Core Emergence in a Massive Infrared Dark Cloud: A Comparison between Mid-IR Extinction and 1.3 mm Emission

Shuo Kong, Jonathan C. Tan, Héctor G. Arce, Paola Caselli, Francesco Fontani, and Michael J. Butler

1 Department of Astronomy, University of Virginia, Charlottesville, VA 22904, USA
2 Department of Astronomy, Yale University, New Haven, CT 06511, USA
3 Department of Astronomy, University of Virginia, Charlottesville, VA 22904, USA
4 Max-Planck-Institute for Extraterrestrial Physics (MPE), Giessenbachstr. 1, D-85748 Garching, Germany
5 INAF—Osservatorio Astrofisico di Arcetri, L.go E. Fermi 5 I-50125 Florence, Italy
6 Max Planck Institute for Astronomy, Königstuhl 17, D-69117 Heidelberg, Germany

Received 2017 November 9; revised 2018 February 16; accepted 2018 February 20; published 2018 March 13

Abstract

Stars are born from dense cores in molecular clouds. Observationally, it is crucial to capture the formation of cores in order to understand the necessary conditions and rate of the star formation process. The Atacama Large Millimeter/submillimeter Array (ALMA) is extremely powerful for identifying dense gas structures, including cores, at millimeter wavelengths via their dust continuum emission. Here, we use ALMA to carry out a survey of dense gas and cores in the central region of the massive (∼10^5 M_☉) infrared dark cloud (IRDC) G28.37+0.07. The observation consists of a mosaic of 86 pointings of the 12 m array and produces an unprecedented view of the densest structures of this IRDC. In this first Letter about this data set, we focus on a comparison between the 1.3 mm continuum emission and a mid-infrared (MIR) extinction map of the IRDC. This allows estimation of the “dense gas” detection probability function (DPF), i.e., as a function of the local mass surface density, Σ, for various choices of thresholds of millimeter continuum emission to define “dense gas.” We then estimate the dense gas mass fraction, f_{dp}, in the central region of the IRDC and, via extrapolation with the DPF and the known Σ probability distribution function, to the larger-scale surrounding regions, finding values of about 5% to 15% for the fiducial choice of threshold. We argue that this observed dense gas is a good tracer of the protostellar core population and, in this context, estimate a star formation efficiency per free-fall time in the central IRDC region of ε_{ff} ∼ 10%, with approximately a factor of two systematic uncertainties.

Key words: dust, extinction – ISM: clouds – ISM: structure – radio continuum: ISM – stars: formation – stars: protostars

Supporting material: data behind figure

1. Introduction

Dense cores, as the birthplace of stars, are the focus of intense theoretical and observational study, in particular for understanding the initial conditions and efficiency of star formation (see, e.g., Bergin & Tafalla 2007; Offner et al. 2014; Tan et al. 2014). One theory of core formation is that of gravito-turbulent fragmentation where dense, gravitationally unstable cores are created in density perturbations arising from compressions in supersonically turbulent molecular gas (e.g., Padoan & Nordlund 2002; Krumholz & McKee 2005; Hennebelle & Chabrier 2008; Chabrier et al. 2014). In the theory of Krumholz & McKee (2005; see also Padoan & Nordlund 2011), the rate of star formation is linked to the Mach number and virial parameter (i.e., degree of gravitational boundedness) of the cloud, although it should be remembered that most of the numerical simulation results that underpin these models are of periodic box turbulence (rather than of global cloud structures) and have relatively weak magnetic fields (i.e., are globally magnetically supercritical; see Tan 2016). Models of core formation mediated by magnetic fields, e.g., via ambipolar diffusion (Kunz & Mouschovias 2009; Christie et al. 2017), have also been proposed.

Giant molecular clouds (GMCs) in the Milky Way are known to have a low star formation efficiency (SFE; Zuckerman & Evans 1974; Kennicutt & Evans 2012), where this quantity is most naturally evaluated as the fraction of mass that forms stars in one local free-fall time of the gas, i.e., ε_{ff}, which is seen to have a value of about 0.02. Krumholz & Tan (2007) extended the analysis methods of Zuckerman & Evans (1974) to denser gas structures, such as infrared dark clouds (IRDCs) and embedded clusters, finding similar values of ε_{ff}. Murray (2011) and Lee et al. (2016) have pointed out there is a large dispersion in ε_{ff} in Galactic GMCs, but the average value in the population is still low and consistent with prior estimates. Theoretical models of star formation rates (SFRs) regulated by turbulence (e.g., Krumholz & McKee 2005; Hennebelle & Chabrier 2011; Krumholz et al. 2012) are one way to explain the low averaged observed values of ε_{ff}. However, Lee et al. (2016) have noted that the high dispersion in ε_{ff} is not accounted for in these models. This may imply a role for more stochastic processes, such as triggering by collisions of magnetically supported GMCs inducing bursts of star formation activity (Scoville et al. 1986; Tan 2000; Wu et al. 2015, 2017).

In order to observationally capture the formation of cores and have a sneak peek of the signature of low SFE, we can compare features that are specifically sensitive to the core formation with features representing the host cloud, Dust continuum emission at millimeter wavelengths is sensitive to star-forming cores and often used as a core tracer (e.g., Bergin & Tafalla 2007; Sanhueza et al. 2017). If observed with interferometers like the Atacama Large Millimeter/submillimeter Array (ALMA), millimeter continuum emission can potentially pinpoint dense, star-forming cores embedded in a molecular cloud.
because the extended emission is resolved out by interferometers. Moreover, continuum emission does not depend on chemical abundance variations (such as chemical depletion) which can affect molecular-line-defined cores. The host cloud, and its density structure, can be traced by far-infrared dust emission (André et al. 2014) or dust extinction (Butler & Tan 2009; Lombardi 2009). The latter has been used to construct \( \Sigma \) probability distribution functions (\( \Sigma \)-PDF; e.g., Kainulainen et al. 2009; Butler et al. 2014; Stutz & Kainulainen 2015; Lim et al. 2016). Such distributions appear to have a log-normal component, perhaps arising from turbulence, together with a high-\( \Sigma \) power-law tail, perhaps due to self-gravity. However, observationally it is challenging to accurately measure the \( \Sigma \)-PDF (e.g., Alves et al. 2017; Chen et al. 2017). Numerical simulations of molecular clouds also aim to reproduce the \( \Sigma \)-PDF and understand its dependence on the included physics (see, e.g., Collins et al. 2011; Federrath & Klessen 2013; Myers 2015).

In this Letter, we study the relationship between millimeter continuum emitting dense gas structures and the local mass surface density of their hosting cloud to better understand the conditions of core formation. Our target cloud is IRDC G28.37+0.07 (also referred to as IRDC C) from the sample of Butler & Tan (2009, 2012, hereafter BT09, BT12) at an estimated kinematic distance of 5 kpc. Specifically, we compare the mid-infrared (MIR) extinction map of the IRDC (Butler et al. 2014, hereafter BTK14) with a newly acquired 1.3 mm continuum image, observed with ALMA.

2. Data

2.1. ALMA Observations

The observations were carried out between 2016 June 24 and 2016 June 30 (UTC) (ALMA Cycle 3), under the project 2015.1.00183.S (PI: Kong). Forty-six 12 m antennas were used during the observation in C40-4 configuration. A custom mosaic with 86 pointings was used to cover the majority of the central dark regions of IRDC G28.37+0.07. The observations were in Band 6 (\( \sim 231 \) GHz). A baseband of 1.8 GHz was used for the 1.3 mm continuum observation (the effective bandwidth for continuum imaging is \( \sim 1.4 \) GHz due to the exclusion of the \(^{12}\text{CO}(2-1)\) molecular line). Three other basebands were set for molecular line observations. In this Letter, we focus on the continuum data; we defer the analysis of the molecular line data to a future paper.

The observations consist of six scheduling blocks, each having roughly 50 minutes on-source integration time. J1751+0939 and J1924-2914 were used as bandpass calibrators. J1751+0939, J1924-2914, and Titan were used as flux calibrators. J1851+0035 was used as the phase calibrator. The typical system temperature was 80 K. The mosaic image was cleaned using the standard clean task in CASA. Briggs weighting with a robust number of 0.5 was applied. No self-calibration was done. In order to perform a pixel-by-pixel comparison with the MIREX image (see the following section), we applied an outer uv-taper to match the ALMA synthesized beam to the MIREX beam (2\( ^{\prime} \)). A pixel scale of 0.4\( ^{\prime\prime} \) was adopted in the clean task. Then we re-binned the images to have 1\( ^{\prime\prime} \) pixels to match the MIREX pixel scale. The resulting sensitivity at map center is \( c_{\text{center}} = 0.2 \) mJy per 2\( ^{\prime\prime} \) beam. The maximum recoverable scale of the ALMA continuum image is \( \sim 20^{\prime\prime} \) (corresponding to the shortest baseline of 10 k\( \lambda \) with \( \lambda \) being 1.3 mm).

2.2. The MIREX Image

The MIR extinction (MIREX) map of IRDC G28.37+0.07 was first developed by BT09 and BT12 using Spitzer 8 \( \mu \)m GLIMPSE imaging data (Churchwell et al. 2009). It was merged with a lower-resolution MIR extinction map by Kainulainen & Tan (2013), which improves accuracy at lower values of \( \Sigma \). Finally, the map was refined by Butler et al. (2014) by using an analysis of deeper archival Spitzer-IRAC imaging, which enables the highest dynamic range of \( \Sigma \) to be probed. In general, the method of MIREX mapping involves estimating the intensity of the diffuse background emission, i.e., from the diffuse Galactic interstellar medium (ISM), via interpolation from surrounding regions, and estimating, empirically, the level of the foreground emission. Then, given an estimate of the dust opacity at 8 \( \mu \)m (averaged over the Spitzer-IRAC Band 4) and a dust-to-gas mass ratio, the total mass surface density can be calculated by solving the simple 1D radiative transfer equation, given the observed intensities emerging from the cloud. The spatial resolution achieved in the map is 2\( ^{\prime\prime} \) with a pixel scale of 1\( ^{\prime\prime} \), set by the resolution of the Spitzer-IRAC data.

There are several effects that lead to systematic errors in the MIREX maps. One problem is that in regions containing local bright MIR source, the extinction is contaminated by the source. Another problem is that in some regions the IRDCs become very optically thick, so only a lower limit on \( \Sigma \) can be estimated. These regions are referred to as being “saturated” in the MIREX map (their presence allows the measurement of the diffuse foreground emission, assumed to be spatially constant). Local fluctuations in the background will lead to errors, since it is modeled as a smoothly varying source. Zero point offsets of up to \( \sim 0.1 \) g cm\(^{-2} \) are present, which are partially corrected for by calibration with NIR extinction maps (Kainulainen & Tan 2013). Still, the zero-point uncertainty is present at a level estimated to be \( \sim 0.02 \) g cm\(^{-2} \) (i.e., \( A_V \sim 4 \) mag or so).

3. Results

3.1. Comparison between 1.3 mm Dust Continuum Emission and MIR Extinction

Figure 1 shows the comparison between the cloud mass surface density from the MIREX map and the 1.3 mm dust continuum emission observed by ALMA. In general, the MIREX image shows mainly \( \Sigma \geq 0.2 \) g cm\(^{-2} \) pixels in the ALMA-mosaicked region. They correspond to relatively dark regions in the original Spitzer-IRAC 8 \( \mu \)m image. The MIREX map reveals features via dust absorption (depending on total \( \Sigma \)), while the ALMA image shows dust emission (depending on total \( \Sigma \) and dust temperature). Another difference arises due to ALMA filtering out low spatial frequency (larger-scale) structures. In our case, the recoverable physical scales range from 10,000 au (0.05 pc, 2\( ^{\prime} \) after uv-tapering) to approximately 100,000 au (0.48 pc, 20\( ^{\prime} \)). We note that the Jeans length

\[
\lambda_J = 0.10 \left( \frac{T}{15 \text{ K}} \right)^{1/2} \left( \frac{n_1}{10^5 \text{ cm}^{-3}} \right)^{-1/2} \text{ pc}
\]

is in the range of recovered scales, given typical conditions of ambient gas in the IRDC. Consequently, while the extinction map tracks the total column density, the ALMA continuum
image pinpoints compact, dense, and warmer structures, i.e., expected to be protostellar cores. Thus, through comparison with the extinction map, the ALMA image shows us where such dense, likely star-forming, structures emerge from the cloud.

We now give a brief overview of several of the regions seen in the map. Dense “cores/clumps” C1 to C16 were identified in the MIREX map by BT12 and BTK14. The continuum cores in the southwest C1 region were studied by Tan et al. (2013, 2016) and Kong et al. (2017). C1a-Sa and C1-Sb have been identified as protostellar cores and C1a and C1b as candidate protostellar cores. A massive pre-stellar core candidate, C1-S, identified by N2D+(3 – 2) emission by Tan et al. (2013), sits between C1-Sa and C1-Sb, but has relatively faint 1.3 mm continuum emission. C1 is the location of the C1-N core, which is another massive pre-stellar core candidate identified by its N2D+(3 – 2) emission. We note that most of the protostellar cores (including the relatively low-mass ~2 M⊙ C1-Sb core) and some massive pre-stellar cores are well detected in the ALMA continuum image.

Moving to the northeast, several other sources are seen in the region, including the C14, C15, and C16 core/clumps. Next, we come to the C2 region, which corresponds to the “P1 clump” studied by Zhang et al. (2009, 2015). They identified a linear chain of five main continuum structures, with a hint of a sixth core/clump at the southwest end. Here, we confirm the detection of this sixth, weaker continuum structure. Like the other cores, it also corresponds to a high-Σ peak in the MIREX map. With the higher-resolution (~0″7) observations

Figure 1. Grayscale: MIREX mass surface density map from BTK14 (scale in g cm⁻²). The angular resolution of the map is shown as the gray filled circle at the lower left. “C1, C2, C3...” label extinction peaks from BTK14. The white arrows point to possible embedded protostars that show as local enhancements in the 8 μm image, which produce local “holes” in the MIREX map. Contours: ALMA 1.3 mm continuum mosaic. The contours range from S/N = 2, 3, 5, 10, 20, 40, ... with the rms noise at map center σ_{cont} ~ 0.2 mJy beam⁻¹. The two red contours highlight S/N = 3 and 10. The synthesized beam is shown as the red filled ellipse at the lower right. The two white enclosing contours show primary-beam responses of 0.3 (outer) and 0.5 (inner). (The data used to create this figure are available.)
of Zhang et al. (2015) a few tens of cores were identified in the C2 region down to subsolar masses, with many of these seen to be protostellar by the presence of bipolar CO outflows.

Northwest of C2 is a region containing C4, C10, and C13, with most of the mass concentrated near C4 and C13. Several distinct millimeter continuum peaks are visible in this region. Continuing northeast from C2 is the sequence of MIR dark core/clumps C5 and C6, which contain a cluster of millimeter emission cores, then the sparser C11 and C12. Between C11 and C12 there is a narrow filament seen in millimeter continuum emission, which closely follows the morphology seen in the MIREX map. This filament shows signs of fragmenting into several cores (including C12), but may be at an earlier stage of evolution compared to the more fragmented regions described above, such as C5/C6, C4/C13, and perhaps C2.

Overall, Figure 1 shows that the 1.3 mm continuum structures follow the extinction features quite well, i.e., they tend to be found in high-$\Sigma$ regions of the MIREX map. For example, in the region around C4, the cloud shows very good agreement between the continuum emission and high-$\Sigma$ pixels. On the other hand, MIREX high-$\Sigma$ regions do not always show millimeter continuum emission. This is illustrated in the region around C11, where it shows few robust 1.3 mm continuum detections. Being in a high-$\Sigma$ region is a necessary, but not sufficient, condition for the presence of strong millimeter continuum cores.

In order to reveal more quantitatively the large-scale mass surface density conditions needed for the formation of 1.3 mm continuum emitting structures, we make a pixel-by-pixel comparison between the ALMA image and the MIREX image (Figure 2). We show two different types of comparisons. In panel (a), we compare signal-to-noise ratio ($S/N$) and $\Sigma$. In panel (b), we compare the 1.3 mm continuum flux density $F_{1.3\,\text{mm}}$ with $\Sigma$. The continuum image is primary-beam corrected, so the map boundary regions have higher noise levels. Both comparisons are restricted to regions where the ALMA primary-beam response $> 0.5$. In both panels, we show the $3\sigma_{\text{center}}$ noise level with a blue dashed horizontal line. The zero point is shown as the red dashed horizontal line. A $3\sigma_{\text{center}}$ noise corresponds to a continuum-derived mass surface density $\Sigma_{\text{mm}} = 0.044 \, \text{g cm}^{-2}$ (using Equation (1) in Kong et al. 2017), assuming a dust temperature of 20 K, $\kappa_{\nu} = 5.95 \times 10^{-3} \, \text{cm}^{2} \, \text{g}^{-1}$ (the moderately coagulated thin ice mantle model of Ossenkopf & Henning 1994), i.e., with a dust-to-gas mass ratio of 1:141 (Draine 2011). For a mean particle mass of 2.33$\,m_{\text{H}}$ (i.e., $n_{\text{H}} = 0.1\,n_{\text{H}}$, this corresponds to a total column density $N_{\text{H}} = 1.9 \times 10^{22} \, \text{cm}^{-2}$, i.e., a visual extinction of $A_{V} = 9.4$ mag (assuming an extinction to column density relation $A_{V} = (N_{\text{H}}/2.0 \times 10^{21} \, \text{cm}^{-2})$ mag). We note that our restriction of analysis to the region where the primary-beam correction factor is $< 2$ means that uncertainties associated with this correction are minimized to this level or smaller.

At first glance, the plots show no clear correlation between the millimeter continuum flux and MIREX $\Sigma$. A similar situation was found by Johnstone et al. (2004) comparing 0.85 mm continuum emission (observed with JCMT) and near-infrared extinction (derived from 2MASS data). However, Figure 2 shows a hint of detection deficit of millimeter continuum emission at $\Sigma \lesssim 0.15 \, \text{g cm}^{-2}$, although there are still a modest number of relatively high $S/N$ and flux density values in this regime. However, one important systematic error associated with the MIREX map is the presence of MIR-bright sources, which lead to an underestimation of $\Sigma$ at these locations. We carry out a visual identification of potential MIR sources in the Spitzer-IRAC 8 $\mu$m image and mark their locations in Figure 1. We then remove these pixels from the analysis, showing the results in Figure 2(c), (d). There are now significantly fewer low $\Sigma$ (i.e., $\lesssim 0.3 \, \text{g cm}^{-2}$) points with high $S/N$ or flux density values.

Focusing on the results in Figure 2(c), (d), we first note that there are very few pixels with $\Sigma \lesssim 0.1 \, \text{g cm}^{-2}$, since even the boundary of the mapped region still corresponds to quite deeply embedded parts of the molecular cloud. Also, there are relatively few points with $\Sigma \gtrsim 0.6 \, \text{g cm}^{-2}$, which is partly due to the effects of approaching the saturation limit in the MIREX map (BTK14). Then, we see that the cloud of points within $-3\sigma_{\text{center}} \lesssim F_{1.3\,\text{mm}} \lesssim 3\sigma_{\text{center}}$ shows the rms noise in the continuum image. At $\Sigma \lesssim 0.5 \, \text{g cm}^{-2}$, most of the pixels still aggregate within $\pm 3\sigma_{\text{center}}$ rms noise. However, starting from $\Sigma \sim 0.2 \, \text{g cm}^{-2}$, we see increased numbers of high $S/N$ and flux density values. By $\Sigma \gtrsim 0.65 \, \text{g cm}^{-2}$, nearly all points are above the $3\sigma_{\text{center}}$ line. In other words, with the increase of $\Sigma$, it is more likely to detect 1.3 mm continuum flux with ALMA (given the recoverable angular scales). When the IRDC has a high enough mass surface density ($\Sigma \gtrsim 0.65 \, \text{g cm}^{-2}$), the 1.3 mm continuum emitting dense structures are always present. If the continuum detections indicate current/future star-forming cores, this would indicate that core/star formation is more likely to happen in high-$\Sigma$ regions of IRDCs.

### 3.2. Dense Gas Detection Probability Function (DPF)

To further quantify the relation between presence of 1.3 mm continuum emission and mass surface density of the parent cloud, we plot the detection probability, $P_{1.3\,\text{mm}}$, versus $\Sigma$ in Figures 3 and 4, using the data set with pixels containing MIR sources removed (see above). Here, $P_{1.3\,\text{mm}}$ is defined as the fraction of “detected” pixels at a given $\Sigma$. The definition of detection differs by cases. In the first case (Figure 3), a pixel is defined to be detected when its $S/N$ is greater than a given threshold. A low threshold is more likely to have false detections, and vice versa. We adopt a fiducial threshold of $S/N = 3$, and show the effects from using $S/N = 2$ and $S/N = 4$. In the second case (Figure 4), a pixel is defined to be detected when its flux density is greater than a given threshold. Here, we use the primary-beam corrected image. The fiducial threshold is $3\sigma_{\text{center}}$ at the map center, where the primary-beam response is 1. We also show the effects of using $2\sigma_{\text{center}}$ and $4\sigma_{\text{center}}$.

In the first case of a constant $S/N$ threshold, it is possible that we miss some weak features at the map boundary where the rms noise $\sigma$ is a factor of 2 higher than $\sigma_{\text{center}}$. In the second case of a constant absolute flux density threshold, while this is closer to a constant physical limit, i.e., of constant core column density for fixed dust temperature and dust opacity, the disadvantage is that we may be overestimating $P_{1.3\,\text{mm}}$ near the map boundary due to increased contamination from noise fluctuations.

In these analyses, we adopt a bin size of $\Delta\Sigma = 0.02 \, \text{g cm}^{-2}$ ($A_{V} \sim 4$ mag). In the left columns of Figures 3 and 4, we show the $P_{1.3\,\text{mm}},\Sigma$ relation with a linear scale. In the right columns, we show the relation with a logarithmic scale. Each row of
panels shows the relation with a different detection threshold, as noted on the top left corner.

In each \( \Sigma \) bin, \( P_{1.3 \text{ mm}} \) is detection total of \( 1.3 \text{ mm} \) point obeys the Bernoulli distribution with success probability \( p \), i.e.,

\[
P(X) = \begin{cases} 
p & \text{if } X = 1 \\
1 - p & \text{if } X = 0,\end{cases}
\]

where \( X = 1 \) means detection, then \( P_{1.3 \text{ mm}} \) is the expectation of \( \sum_{i=1}^{n} X_i/n \), given \( X_1,..X_n \) are independent, identically distributed random variables. The standard deviation of \( \sum_{i=1}^{n} X_i/n \) is \( \sqrt{p(1-p)/n} \), which is adopted as the error bar for each bin. We use the observed probability as an estimate of the Bernoulli success probability. Note that by this method, estimating the error bar becomes problematic when the success probability equals 0 or 1. Such points are excluded from the functional fitting (see below).

At \( \Sigma \lesssim 0.04 \text{ g cm}^{-2} \) (\( A_V \sim 8 \text{ mag} \)), there are very few (i.e., only about 5) pixels in the mapped region. While these pixels do not tend to show millimeter continuum flux detections via the various thresholds, there are too few for us to test scenarios of there being a threshold for star formation at about this level (e.g., McKee 1989; Johnstone et al. 2004; Lada et al. 2010). Also, we note that the MIREX map, even with NIR extinction correction, can have relatively large systematic errors in this low-Sigma regime. Indeed, such problems, including incomplete removal of MIR sources, lead us to be cautious of results for \( \Sigma \lesssim 0.15 \text{ g cm}^{-2} \), where \( P_{1.3 \text{ mm}} \) is seen to sometimes have finite values, but typically with large errors.

However, in the main region of interest for our study, i.e., for \( \Sigma \gtrsim 0.15 \text{ g cm}^{-2} \), in all the cases the detection probability increases steadily to reach approximately 100% by \( \Sigma \sim 0.65 \text{ g cm}^{-2} \). In Figures 3(b), (d), (f) and 4(b), (d), (f), the plots show that \( P_{1.3 \text{ mm}} \) follows an approximate power-law relation with \( \Sigma \) between \( \Sigma \sim 0.15 \text{ g cm}^{-2} \) and \( \Sigma \sim 0.65 \text{ g cm}^{-2} \). We fit...
Figure 3. (a) 1.3 mm continuum detection probability $P_{1.3\,\text{mm}}$ as a function of $\Sigma$. Here, the continuum detection threshold is $S/N = 2$ with the rms noise being 0.2 mJy per 2" beam. The error bars are the square root of the variance of the Bernoulli distribution (see the text). (b) The same as (a), but in logarithmic scale. The blue dashed line shows a power-law fit over the indicated range of $\Sigma$. The parameters $a$ and $b$ follow Equation (3). (c) Same as (a), but with a detection threshold of $S/N = 3$. (d) Same as (c), but in logarithmic scale. (e) Same as (a), but with a detection threshold of $S/N = 4$. (f) Same as (e), but in logarithmic scale.
Figure 4. (a) 1.3 mm continuum detection probability $P_{1.3 \text{ mm}}$ as a function of $\Sigma$. Here, the continuum detection threshold is $2\sigma_{\text{center}}$, where $\sigma_{\text{center}}$ (0.2 mJy per 2″ beam) is the rms noise at map center (primary-beam response = 1). The error bars are the square root of the variance of the Bernoulli distribution (see the text). (b) The same as (a), but in logarithmic scale. The blue dashed line shows a power-law fit over the indicated range of $\Sigma$. The parameters $a$ and $b$ follow Equation (3). (c) Same as (a), but with a detection threshold of $3\sigma_{\text{center}}$. (d) Same as (c), but in logarithmic scale. (e) Same as (a), but with a detection threshold of $4\sigma_{\text{center}}$. (f) Same as (e), but in logarithmic scale.
Table 1

| Thresholds | \( \Sigma_{\text{mm}} \) (g cm\(^{-2}\)) | \( a \) | \( b \) | \( f_{\text{dg,mm}} \) | \( f_{\text{dpf,mm}} \) | \( f_{\text{dpf,mirex}} \) | \( f_{\text{dpf,dpf,gmc}} \) |
|------------|-----------------|------|------|-----------------|-----------------|----------------|----------------|
| \( S/N \geq 2 \) | 0.029 | 3.3 | 2.6 | 10\%\(^{15}\) | 20\% | 17\% | 9.2\% |
| \( S/N \geq 3 \) | 0.044 | 4.6 | 3.3 | 8.7\%\(^{15}\) | 13\% | 12\% | 6.5\% |
| \( S/N \geq 4 \) | 0.058 | 4.5 | 3.6 | 8.0\%\(^{15}\) | 10\% | 9.4\% | 5.7\% |
| \( f_{1.3 \text{ mm}} \geq 2 \sigma_{\text{center}} \) | 0.029 | 2.4 | 2.0 | 11\%\(^{15}\) | 24\% | 22\% | 13\% |
| \( f_{1.3 \text{ mm}} \geq 3 \sigma_{\text{center}} \) | 0.044 | 3.1 | 2.6 | 9.5\%\(^{15}\) | 17\% | 15\% | 8.2\% |
| \( f_{1.3 \text{ mm}} \geq 4 \sigma_{\text{center}} \) | 0.058 | 3.7 | 3.2 | 8.6\%\(^{15}\) | 12\% | 11\% | 6.3\% |

Note. \(^{a}\) The superscripts and subscripts correspond to using the lower (15 K) and higher (30 K) temperature assumptions in the mass estimation based on 1.3 mm continuum flux.

The Astrophysical Journal Letters, 855:L25 (11pp), 2018 March 10

Kong et al.

The function \( P_{1.3 \text{ mm}} = a(\Sigma/1 \text{ g cm}^{-2})^b \) by minimizing \( \chi^2 \) (normalized by the errors), which is shown as the blue dashed lines in these figures. Note that we do not include \( P_{1.3 \text{ mm}} = 1 \) points in the fit. The resulting power-law indices \( b \) and amplitudes \( a \) are displayed in the figures and in Table 1.

With an increase in the level of the detection thresholds, Figures 3 and 4 show a decrease in detection probabilities, as expected. At the same time, the power-law indices become larger, i.e., with a higher detection threshold, the increase of \( P_{1.3 \text{ mm}} \) between \( \Sigma \approx 0.15 \text{ g cm}^{-2} \) and 0.65 g cm\(^{-2}\) becomes steeper. In the next section, we will use such power-law approximations for \( P_{1.3 \text{ mm}} (\Sigma) \) to estimate the mass fraction of “dense” gas in the IRDC and GMC region.

3.3. Dense Gas Fraction

The ALMA observations give us a direct measure of the amount of “dense” gas, i.e., that is detected by some defined criteria of 1.3 mm flux emission, which can be compared to the total mass estimate of the IRDC that overlaps with the region mapped by ALMA. From the MIREX map, this mass is 1.21 \( \times 10^{4} M_{\odot} \), with uncertainties at the level of about 30\% due to opacity per unit mass uncertainties. Distance uncertainties contribute further, but these will cancel out in the ratio of these masses to the millimeter-continuum-derived mass.

The total millimeter flux in the observed, analyzed region (i.e., where the primary-beam correction factor is \( \leq 2 \)) is 1.42 Jy (based on detections above \( 3\sigma_{\text{center}} \)) which translates into a total mass of 1.16 \( \times 10^{4} M_{\odot} \) given our fiducial assumptions, including \( T = 20 \text{ K} \). Thus, the direct measure of dense gas mass fraction (expressed as percentages) is \( f_{\text{dg,mm}} = 9.5\% \) for this case. This value is listed in column (5) of Table 1 for all the considered cases, showing the effects of varying \( T \) from 15 to 30 K. We see the sensitivity of these dense gas fractions to threshold choice and temperature choice, with fiducial results being about 10\%. Systematic variations arising from the choice of dust temperature are up to a factor of almost two and are the most significant source of uncertainty (see also Goodman et al. 2009).

A second estimate of the dense gas fraction, \( f_{\text{dpf,mirex}} \), can be made by summing the MIREX mass estimate of the pixels that are detected in 1.3 mm continuum. These values are shown in column (6) of Table 1. Fiducial results are now moderately higher at about 15\%.

Next, we utilize our analytic approximations for the DPF, \( P_{1.3 \text{ mm}} (\Sigma) \), combined with analytic forms for the probability distribution function (PDF) of \( \Sigma \) to estimate dense gas fractions. Recall that the observed DPFs have a power-law form in the range from \( \Sigma \approx 0.15 \text{ g cm}^{-2} \) to \( \approx 0.65 \text{ g cm}^{-2} \). At lower values of \( \Sigma \) we extrapolate with a constant that is similar to the \( P_{1.3 \text{ mm}} \) at \( \Sigma = 0.15 \text{ g cm}^{-2} \). Finally, at high values, \( \Sigma > 0.65 \text{ g cm}^{-2} \), we use a constant value of unity. Thus, overall the DPF is described via

\[
P_{1.3 \text{ mm}} = \begin{cases} 
P_{1.3 \text{ mm,min}} \text{ if } \Sigma/(\text{g cm}^{-2}) < 0.15 \\
\quad\quad\quad a \left( \frac{\Sigma}{1 \text{ g cm}^{-2}} \right)^b \text{ if } 0.15 < \Sigma/(\text{g cm}^{-2}) < 0.65 \\
\quad\quad\quad 1 \quad \text{ if } \Sigma/(\text{g cm}^{-2}) > 0.65,
\end{cases}
\]

where \( P_{1.3 \text{ mm,min}} = a(0.15 \text{ g cm}^{-2}/1 \text{ g cm}^{-2})^b \). The fiducial value for \( P_{1.3 \text{ mm,min}} \) is \( \approx 0.02 \). This value acts effectively as a lower limit floor on our estimated values of \( f_{\text{dg}} \).

Then the mass of dense, i.e., 1.3 mm emitting, gas is

\[
M_{\text{dg}} = \int P_{1.3 \text{ mm}} \Sigma A \rho(\ln \Sigma) d(\ln \Sigma),
\]

where \( A \) is the total cloud area being integrated over and \( \rho(\ln \Sigma) \) is the cloud’s PDF of mass surface densities. Based on two independent methods, the \( \Sigma \)-PDF in IRDC G28.37+0.07 and its surroundings (i.e., of a \( \approx 20 \) pc region, equivalent to \( \approx 30 \) pc) has been found to be reasonably well fit by a single log-normal function (Butler et al. 2014; Lim et al. 2016), i.e., of the form

\[
\rho(\ln \Sigma) = \frac{1}{\sqrt{2 \pi} \sigma_{\ln \Sigma}} \exp \left( - \frac{(\ln \Sigma - \bar{\ln \Sigma})^2}{2 \sigma_{\ln \Sigma}^2} \right).
\]

Here, we adopt this empirical \( \Sigma \)-PDF (i.e., area-weighted)\(^7\) in the NIR+MIR extinction map case, i.e., with \( \sigma_{\ln \Sigma} = 1.15, \Sigma = 0.038 \text{ g cm}^{-2}, \) and \( \bar{\ln \Sigma} = -3.93. \) We note that the actual \( \Sigma \)-PDF measured by Lim et al. (2016) has a small power-law tail excess component, emerging at about \( \Sigma \approx 0.3 \text{ g cm}^{-2} \). While the use of the above log-normal leads to a small underestimation of the importance of the higher \( \Sigma \) regions, it is

\(^7\) We have also made the same calculations using their mass-weighted PDF. The results (dense gas fractions) are very similar.
a very modest effect since the fraction of pixels affected by this excess is less than a few percent.

Then the total mass of dense gas can be estimated by integrating Equation (4). If we carry out this exercise for the area corresponding to the analyzed area of the IRDC, i.e., that mapped by ALMA with a primary-beam response >0.5, we obtain 135 $M_{\odot}$. This is much smaller than our previous estimates for $M_{\text{gal}}$, which is primarily because the $\Sigma$-PDF was estimated for a much larger region and contains much more contribution from lower values of $\Sigma$. If we restrict the above integration to the range $\Sigma = 0.15$–0.65 g cm$^{-2}$, then we obtain $M_{\text{gal}} = 1828 M_{\odot}$ (for the $>3\sigma_{\text{center}}$ case), in much closer agreement with our previous estimates. Dense gas fractions calculated via this latter method can be derived by comparison to the total cloud mass observed in the mapped region, i.e., $1.21 \times 10^5 M_{\odot}$, yielding the values $f_{\text{dg,DPF,0.15–0.65}}$ in column (7) of Table 1. These values are very similar to those of $f_{\text{dg,MIREX}}$.

Finally, we can make the extrapolation that the observed DPF of the inner IRDC region mapped by ALMA will hold in the wider GMC region, where the approximately log-normal $\Sigma$-PDF was measured. For this $\sim 30$ pc scale region, the total cloud mass is

$$M_{\text{tot}} = \int \Sigma A_p (\ln \Sigma) d (\ln \Sigma),$$

which has a value of 170,000 $M_{\odot}$. The values of $f_{\text{dg,DPF,GMC}} = M_{\text{dg}} / M_{\text{tot}}$ are shown in column (8) of Table 1. In the fiducial cases, these values are smaller than 10%.

4. Discussion

4.1. Core/Star Formation Efficiency

The MIR extinction map and the ALMA 1.3 mm continuum map both trace dust in the IRDC, which are then used to estimate the masses. However, while the MIREX map traces the total mass surface density without bias at any particular spatial scale and without bias on the temperature (as long as the region is cold enough not to be emitting at $8 \mu m$), the ALMA continuum map misses flux from extended structures ($\gtrsim 20'$) and is biased toward warmer material. We describe the mass associated with the 1.3 mm continuum flux as the “dense” gas component and discuss below that the majority of this material is likely to be directly involved in the star formation process.

We have measured the mass of the component that is detected by our ALMA observation of dust continuum emission and find it to be about $f_{\text{dg,mm}} \sim 10$% of the total mass in the “central”, i.e., mapped region of the IRDC, but with about 50% uncertainties due to assumed dust temperature. If we use the values of the MIREX pixels at the locations where millimeter continuum emission is seen, then the associated mass fraction increases by a factor of about 1.7 (depending on the choice of flux threshold), i.e., to $f_{\text{dg,MIREX}} \sim 17$$\%$. The difference between $f_{\text{dg,MIREX}}$ and $f_{\text{dg,mm}}$ could be due to, e.g., a dense core filling factor of less than one on the scale of the 2$''$ pixels or a systematically lower temperature of the millimeter continuum emitting dust, i.e., $\sim 15$ K rather than 20 K.

If we use the data to define a detection probability of millimeter continuum emission as a function of $\Sigma$ and then apply this to an estimate of the $\Sigma$-PDF of the mapped region of the IRDC, i.e., a log-normal but restricted to the range $\Sigma = 0.15$–0.65 g cm$^{-2}$, then we obtain values of dense gas fractions of $f_{\text{dg,DPF,0.15–0.65}} \approx 15\%$, very similar to the values of $f_{\text{dg,MIREX}}$ (also compare other values in columns (6) and (7)), which indicates that the analytic approximations for the DPF are quite accurate. Extrapolating the observed DPF of the inner IRDC region to the wider GMC region, where the $\Sigma$-PDF was seen to be well fit by a single log-normal (BTK14; Lim et al. 2016), then integration with this PDF leads to estimates of $f_{\text{dg,DPF,GMC}} \approx 8\%$. We note that this mass fraction is very similar to the mass fraction of the GMC that is in the power-law tail part of the $\Sigma$-PDF, $\epsilon_{\text{ol}} \sim 3\%$ to 8\% (Lim et al. 2016) based on lower angular resolution Herschel measurements of submillimeter dust continuum emission from the region.

Our ALMA continuum map detects the C1a, C1-Sa, and C1-Sb protostellar cores from Tan et al. (2016), which includes some lower-mass objects. It also detects the five main continuum structures in C2 (Zhang et al. 2009, 2015), which have been resolved into a population of cores extending down to subsolar masses. Thus, it is likely that the current observations capture a significant fraction of the core mass function (CMF) of protostellar cores. The detected millimeter flux may also contain some contribution from more massive pre-stellar cores, such as C1-S and C1-N (Tan et al. 2013; Kong et al. 2017). Thus, for simplicity, we will assume that our detected 1.3 mm continuum fluxes give a near complete census of the protostellar CMF and ignore the possibility that it may include some contribution from the pre-stellar CMF. These effects of protostellar CMF incompleteness and pre-stellar CMF contribution will offset each other to some extent. Under this assumption, then the total current star-forming core efficiency is simply the same as $f_{\text{dg}}$. If we next further assume that the star formation efficiency from individual cores is about 50%, which is expected based on models of outflow feedback (Matzner & McKee 2000; Zhang et al. 2014), then the total mass of stars that would form from the currently observed cores is about half of $f_{\text{dg}}$, i.e., $\epsilon_{\text{s}} \approx 5\%$ to $\sim 8\%$.

4.2. Star Formation Rates

A number of star formation models involve protostellar cores collapsing at rates similar to their local free-fall rate (e.g., Shu et al. 1987; McKee & Tan 2003; Krumholz & McKee 2005). The Turbulent Core Model (McKee & Tan 2003, hereafter MT03) assumes core properties are set by the mean pressure in their surrounding, self-gravitating clump, which then leads to a simple relation between the individual star formation time and the average free-fall time of the clump. In the fiducial case, the timescale for star formation is $t_{\text{sf}} = 1.3 \times 10^7 (M_e / 60 M_{\odot})^{1/4} (\Sigma_e / 1 g cm^{-2})^{-3/4}$ year (cf. Equation (44) in MT03), which has a very weak dependence on core mass, $M_e$, and clump mass surface density, $\Sigma_e$. This timescale is related to the clump’s mean free-fall time via $t_{\text{sf}} / t_{\text{ff,cl}} = 0.98 (M_e / 60 M_{\odot})^{1/4} (M_e / 4000 M_{\odot})^{-1/4} (\text{cf.} \quad \text{Equation (37) in MT03})$, i.e., they are quite similar.

For a CMF that is a Salpeter (1955) power law of form $dN / d \log M \propto M^{-\alpha}$ with $\alpha = 1.35$ with lower limit of $M_e = 1 M_{\odot}$ and upper limit of 240$M_{\odot}$ (so that resulting stellar initial mass function (IMF) with 50% formation efficiency from the core is in the range from $m_* = 0.5 M_e$ to 120 $M_e$, which is, for our purposes, a reasonable approximation of the actual observed IMF), then half of the mass of the core population has $M_e \gtrsim 5 M_{\odot}$. Thus we take $M_e = 5 M_{\odot}$ as a typical core mass. The mapped region of the IRDC has a total mass of $\sim 1.21 \times 10^4 M_{\odot}$, which we will approximate as $10^4 M_{\odot}$. Under these two conditions, $t_{\text{sf}} / t_{\text{ff,cl}} \rightarrow 0.42$. 

Kong et al.
Assuming the SFR is steady and the CMF is evenly populated, then the observed cores will represent those objects that have formed in the last average individual star formation time, $t_{ff}$, i.e., the last $0.42t_{ff,cl}$. Taking the mass fraction in dense gas (defined at $F_{1.3\,\text{mm}} \geq 3\sigma_{\text{center}}$) as $f_{1.3\,\text{mm}} = 0.093^{+0.14}_{-0.058}$ as the most accurate estimate of the current mass fraction in protostellar cores in the observed region of the IRDC, then we find that, for $\epsilon_{\text{core}} = 0.5$ (Matzner & McKee 2000; Zhang et al. 2014), the star formation efficiency per free-fall time is $\epsilon_{ff} = 0.1^{+0.17}_{-0.069}$.

This estimate of $\epsilon_{ff} \sim 0.1$ is about a factor of two larger than the value estimate inside the half-mass radius of the Orion Nebula Cluster by Da Rio et al. (2014), which was estimated from observed age spreads of young stellar objects. However, the uncertainties arising solely from the uncertain temperatures of protostellar cores (15–30 K range adopted here) lead to almost a factor of two uncertainty in $\epsilon_{ff}$. The mean mass surface density in the analyzed region of the IRDC is $\approx 0.26$ g cm$^{-2}$. The protostellar core models of Zhang & Tan (2015), i.e., for $M_c = 2, 3, 4M_\odot$, in $\Sigma \approx 0.3$ g cm$^{-2}$ clump environments have mean envelope temperatures near 20 K (set mostly by accretion luminosities), but can exceed 30 K in $\Sigma \approx 1$ g cm$^{-2}$ regions that have higher accretion rates. Also, more massive cores forming more massive protostars will tend to have warmer envelope temperatures, which would lower our estimates of the mass of the core population and thus $\epsilon_{ff}$. These uncertainties can be reduced by carrying out temperature measurements of each protostellar core (e.g., of the dust via spectral energy distribution observations and modeling or of associated gas via, e.g., NH$_3$ observations).

In addition to the effects of core temperature uncertainties, additional systematic uncertainties include that the analysis has assumed a fixed value of the star formation efficiency from the core, a particular relation between star formation time and clump free-fall time (fiducial case from MT03) and equates the observed 1.3 mm continuum structures with the total protostellar core population. These assumptions and uncertainties can be improved with future work. For example, observations of CO outflows can be used to confirm that millimeter continuum sources are indeed protostellar cores. Better sensitivity of millimeter continuum data can help to probe further down the protostellar CMF (although with the half-mass point estimated to be near $5 M_\odot$, we expect that the bulk of the population containing most of the mass has already been detected). Assumptions about star formation efficiency from the core can be tested with improved theoretical and numerical models (e.g., Matsushita et al. 2017; Tanaka et al. 2017). The relation of individual star formation time to mean clump free-fall time is more difficult to test observationally, and may depend on the uncertain degree of magnetization in the cores (Li & Shu 1997, MT03). One observational test involves measuring the mass accretion rates of the protostars, potentially from modeling their spectral energy distributions (see, e.g., Zhang & Tan 2015, 2017; De Buizer et al. 2017) or from measuring their mass outflow rates that are expected to be proportional to accretion rates (see, e.g., Beltrán & de Wit 2016).

5. Conclusions

In this Letter, we have presented first results from an ALMA 1.3 mm continuum mosaic observation using the 12 m array of the central regions of a massive IRDC, which is a potential site of massive star cluster formation. We have focused on carrying out a detailed comparison of the 1.3 mm emission (which is sensitive to structures $\lesssim 20''$ in size) with an MIR-derived extinction map of the cloud. In particular, we argue that the 1.3 mm structures likely trace “dense,” protostellar cores, and have studied the prevalence of such sources in the IRDC as a function of its local mass surface density, $\Sigma$. Based on various definitions of 1.3 mm continuum detection, i.e., at a fixed signal-to-noise ratio or a fixed absolute flux density, we find that the DPF, $P_{1.3\,\text{mm}}(\Sigma)$, rises as a power law, i.e., $\propto (\Sigma/1\,\text{g cm}^{-2})^{0.3}$ with $b \approx 3$ in the fiducial cases, over the range $0.15 \lesssim \Sigma < 1\,\text{g cm}^{-2}$ $\sim 0.65$. At higher values of $\Sigma$, we find that $P_{1.3\,\text{mm}} \sim 1$. At lower values of $\Sigma$, which are not so common in the mapped region, we have weaker constraints on $P_{1.3\,\text{mm}}$ but approximate it as a constant of $\sim 10^{-2}$ in the fiducial cases. Such an empirical relation can provide a test of theoretical/numerical models of star formation.

We have then utilized the continuum image and the estimated form of $P_{1.3\,\text{mm}}(\Sigma)$ to carry out various estimates of the “dense” gas mass fraction, $f_{\text{dg}}$, in the IRDC and, by extrapolation with the observed $\Sigma$-PDF, in the larger-scale GMC region. The mass estimate in the mapped region of the IRDC made directly from the observed 1.3 mm flux depends on adopted dust opacities and temperatures, but has a fiducial value of just under 10%. Using the MIREX $\Sigma$ at a location of 1.3 mm flux detection leads to mass fraction estimates that are about a factor of 1.5 times higher. Extrapolating to the larger-scale region, given its observed log-normal $\Sigma$-PDF, we find values of $f_{\text{dg}} \sim 7\%$.

Finally, assuming that the detected 1.3 mm structures mostly trace protostellar cores and capture the bulk of the mass of the core population, we use these results to estimate the SFR in the IRDC, in particular the star formation efficiency per free-fall time, $\epsilon_{ff}$. This analysis requires a model to link core properties to ambient clump properties, for which we utilize the Turbulent Core Model of McKee & Tan (2003). Then, individual star formation times are, on average, about half of the clump free-fall time. Given an expected core to star formation efficiency, $\epsilon_{\text{core}}$, of about 50% leads to estimates of $\epsilon_{ff} \approx f_{\text{dg}} \approx 10\%$.

Future improvements in this measurement have been outlined, including better temperature and thus mass estimates of the protostellar cores and confirmation of protostellar activity via analysis of outflow properties. Future work may also include extension of these methods to a larger sample of IRDCs and star-forming regions.

We thank Alyssa Goodman for constructive suggestions to the Letter. We thank Charles Lada, Richard Larson, Nick Scoville, Adam Leroy, Gus Oemler, Qizhou Zhang, John Carpenter, and Wanggi Lim for fruitful discussions. S.K. was funded by NSF award AST-1140063 while conducting this study. J.C.T. acknowledges NSF grant AST1411527. P.C. acknowledges the financial support of the European Research Council (ERC: project PALs 320620). This Letter makes use of the following ALMA data: ADS/JAO.ALMA#2015.1.00183. S. ALMA is a partnership of ESO (representing its member states), NSF (USA) and NINS (Japan), together with NRC (Canada), NSC and ASIAA (Taiwan), and KASI (Republic of Korea), in cooperation with the Republic of Chile. The Joint ALMA Observatory is operated by ESO, AUI/NRAO and NAOJ. The National Radio Astronomy Observatory is a
facility of the National Science Foundation operated under cooperative agreement by Associated Universities, Inc.

Software: Astropy (Astropy Collaboration et al. 2013), Numpy (van der Walt et al. 2011), APLpy (Robitaille & Bressert 2012), Matplotlib (Hunter 2007).

Facilities: ALMA, Spitzer.

ORCID iDs
Shuo Kong © https://orcid.org/0000-0002-8469-2029
Jonathan C. Tan © https://orcid.org/0000-0002-3389-9142
Héctor G. Arce © https://orcid.org/0000-0001-5653-7817
Paola Caselli © https://orcid.org/0000-0003-1481-7911
Francesco Fontani © https://orcid.org/0000-0003-0348-3418
Michael J. Butler © https://orcid.org/0000-0003-1031-898X

References
Alves, J., Lombardi, M., & Lada, C. J. 2017, A&A, 606, L2
André, P., Di Francesco, J., Ward-Thompson, D., et al. 2014, in Protostars and Planets VI, ed. H. Beuther et al. (Tucson, AZ: Univ. Arizona Press), 27
Astropy Collaboration, Robitaille, T. P., Tollerud, E. J., et al. 2013, A&A, 558, A33
Beltrán, M. T., & de Wit, W. J. 2016, A&ARv, 24, 6
Bergin, E. A., & Tafalla, M. 2007, ARA&A, 45, 339
Butler, M. J., & Tan, J. C. 2009, ApJ, 696, 484
Butler, M. J., & Tan, J. C. 2012, ApJ, 754, 5
Butler, M. J., Tan, J. C., & Kainulainen, J. 2014, ApJL, 782, L30
Chabrier, G., Hennebelle, P., & Charlot, S. 2014, ApJL, 796, 75
Chen, H., Burkhart, B., Goodman, A. A., & Collins, D. C. 2017, arXiv:1707.09356
Christie, D., Wu, B., & Tan, J. C. 2017, ApJ, 848, 50
Churchwell, E., Babler, B. L., Meade, M. R., et al. 2009, PASP, 121, 213
Collins, D. C., Padoan, P., Norman, M. L., & Xu, H. 2011, ApJ, 731, 59
Da Rio, N., Tan, J. C., & Jaehnig, K. 2014, ApJ, 795, 55
De Buizer, J. M., Liu, M., Tan, J. C., et al. 2017, ApJ, 843, 33
Draine, B. T. 2011, Physics of the Interstellar and Intergalactic Medium (Princeton, NJ: Princeton Univ. Press)
Federrath, C., & Klessen, R. S. 2013, ApJ, 763, 51
Goodman, A. A., Pineda, J. E., & Schnee, S. L. 2009, ApJ, 692, 91
Hennebelle, P., & Chabrier, G. 2008, ApJ, 684, 395
Hennebelle, P., & Chabrier, G. 2011, ApJL, 743, L29
Hunter, J. D. 2007, CSE, 9, 90
Johnstone, D., Di Francesco, J., & Kirk, H. 2004, ApJL, 611, L45
Kainulainen, J., Beuther, H., Henning, T., & Plume, R. 2009, A&A, 508, L35
Kainulainen, J., & Tan, J. C. 2013, A&A, 549, A53
Kenne tit, R. C., & Evans, N. J. 2012, ARA&A, 50, 531
Kong, S., Tan, J. C., Caselli, P., et al. 2017, arXiv:1701.05953
Krumholz, M. R., Klein, R. I., & McKee, C. F. 2012, ApJ, 754, 71
Krumholz, M. R., & McKee, C. F. 2005, ApJ, 630, 250
Krumholz, M. R., & Tan, J. C. 2007, ApJ, 654, 304
Kunz, M. W., & Mouschovias, T. C. 2009, MNRAS, 399, L94
Lada, C. J., Lombardi, M., & Alves, J. F. 2010, ApJ, 724, 687
Lee, E. J., Miville-Deschênes, M.-A., & Murray, N. W. 2016, ApJ, 833, 229
Li, Z.-Y., & Shu, F. H. 1997, ApJ, 475, 237
Lim, W., Tan, J. C., Kainulainen, J., Ma, B., & Butler, M. J. 2016, ApJL, 829, L19
Lombardi, M. 2009, A&A, 493, 735
Matsushita, Y., Machida, M. N., Sakurai, Y., & Hosokawa, T. 2017, MNRAS, 470, 1026
Matzner, C. D., & McKee, C. F. 2000, ApJ, 545, 364
McKee, C. F. 1989, ApJ, 345, 782
McKee, C. F., & Tan, J. C. 2003, ApJ, 585, 850
Murray, N. 2011, ApJ, 729, 133
Myers, P. C. 2015, ApJ, 806, 226
Offner, S. R. S., Clark, P. C., Hennebelle, P., et al. 2014, in Protostars and Planets VI, ed. H. Beuther et al. (Tucson, AZ: Univ. Arizona Press), 53
Ossenkopf, V., & Henning, T. 1994, A&A, 291, 943
Padoan, P., & Nordlund, Å. 2002, ApJL, 576, 870
Padoan, P., & Nordlund, Å. 2011, ApJL, 741, L22
Robitaille, T., & Bressert, E. 2012, APLpy: Astronomical Plotting Library in Python, Astrophysics Source Code Library, ascl:1208.017
Salpeter, E. E. 1955, ApJ, 121, 161
Sanhueza, P., Jackson, J. M., Zhang, Q., et al. 2017, ApJ, 841, 97
Scoville, N. Z., Sanders, D. B., & Clemens, D. P. 1986, ApJL, 310, L77
Shu, F. H., Adams, F. C., & Lizano, S. 1987, ARA&A, 25, 23
Stutz, A. M., & Kainulainen, J. 2015, A&A, 577, L6
Tan, J. C. 2000, ApJ, 536, 173
Tan, J. C. 2016, in IAU Symp. 315, From Interstellar Clouds to Star-forming Galaxies: Universal Processes?, ed. P. Jablonka, P. André, & F. van der Tak (Cambridge: Cambridge Univ. Press), 154
Tan, J. C., Beltrán, M. T., Caselli, P., et al. 2014, in Protostars and Planets VI, ed. H. Beuther et al. (Tucson, AZ: Univ. Arizona Press), 149
Tan, J. C., Kong, S., Butler, M. J., Caselli, P., & Fontani, F. 2013, ApJ, 779, 96
Tan, J. C., Kong, S., Zhang, Y., et al. 2016, ApJL, 821, L3
Tanaka, K. E. I., Tan, J. C., & Zhang, Y. 2017, ApJ, 835, 32
van der Walt, S., Colbert, S. C., & Varoquaux, G. 2011, CSE, 13, 22
Wu, B., Tan, J. C., Christie, D., et al. 2017, ApJ, 841, 88
Wu, B., Van Loo, S., Tan, J. C., & Bruderer, S. 2015, ApJ, 811, 56
Zhang, Q., Wang, K., Lu, X., & Jiménez-Serra, I. 2015, ApJ, 804, 141
Zhang, Q., Wang, Y., Pillai, T., & Rathborne, J. 2009, ApJ, 696, 268
Zhang, Y., & Tan, J. C. 2015, ApJL, 802, L15
Zhang, Y., & Tan, J. C. 2017, arXiv:1708.08853
Zhang, Y., Tan, J. C., & Hosokawa, T. 2014, ApJ, 788, 166
Zuckerman, B., & Evans, N. J., II 1974, ApJL, 192, L149