 2018) may be one of the causes of the systematic distance discrepancy. At a typical distance of 500 pc, 0.1 mas error corresponds to ~ 25 pc shift, which can explain this systematic shift. Furthermore, the extinction A_G , which has large uncertainties, may contain unknown systematic errors.

One means of examining the discrepancy is to compare the distance modulus in Schlafly et al. (2014) with the *Gaia* DR2 parallaxes. However, this is beyond the scope of this work, and alternatively, in Fig. 8, we display the distances of MBM11 and MBM7, which show large distance discrepancies. As shown in Fig. 8, the distances of MBM11 and MBM7 are 279^{+3}_{-3} and 253^{+6}_{-5} pc, respectively. However, in the distance catalog of

Schlafly et al. (2014), the distance of the MBM11 molecular cloud is approximately 206 ± 23 pc, which is the average distance of MBM11, 12, 13, and 14, while the distance of MBM7 is 148^{+13}_{-11} pc. Based on *Gaia* DR2, the distances of these two molecular clouds, particularly MBM11, are well determined.

5.2. Individual molecular cloud distances

In this subsection, we discuss the distances of several individual molecular clouds. Schlafly et al. (2014) obtained a distance of about 350 pc for the Ursa Major molecular clouds, which is much farther than previous studies, ~ 110 pc (Penprase 1993). Using

Gaia DR2 data, we have derived a distance of 412^{+20}_{-12} for the Ursa Major molecular clouds, which is close to the distance given by Schlafly et al. (2014).

Li et al. (2015) performed a survey of two CO isotopologue lines toward the same Gemini region as shown in Fig. 4. They used a distance of 400 pc, but *Gaia* DR2 clearly shows that its distance is about 725 pc. The masses and sizes of Gemini molecular cores calculated by Li et al. (2015) should be revised accordingly.

Cepheus (Grenier et al. 1989; Yonekura et al. 1997; Kirk et al. 2009) is an interesting region. As shown in Fig. 9, there are two obvious components along the line of sight. The distance of the nearer component is 346^{+3}_{-3} pc, and the distance of the farther is 1043^{+6}_{-7} pc. According to the CO observations of Grenier et al. (1989), the radial velocity range of the far component is about -12 km s^{-1} , while the nearer component has a velocity of about 0 km s^{-1} . The distances of the two components derived by Grenier et al. (1989) are ~ 300 and 800 pc, respectively, which are consistent with our results.

6. Summary

Using the parallaxes and extinction A_G provided by *Gaia* DR2, we derived the distances for 52 molecular clouds, most of which are at high Galactic latitudes, that is, $|b| > 10^\circ$. The systematic error of the distances is about 5%, and we determined reliable distances for 13 molecular clouds for the first time. In addition, we have confirmed the distances of many star-forming regions, such as Orion, Taurus, Cepheus, and Mon R2.

We used Planck 857 GHz data rather than CO data to trace molecular clouds because CO observations are incomplete at high Galactic latitudes. However, at low Galactic latitudes ($|b| \leq 10^\circ$), multiwavelength observations with high spatial resolutions would classify on- and off-cloud stars more accurately.

Gaia DR2 has enabled us to determine the distances of many nearby molecular clouds efficiently. Although the large errors of extinction A_G in *Gaia* DR2 prevent us from examining distant (>2 kpc) molecular clouds, with the improved qualities of distances and extinction A_G in future *Gaia* data releases, we would be able to obtain the distances of many more molecular clouds.

Acknowledgements. We thank Mario G. Lattanzi for his valuable comments and careful proofreading. We are grateful to an anonymous referee for their constructive comments, particularly on the distance likelihood and the MCMC method. This work was sponsored by the Ministry of Science and Technology (MOST) Grant No. 2017YFA0402701, the 100 Talents Project of the Chinese Academy

of Sciences (CAS), the National Science Foundation of China under Grand NO. 11673051 and 11873019, the CAS Grand No. QYZDJ-SSW-SLH047, and the Key Laboratory for Radio Astronomy, CAS.

References

- Andrae, R., Fouesneau, M., Creevey, O., et al. 2018, *A&A*, 616, A8
 Arenou, F., Luri, X., Babusiaux, C., et al. 2018, *A&A*, 616, A17
 Bailer-Jones, C. A. L. 2015, *PASP*, 127, 994
 Bok, B. J. 1937, *The Distribution of the Stars in Space* (Chicago: University of Chicago Press)
 Dame, T. M., Hartmann, D., & Thaddeus, P. 2001, *ApJ*, 547, 792
 Dzib, S. A., Ortiz-León, G. N., Loinard, L., et al. 2016, *ApJ*, 826, 201
 Foreman-Mackey, D., Hogg, D. W., Lang, D., & Goodman, J. 2013, *PASP*, 125, 306
 Gaia Collaboration (Prusti, T., et al.) 2016, *A&A*, 595, A1
 Gaia Collaboration (Brown, A. G. A., et al.) 2018, *A&A*, 616, A1
 Gelman, A., & Rubin, D. B. 1992, *Stat. Sci.*, 7, 457
 Geman, S., & Geman, D. 1984, *IEEE Trans. Pattern Anal. Mach. Intell.*, 6, 721
 Goodman, J., & Weare, J. 2010, *Commun. Appl. Math. Comput. Sci.*, 5, 65
 Green, G. M., Schlafly, E. F., Finkbeiner, D. P., et al. 2014, *ApJ*, 783, 114
 Grenier, I. A., Lebrun, F., Arnaud, M., Dame, T. M., & Thaddeus, P. 1989, *ApJ*, 347, 231
 Hastings, W. K. 1970, *Biometrika*, 57, 97
 Kaiser, N., Burgett, W., Chambers, K., et al. 2010, *Proc. SPIE*, 7733, 77330E
 Kirk, J. M., Ward-Thompson, D., Di Francesco, J., et al. 2009, *ApJS*, 185, 198
 Knude, J., & Hog, E. 1998, *A&A*, 338, 897
 Kounkel, M., Hartmann, L., Loinard, L., et al. 2017, *ApJ*, 834, 142
 Li, Y., Xu, Y., Yang, J., et al. 2015, *AJ*, 150, 60
 Lindgren, L., Hernández, J., Bombrun, A., et al. 2018, *A&A*, 616, A2
 Magnani, L., & de Vries C. P. 1986, *A&A*, 168, 271
 Magnani, L., Blitz, L., & Mundy, L. 1985, *ApJ*, 295, 402
 Magnani, L., Hartmann, D., & Speck, B. G. 1996, *ApJS*, 106, 447
 Metropolis, N., Rosenbluth, A. W., Rosenbluth, M. N., Teller, A. H., & Teller, E. 1953, *J. Chem. Phys.*, 21, 1087
 Miville-Deschênes, M.-A., & Lagache, G. 2005, *ApJS*, 157, 302
 Ortiz-León, G. N., Dzib, S. A., Kounkel, M. A., et al. 2017a, *ApJ*, 834, 143
 Ortiz-León, G. N., Loinard, L., Kounkel, M. A., et al. 2017b, *ApJ*, 834, 141
 Ortiz-León, G. N., Loinard, L., Dzib, S. A., et al. 2018, *ApJ*, 865, 73
 Penprase, B. E. 1993, *ApJS*, 88, 433
 Pineda, J. L., Goldsmith, P. F., Chapman, N., et al. 2010, *ApJ*, 721, 686
 Planck Collaboration I. 2014, *A&A*, 571, A1
 Rice, T. S., Goodman, A. A., Bergin, E. A., Beaumont, C., & Dame, T. M. 2016, *ApJ*, 822, 52
 Roman-Duval, J., Jackson, J. M., Heyer, M., et al. 2009, *ApJ*, 699, 1153
 Schlafly, E. F., Green, G., Finkbeiner, D. P., et al. 2014, *ApJ*, 786, 29
 Schlegel, D. J., Finkbeiner, D. P., & Davis, M. 1998, *ApJ*, 500, 525
 Torres, R. M., Loinard, L., Mioduszewski, A. J., & Rodríguez, L. F. 2009, *ApJ*, 698, 242
 Ward-Thompson, D., Pattle, K., Kirk, J. M., et al. 2016, *MNRAS*, 463, 1008
 Xu, Y., Reid, M. J., Menten, K. M., et al. 2009, *ApJ*, 693, 413
 Yonekura, Y., Dobashi, K., Mizuno, A., Ogawa, H., & Fukui, Y. 1997, *ApJS*, 110, 21
 Zhang, B., Reid, M. J., Menten, K. M., et al. 2013, *ApJ*, 775, 79

Appendix A: Additional table

Table A.1. Molecular cloud distances.

Cloud ^a	l^a (°)	b^a (°)	D^b (pc)	$\sigma_{l,b}$ (mag)	$\mu_{2,b}$ (mag)	$\sigma_{2,b}$ (mag)	Center ^c (°, °)	Extension ^c (°, °)	Noise ^d (mJy sr ⁻¹)	Signal ^d (mJy sr ⁻¹)	Cutoff ^e (pc)	N^e (13)	D_{Schlafly}^f (pc)	Note ^g (15)
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)
MBM155	1.6	-21.3	148 ⁺³ ₋₃	0.146 ^{+0.028} _{-0.030}	0.378 ^{+0.083} _{-0.088}	0.473 ^{+0.057} _{-0.063}	(0.7, -20.8)	(5.0, 8.7)	3.9	8.3	300	626	Y	MBM154,153
MBM35	6.6	38.1	107 ⁺³ ₋₄	0.251 ^{+0.047} _{-0.052}	0.502 ^{+0.057} _{-0.063}	0.438 ^{+0.049} _{-0.049}	(5.4, 38.1)	(6.6, 8.2)	4.2	8.2	300	617	89 ⁺¹⁷ ₋₂₅	MBM38,36,37,34,CB 63
MBM145	8.5	21.9	123 ⁺³ ₋₄	0.282 ^{+0.050} _{-0.050}	0.758 ^{+0.058} _{-0.055}	0.503 ^{+0.048} _{-0.047}	(11.3, 23.2)	(9.2, 7.1)	10.8	16.4	300	677	152 ⁺¹⁹ ₋₂₅	MBM146,149,148,139,150
MBM151	21.5	20.9	138 ⁺² ₋₂	0.279 ^{+0.038} _{-0.036}	0.862 ^{+0.047} _{-0.048}	0.472 ^{+0.039} _{-0.044}	(20.8, 21.8)	(6.4, 9.4)	5.9	11.7	300	766	122 ⁺⁸ ₋₈	MBM158
MBM159	27.4	-21.1	133 ⁺³ ₋₃	0.183 ^{+0.035} _{-0.039}	0.330 ^{+0.070} _{-0.065}	0.430 ^{+0.049} _{-0.049}	(27.2, -21.3)	(6.5, 9.4)	5.1	7.3	300	786	143 ⁺⁸ ₋₁₀	MBM158
Aquila S	38.0	-17.0	141 ⁺¹³ ₋₁₁	0.395 ^{+0.064} _{-0.062}	0.666 ^{+0.044} _{-0.046}	0.394 ^{+0.038} _{-0.036}	(38.0, -17.9)	(6.1, 7.7)	5.0	8.9	300	732	114±18	MBM47,48
MBM46	40.5	-35.5	541 ⁺¹² ₋₁₀	0.335 ^{+0.016} _{-0.015}	0.425 ^{+0.022} _{-0.021}	0.364 ^{+0.017} _{-0.018}	(41.7, -35.4)	(8.6, 5.2)	2.7	4.9	1000	5130	490 ⁺²⁵ ₋₂₃	MBM47,48
MBM160	44.0	-23.3	717 ⁺¹¹ ₋₁₁	0.379 ^{+0.009} _{-0.009}	0.360 ^{+0.008} _{-0.008}	0.319 ^{+0.007} _{-0.007}	(45.9, -21.3)	(6.7, 10.2)	3.4	5.0	1500	22411	Y	Y
Hercules	44.0	8.6	229 ⁺⁴ ₋₃	0.303 ^{+0.040} _{-0.040}	1.171 ^{+0.054} _{-0.052}	0.475 ^{+0.045} _{-0.048}	(45.5, 9.1)	(6.0, 2.6)	12.4	16.9	500	575	193±15	MBM53
Pegasus	92.2	-34.7	268 ⁺⁶ ₋₈	0.251 ^{+0.026} _{-0.030}	0.457 ^{+0.054} _{-0.056}	0.439 ^{+0.041} _{-0.044}	(91.6, -32.8)	(4.5, 6.8)	2.5	5.4	500	1175	230±24	MBM53
MBM54	93.0	-37.5	257 ⁺⁸ ₋₇	0.257 ^{+0.037} _{-0.033}	0.477 ^{+0.069} _{-0.068}	0.415 ^{+0.056} _{-0.058}	(94.5, -37.8)	(5.8, 3.0)	2.7	5.4	500	591	231 ⁺¹¹ ₋₁₂	MBM53
Lacerta	96.0	-12.0	512 ⁺⁶ ₋₆	0.364 ^{+0.021} _{-0.019}	0.757 ^{+0.021} _{-0.021}	0.392 ^{+0.019} _{-0.019}	(96.1, -12.3)	(3.1, 6.5)	5.5	8.9	1000	2758	498±64	Y
MBM56	103.1	-26.1	288 ⁺⁷ ₋₇	0.334 ^{+0.031} _{-0.029}	0.390 ^{+0.022} _{-0.022}	0.351 ^{+0.019} _{-0.019}	(102.5, -27.4)	(5.4, 5.2)	2.9	4.5	1000	3122	227 ⁺¹⁷ ₋₁₇	G102-27
MBM157	103.2	22.7	382 ⁺⁵ ₋₄	0.285 ^{+0.014} _{-0.014}	0.463 ^{+0.012} _{-0.012}	0.344 ^{+0.010} _{-0.010}	(100.3, 23.8)	(9.5, 7.0)	2.0	4.2	1000	8883	325 ⁺¹¹ ₋₉	MBM156
Cepheus(Far) ^h	107.0	12.0	1043 ⁺⁶ ₋₇	0.454 ^{+0.014} _{-0.015}	2.143 ^{+0.017} _{-0.018}	0.382 ^{+0.014} _{-0.013}	(107.5, 13.1)	(8.7, 4.5)	16.1	24.2	2000	9048	909±108	Y
MBM162	111.7	20.1	377 ⁺³ ₋₃	0.399 ^{+0.025} _{-0.025}	0.971 ^{+0.019} _{-0.018}	0.374 ^{+0.017} _{-0.017}	(111.7, 21.3)	(4.4, 5.1)	4.8	9.1	1000	2523	304 ⁺⁸ ₋₉	Y
Cepheus	113.0	17.0	346 ⁺³ ₋₃	0.350 ^{+0.024} _{-0.024}	1.454 ^{+0.024} _{-0.024}	0.487 ^{+0.020} _{-0.021}	(113.8, 17.1)	(3.9, 6.1)	8.4	14.5	1000	2347	351±50	Y
MBM163	115.8	20.2	352 ⁺² ₋₂	0.273 ^{+0.018} _{-0.019}	0.840 ^{+0.012} _{-0.013}	0.339 ^{+0.012} _{-0.012}	(117.0, 21.8)	(7.2, 5.8)	4.7	7.6	1000	4266	293 ⁺⁶ ₋₇	MBM161,166,165,164
MBM2	117.4	-52.3	261 ⁺⁸ ₋₇	0.215 ^{+0.020} _{-0.020}	0.339 ^{+0.025} _{-0.023}	0.411 ^{+0.016} _{-0.020}	(111.5, -51.2)	(17.6, 7.2)	2.8	4.2	1000	4791	206 ⁺¹⁴ ₋₁₂	Y
HSVMT 03	120.5	29.6	374 ⁺⁹ ₋₉	0.344 ^{+0.028} _{-0.028}	0.596 ^{+0.019} _{-0.017}	0.325 ^{+0.018} _{-0.016}	(118.9, 29.0)	(5.7, 5.5)	3.8	5.5	1000	2576	Y	HSVMT 01
HSVMT 5	121.1	21.8	352 ⁺³ ₋₃	0.279 ^{+0.021} _{-0.019}	0.849 ^{+0.013} _{-0.014}	0.317 ^{+0.013} _{-0.012}	(119.4, 21.6)	(7.0, 4.7)	4.6	7.5	1000	3549	Y	MBM166,165,164
HSVMT 08	122.5	29.0	347 ⁺⁸ ₋₇	0.235 ^{+0.043} _{-0.044}	0.759 ^{+0.030} _{-0.031}	0.295 ^{+0.030} _{-0.029}	(122.3, 29.9)	(2.7, 3.3)	4.8	7.0	1000	623	Y	Y
HSVMT 12	125.2	32.5	361 ⁺⁵ ₋₇	0.261 ^{+0.015} _{-0.016}	0.550 ^{+0.017} _{-0.019}	0.427 ^{+0.014} _{-0.016}	(127.7, 29.6)	(6.8, 9.1)	1.4	4.1	1000	5868	Y	HSVMT 14
Polaris	126.0	21.2	489 ⁺¹⁰ ₋₉	0.466 ^{+0.068} _{-0.064}	0.769 ^{+0.018} _{-0.017}	0.319 ^{+0.015} _{-0.017}	(126.6, 20.7)	(4.7, 4.2)	4.2	7.7	1000	2640	399±37	Y
Ursa Major	143.0	38.0	412 ⁺²⁰ ₋₁₂	0.281 ^{+0.023} _{-0.021}	0.389 ^{+0.035} _{-0.038}	0.411 ^{+0.028} _{-0.028}	(142.3, 36.1)	(4.9, 7.7)	1.3	3.2	1000	2411	347±48	MBM29,28,27,30

Notes. ^(a)Molecular clouds (Col. 1) and their Galactic coordinates (Cols. 2 and 3), most of which are cataloged by Magnani et al. (1985, 1996) and Schlafly et al. (2014). ^(b)The mean and 95% HPD errors of the distance (Col. 4), the extinction A_G dispersion of foreground stars (Col. 5), the extinction A_G of background stars (Col. 6), and their dispersion (Col. 7). The distance systematic error is about 5%, which is not included in this table. ^(c)The center (Col. 8) and angular extinction (Col. 9) of region boxes in galactic coordinates, not their actual angular sizes. Those region boxes contain both on- and off-cloud *Gaia* DR2 stars. ^(d)The noise (Col. 10) and signal threshold (Col. 11) used to classify on- and off-molecular clouds. ^(e)The number of on-cloud *Gaia* stars (Col. 13) nearer than the distance cutoff (Col. 12). ^(f)Distances in Schlafly et al. (2014; Col. 14). Because distances of multiple positions are provided for molecular clouds in Table 1 of Schlafly et al. (2014), we averaged the distances of those molecular clouds weighted by their errors. Molecular clouds that not measured by Schlafly et al. (2014) or other reliable methods are marked with Y, 13 in total. ^(g)Other molecular clouds covered by the region box (Col. 15). ^(h)The far component towards the Cepheus molecular clouds.

Table A.1. continued.

Cloud ^a	l^a (°)	b^a (°)	D^b (pc)	σ_1^b (mag)	μ_2^b (mag)	σ_2^b (mag)	Center ^c (°, °)	Extension ^c (°, °)	Noise ^d (mJy sr ⁻¹)	Signal ^d (mJy sr ⁻¹)	Cutoff (pc)	N^e (13)	D_{Schlafly}^f (pc)	Note ^g
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)
MBM31	146.4	39.6	400 ⁺⁸ ₋₈	0.306 ^{+0.032} _{-0.030}	0.534 ^{+0.038} _{-0.036}	0.379 ^{+0.034} _{-0.031}	(147.2, 39.2)	(4.0, 4.7)	2.6	4.3	1000	1145	325 ⁺²⁷ ₋₂₆	MBM32, HSNVT 24
Cam	147.5	17.8	225 ⁺⁴ ₋₆	0.248 ^{+0.034} _{-0.034}	0.429 ^{+0.054} _{-0.043}	0.428 ^{+0.044} _{-0.043}	(146.6, 18.0)	(6.6, 4.1)	4.5	6.4	500	995	263 ^{±88}	
MBM7	150.4	-38.1	253 ⁺⁶ ₋₅	0.320 ^{+0.024} _{-0.024}	0.458 ^{+0.040} _{-0.047}	0.461 ^{+0.034} _{-0.032}	(150.9, -39.6)	(14.0, 6.7)	3.6	5.9	500	2080	148 ⁺¹³ ₋₁₁	MBM8, 6.G154.7-39.8
HSNVT 27	153.6	36.9	397 ⁺⁸ ₋₈	0.274 ^{+0.020} _{-0.021}	0.353 ^{+0.035} _{-0.034}	0.375 ^{+0.026} _{-0.027}	(152.9, 37.4)	(6.0, 6.0)	2.1	3.3	1000	2204	Y	
MBM26	156.4	32.6	404 ⁺¹⁰ ₋₁₀	0.297 ^{+0.024} _{-0.026}	0.354 ^{+0.043} _{-0.046}	0.421 ^{+0.030} _{-0.034}	(158.1, 33.8)	(6.1, 4.8)	2.1	3.3	1000	1802	Y	HSNVT 28
California	157.0	-12.0	500 ⁺⁷ ₋₇	0.742 ^{+0.034} _{-0.030}	2.576 ^{+0.024} _{-0.020}	0.209 ^{+0.021} _{-0.020}	(161.2, -8.2)	(8.2, 2.9)	18.9	30.3	1000	1641	414 ^{±27}	
MBM11	158.0	-35.1	279 ⁺³ ₋₃	0.313 ^{+0.025} _{-0.025}	0.861 ^{+0.044} _{-0.046}	0.482 ^{+0.038} _{-0.039}	(159.3, -34.6)	(7.2, 7.4)	4.5	8.9	500	1206	185 ⁺²¹ ₋₂₀	MBM13, 12, 14
UT 8d	158.2	-26.3	289 ⁺³ ₋₂	0.368 ^{+0.027} _{-0.024}	1.318 ^{+0.020} _{-0.023}	0.469 ^{+0.018} _{-0.018}	(158.1, -24.0)	(5.5, 7.8)	7.1	13.0	1000	2628	Y	MBM101, 103, 102, 104
Perseus	160.0	-20.0	310 ⁺⁴ ₋₄	0.595 ^{+0.046} _{-0.049}	2.059 ^{+0.039} _{-0.041}	0.460 ^{+0.034} _{-0.034}	(159.8, -19.3)	(4.4, 4.9)	10.2	25.1	1000	980	280 ^{±39}	MBM101, 103, 102, 104, IC 348
G161.9-43.3	161.9	-43.3	329 ⁺⁷ ₋₇	0.361 ^{+0.023} _{-0.023}	0.586 ^{+0.017} _{-0.017}	0.343 ^{+0.016} _{-0.015}	(164.0, -44.9)	(11.2, 6.3)	4.0	6.7	1000	3314	Y	
UT 4	165.3	-26.4	294 ⁺¹³ ₋₉	0.258 ^{+0.066} _{-0.065}	0.836 ^{+0.033} _{-0.032}	0.323 ^{+0.030} _{-0.031}	(164.0, -26.8)	(4.0, 2.0)	6.2	9.0	1000	602	Y	UT 3
MBM17	167.5	-26.6	371 ⁺¹¹ ₋₁₄	0.524 ^{+0.037} _{-0.038}	0.830 ^{+0.019} _{-0.020}	0.316 ^{+0.020} _{-0.017}	(169.3, -27.5)	(5.5, 4.0)	7.1	9.7	1000	1855	165 ⁺¹⁶ ₋₁₄	UT 6, UT 5
3C 75.0	170.3	-44.9	343 ⁺⁸ ₋₁₁	0.406 ^{+0.022} _{-0.022}	0.679 ^{+0.016} _{-0.019}	0.369 ^{+0.016} _{-0.015}	(170.6, -45.8)	(9.0, 9.5)	3.2	7.5	1000	3705	Y	ir1
MBM16	171.7	-37.7	150 ⁺⁶ ₋₆	0.180 ^{+0.029} _{-0.031}	0.357 ^{+0.091} _{-0.098}	0.523 ^{+0.063} _{-0.059}	(167.9, -36.9)	(10.6, 6.6)	5.8	11.2	300	817	147 ⁺¹⁰ ₋₉	
Taurus	175.3	-16.2	145 ⁺¹² ₋₁₆	0.441 ^{+0.157} _{-0.226}	1.219 ^{+0.172} _{-0.220}	0.664 ^{+0.153} _{-0.119}	(170.2, -16.2)	(12.5, 6.7)	9.1	19.1	300	828	138 ^{±15}	
MBM18	189.1	-36.0	161 ⁺² ₋₂	0.204 ^{+0.022} _{-0.024}	0.305 ^{+0.062} _{-0.056}	0.412 ^{+0.041} _{-0.042}	(190.2, -37.1)	(8.5, 11.4)	4.8	9.3	300	1268	166 ⁺¹⁸ ₋₁₇	3C 105.0
Orion Lam	195.5	-13.7	405 ⁺⁴ ₋₄	0.350 ^{+0.016} _{-0.017}	0.739 ^{+0.021} _{-0.020}	0.499 ^{+0.016} _{-0.018}	(196.0, -16.6)	(5.4, 8.7)	5.4	13.2	1000	5616	418 ^{±34}	
Gemini	200.0	12.0	725 ⁺¹⁶ ₋₁₈	0.344 ^{+0.017} _{-0.015}	0.510 ^{+0.018} _{-0.019}	0.386 ^{+0.015} _{-0.015}	(198.4, 12.7)	(5.8, 5.9)	3.2	5.2	1500	6351	Y	
Mon OB1	201.0	1.0	722 ⁺⁴ ₋₄	0.336 ^{+0.009} _{-0.009}	0.857 ^{+0.027} _{-0.028}	0.781 ^{+0.020} _{-0.020}	(201.4, 5.0)	(4.0, 10.6)	4.8	18.0	1500	13448	886 ^{±13}	
NGC 2068	205.3	-14.3	412 ⁺⁴ ₋₄	0.385 ^{+0.042} _{-0.040}	2.148 ^{+0.082} _{-0.077}	0.486 ^{+0.066} _{-0.065}	(204.6, -14.1)	(2.5, 2.0)	11.2	27.7	1000	403		
Rosette	206.8	-1.2	1460 ⁺¹⁹ ₋₁₇	0.280 ^{+0.015} _{-0.015}	2.297 ^{+0.042} _{-0.038}	0.490 ^{+0.035} _{-0.033}	(206.4, -2.3)	(4.7, 3.2)	27.4	56.4	3000	6017	1483 ^{±80}	
L1641	212.5	-19.0	408 ⁺⁴ ₋₄	0.275 ^{+0.029} _{-0.030}	1.966 ^{+0.079} _{-0.083}	0.724 ^{+0.074} _{-0.076}	(213.5, -19.9)	(3.0, 3.2)	7.5	30.0	1000	828		
Mon R2	215.3	-12.9	862 ⁺¹⁰ ₋₁₀	0.327 ^{+0.028} _{-0.027}	2.161 ^{+0.050} _{-0.056}	0.527 ^{+0.046} _{-0.044}	(214.0, -12.3)	(4.6, 4.4)	13.7	26.2	1500	2204	933 ^{±125}	
ir2	235.0	37.0	665 ⁺²⁵ ₋₂₃	0.328 ^{+0.017} _{-0.017}	0.284 ^{+0.042} _{-0.038}	0.405 ^{+0.028} _{-0.028}	(236.4, 38.8)	(6.5, 5.7)	1.7	3.1	1500	2877	Y	235.9+38.2
Lupus	342.5	9.0	168 ⁺⁴ ₋₄	0.273 ^{+0.038} _{-0.036}	1.215 ^{+0.044} _{-0.043}	0.644 ^{+0.037} _{-0.036}	(342.3, 9.8)	(5.0, 6.5)	14.7	25.7	500	1427		
Ophiuchus	355.2	16.0	131 ⁺² ₋₂	0.201 ^{+0.047} _{-0.047}	1.316 ^{+0.087} _{-0.092}	0.698 ^{+0.074} _{-0.074}	(351.3, 16.5)	(7.3, 6.2)	14.6	38.6	300	443	121 ^{±9}	
MBM128	355.6	20.6	128 ⁺² ₋₃	0.283 ^{+0.056} _{-0.055}	0.848 ^{+0.115} _{-0.121}	0.713 ^{+0.092} _{-0.088}	(357.0, 21.6)	(5.4, 5.9)	11.6	21.8	300	395	134 ⁺¹¹ ₋₁₁	MBM130, 133, 127, 126, 125, 129, 131