THE HIGH-MASS STELLAR INITIAL MASS FUNCTION IN M31 CLUSTERS

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ABSTRACT

We have undertaken the largest systematic study of the high-mass stellar initial mass function (IMF) to date using the optical color–magnitude diagrams (CMDs) of 85 resolved, young (4 Myr < t < 25 Myr), intermediate mass star clusters (105–107 M$_\odot$), observed as part of the Panchromatic Hubble Andromeda Treasury program. We fit each cluster’s CMD to measure its mass function (MF) slope for stars $\geq 2 M_\odot$. By modeling the ensemble of clusters, we find the distribution of MF slopes is best described by $\Gamma = +1.45^{+0.03}_{-0.06}$ with a very small intrinsic scatter and no drastic outliers. This model allows the MF slope to depend on cluster mass, size, and age, but the data imply no significant dependencies within this regime of cluster properties. The lack of an age dependence suggests that the MF slope has not significantly evolved over the first $\sim$25 Myr and provides direct observational evidence that the measured MF represents the IMF. Taken together, this analysis—based on an unprecedented large sample of young clusters, homogeneously constructed CMDs, well-defined selection criteria, and consistent modeling—implies that the high-mass IMF slope in M31 clusters is universal. The IMF has a slope ($\Gamma = +1.45^{+0.03}_{-0.06}$; statistical uncertainties) that is slightly steeper than the canonical Kroupa ($+1.30$) and Salpeter ($+1.35$) values, and our measurement of it represents a factor of $\sim$20 improvement in precision over the Kroupa IMF ($+1.30 \pm 0.7$). Using our inference model on select Milky Way (MW) and LMC high-mass IMF studies from the literature, we find $\Gamma_{\text{MW}} \simeq 1.15 \pm 0.1$ and $\Gamma_{\text{LMC}} \simeq 1.3 \pm 0.1$, both with intrinsic scatter of $\sim$0.3–0.4 dex. Thus, while the high-mass IMF in the Local Group may be universal, systematics in the literature of IMF studies preclude any definitive conclusions; homogenous investigations of the high-mass IMF in the local universe are needed to overcome this limitation. Consequently, the present study represents the most robust measurement of the high-mass IMF slope to date. To facilitate practical use over the full stellar mass spectrum, we have drafted the M31 high-mass IMF slope onto widely used sub-solar mass Kroupa and Chabrier IMFs. The increased steepness in the M31 high-mass IMF slope implies that commonly used UV- and H$\alpha$-based star formation rates should be increased by a factor of $\sim$1.3–1.5 and the number of stars with masses $> 8 M_\odot$ is $\sim$25% fewer than expected for a Salpeter/Kroupa IMF.

Key words: galaxies: star clusters; general – galaxies: star formation – Hertzsprung–Russell and C–M diagrams – Local Group – stars: luminosity function, mass function

1. INTRODUCTION

The stellar initial mass function (IMF), for $M \geq 1 M_\odot$, plays a central role in a wide variety of astrophysics. The relative number of such stars is essential to interpreting the stellar populations of star-forming galaxies across cosmic time, testing and validating theories of star formation, constraining chemical evolution models, the formation of compact objects and gravitational waves, and understanding the interplay between stars and gas (e.g., Schmidt 1959; Tinsley 1968; Searle et al. 1973; Talbot & Arnett 1973; Ostriker et al. 1974; Audouze & Tinsley 1976; Scalp 1986; Kennicutt 1998a; Elmegreen 1999; Kroupa 2001; Massey 2003; Alves et al. 2007; Fardal et al. 2007; McKee & Ostriker 2007; Wilkins et al. 2008; Pfennig Altenburg & Kroupa 2009; Kennicutt & Evans 2012; Narayanan & Dave 2012; Conroy 2013; Kroupa et al. 2013; Belczynski et al. 2014; Krumholz 2014; Madau & Dickinson 2014).

1.1. Current State of the IMF Above $\sim$1 $M_\odot$

Our best constraints on the high-mass IMF come from studies in the Milky Way (MW) and nearby galaxies. For the closest galaxies, it is possible to count individual stars and
make a direct measurement of the high-mass IMF slope. While measuring the IMF from direct star counts is straightforward in principle, it is far more complicated in practice. Accurate measurements are challenging due to a variety of observational and physical effects including completeness, dynamical evolution, distance/extinction ambiguity, stellar multiplicity, the accuracy of stellar evolution models, degeneracies between star formation history (SFH) and the IMF slope, and a paucity of massive stars in the nearby universe, all of which complicate the translation of the present day mass function (MF) into the IMF (e.g., Lequeux 1979; Miller & Scalo 1979; Scalo 1986; Mateo et al. 1993; Massey et al. 1995a; de La Fuente Marcos 1997; Maíz Apellániz & Úbeda 2005; Elmegreen & Scalo 2006; Maíz Apellániz 2008; Ascenso et al. 2009; Kruíjssen 2009; Bastian et al. 2010; De Marchi et al. 2010; Portegies Zwart et al. 2010; Kroupa et al. 2013; Schneider et al. 2015).

Observations of more distant, unresolved galaxies alleviate some of these challenges. For example, by studying an entire galaxy, effects of stellar dynamics (e.g., migration, mass segregation) and distance/extinction confusion are less important. Further, distant galaxies offer far more diversity than what is available in the very local universe and provide better opportunities to explore the high-mass IMF as a function of environment (metallicity, mass, redshift, etc.; e.g., Baldry & Glazebrook 2003; Fardal et al. 2007; Davé et al. 2011; Narayanan & Davé 2013).

However, the unresolved nature of these observations severely limits the precision and accuracy of any high-mass IMF measurement. Most integrated observations provide insufficient leverage to mitigate the SFH-IMF degeneracy, and we are forced to make simplifying assumptions about a galaxy’s recent SFH (e.g., constant, single short burst) and dust (e.g., a single value for an entire galaxy) to provide any constraint on the high-mass IMF (e.g., Miller & Scalo 1979; Elmegreen & Scalo 2006). Although studies of unresolved clusters may provide a new avenue for IMF studies (e.g., Calzetti et al. 2010; Andrews et al. 2013; although see Weidner et al. 2014 for a contrasting perspective), high-mass IMF studies using integrated light typically provide coarse characterizations rather than precise measurements.

As a result of these challenges, our current knowledge of the high-mass IMF remains remarkably poor. The widespread adoption of a “universal” Salpeter (\(\Gamma = +1.35\); Salpeter 1955) or Kroupa (\(\Gamma = +1.3\); Kroupa 2001) IMF above \(1 M_\odot\), presumed not to vary with environment has given the impression that the high-mass IMF is a solved problem (e.g., Kennicutt 1998a). However, compilations of IMF studies in the literature from resolved star counts indicate an “average value” of \(\Gamma = 1.7\) with a scatter \(\sim 0.5\) dex in IMF slope measurements that does not clearly support a “universal” IMF (e.g., Scalo 1986, 1998; Kroupa 2002; Scalo 2005).\(^{13}\) There also does not appear to be any unambiguous systematic dependence of IMF slope on environment and/or initial star-forming conditions (e.g., Bastian et al. 2010), which has significant implications for how stars form in a wide variety of physical conditions (e.g., Elmegreen 1999; McKee & Ostriker 2007; Hopkins 2013; Krumholz 2014).

Controversy over the form and universality of the IMF has persisted since the first studies in the 1950s. Following the seminal work of Salpeter (1955), a handful of studies reported similar IMFs in a number of MW clusters and field environments (Jaschek & Jaschek 1957; Sandage 1957; van den Bergh 1957), prompting early speculation about the universality of the “original MF” (e.g., Limber 1960; van den Bergh & Sher 1960). However, this early agreement with Salpeter gave way to large scatter in IMF measurements beginning in the 1970s, as several studies with more sophisticated completeness corrections and updated stellar physics highlighted the shortcomings in the first generation of IMF studies (e.g., Taff 1974; Lequeux 1979; Tarrab 1982). Furthermore, Miller & Scalo (1979) provided a comprehensive reassessment of the MF field star IMF, in light of the previously unaccounted for SFH-IMF degeneracy. They suggested a high-mass IMF of the with \(\Gamma = 1.5\) for stars \(1 M_\odot \lesssim M_\odot \lesssim 10\) and \(\Gamma = 2.3\) for stars \(M_\odot \gtrsim 10\). However, Kennicutt (1983) showed that such a steep IMF slope above \(\sim 10 M_\odot\) was incompatible with observations of \(B - V\) colors and H\alpha equivalent widths in nearby disk galaxies, suggesting that extrapolating the value of \(\Gamma = 1.5\) from \(1-100 M_\odot\) provided a better match to the data. This conclusion was further solidified in Kennicutt et al. (1994).

The number of observational high-mass IMF studies peaked between the late 1980s and early 2000s. This golden age followed the comprehensive IMF review by Scalo (1986), which concluded that systematic uncertainties were the dominant source of confusion in our poor knowledge of the IMF. Shortly after, the proliferation of uncertainties in reducing these systems, primarily by allowing for quantitative completeness corrections via artificial star tests (ASTs; e.g., Stetson 1987; Mateo et al. 1993). The 1990s saw a flurry of qualitatively new IMF studies including star counts in the LMC, spectroscopy of starburst galaxies, and IMF constraints from stellar abundance patterns (see Leitherer 1998; Massey 1998; Gilmore 2001 for extensive discussion of the literature in these areas). The sheer number and richness of IMF studies from this era precludes a detailed history in this paper, and we instead refer readers to a number of excellent papers that chronicle the history of the IMF (e.g., Scalo 1986, 2005; Massey 1998, 2003; Kennicutt 1998a; Elmegreen 1999, 2006; Eisenhauer 2001; Gilmore 2001; Zinnecker 2005; Bastian et al. 2010; Kroupa et al. 2013).

To briefly summarize, many studies from this era concluded that the high-mass IMF in the local universe was not drastically different from Salpeter over a wide range of environments (e.g., Leitherer 1998; Massey 1998, 2003; Gilmore 2001; Kroupa 2001). However, there was no clear explanation for the observed scatter in the measured IMF slopes, which contradicts a truly universal high-mass IMF. Some have argued that measurement uncertainty and systematics are the dominant source of scatter (e.g., Elmegreen 1999; Kroupa 2001, 2002; Massey 2003). However, without systematically revisiting each study, it is challenging to dismiss the scatter as solely a measurement uncertainty. As a clear counterexample, there are well-known studies of star clusters with some of the same physical properties (e.g., age, mass) that have been analyzed identically, but have very different high-mass IMF slopes (e.g., Phelps & Janes 1993), reinforcing that some of the scatter may be physical in nature. In a broader analysis of the literature, (Weisz et al. 2013) show that >75% of published resolved star IMF studies significantly underreport uncertainties relative to the theoretically possible precision, which underlines the

\(^{13}\) Although a “Kroupa IMF” has a high-mass slope of \(\Gamma = 1.3\), Kroupa (2002) argue that the slope should probably be close \(\Gamma \sim 1.7\), as most historical IMF studies do not account for unresolved binary stars, which, in principle, cause measured MF slopes to be shallower than true IMF slopes.
difficulty in disentangling systematics from intrinsic IMF variation using IMF studies in the literature. Since the early 2000s, studies of the high-mass IMF have diminished in number, as attention turned to the sub-solar regime (e.g., Chabrier 2003; Scalo 2005; Zinnecker 2005). However, focus on the high-mass IMF was recently rekindled following several reports of a systematic deficit of massive stars in lower-luminosity galaxies, based on observations of integrated broadband and/or Hα-to-UV flux ratios (e.g., Hoversten & Glazebrook 2008; Boselli et al. 2009; Lee et al. 2009; Meurer et al. 2009). Yet, whether these observations are indicative of a systematically varying IMF (e.g., Weidner & Kroupa 2005; Pflamm-Altenburg & Kroupa 2009; Kroupa et al. 2013) remains an open question, as other mechanisms such as stochastic sampling of the cluster IMF and/or bursty SFHs can also explain these observed properties (e.g., Fumagalli et al. 2011; da Silva et al. 2012; Weisz et al. 2012). Other recent studies have challenged the universality of the high-mass IMF (e.g., Dib 2014), but broader conclusions are limited by the small sample sizes.

After decades of study, the most commonly used high-mass IMF is a universal Salpeter/Kroupa IMF that is perfectly known (i.e., uncertainties are typically not included). However, this de facto standard IMF is not unambiguously supported by a large number of literature IMF studies. This tension is the direct result of a continued reliance on a rich, but heterogenous set of literature IMF studies, and can only be resolved through a large, homogenous sample of individually resolved young, massive stars.

1.2. The IMF $\geq 2-3 M_\odot$ in M31

A primary goal of the Panchromatic Hubble Andromeda Treasury program (PHAT; Dalcanton et al. 2012) is a comprehensive exploration of the stellar IMF above $\sim 2-3 M_\odot$. This 828-orbit Hubble Space Telescope (HST) multi-cycle program acquired near-UV through the near-IR imaging of $\sim 120$ million resolved stars in $\sim 1/3$ of M31’s star-forming disk (Williams et al. 2014), providing a dataset of unprecedented size and homogeneity for measuring the high-mass IMF.

The focus of the present IMF study is on M31’s young star cluster population. Star clusters provide a more direct path to measuring the IMF compared to field populations, which suffer from IMF-SFH degeneracies for stars more luminous than the oldest main sequence turnoff (e.g., Miller & Scalo 1979; Elmegreen & Scalo 2006); furthermore, the differential extinction found in the disk of M31 (e.g., Draine et al. 2013; J. J. Dalcanton 2015, in preparation), and similarly sized star-forming galaxies, adds a significant layer of complexity to modeling the field stellar populations (e.g., Lewis et al. 2015). In contrast, star clusters provide a relatively simple way to probe the IMF. Their coeval nature genuinely mitigates the effects of SFH, while their vulnerability to foreground differential extinction is minimal due to their small sizes.

In this paper, we focus on determining the MF slopes for PHAT clusters that are $\lesssim 25$ Myr old and $\gtrsim 10^5 M_\odot$ in mass using resolved star color–magnitude diagrams (CMDs). For these clusters, the detectable portion of the main sequence is well-populated from $\sim 2 M_\odot$ to $\gtrsim 25 M_\odot$, and they are of order $\sim 0.1$ relaxation times old, which mitigates the strongest effects of mass segregation, making them an ideal sample with which to study the high-mass IMF. We model the optical CMD of each cluster $k$ to constrain its power-law slope $\Gamma_k$ of the present day high-mass MF, $p(\Gamma_k)$, marginalized over other parameters of that fit (e.g., cluster age, extinction, binarity).

Subsequently, we propose a simple model for the ensemble properties of the cluster MF, with a mean, $\Gamma$, and a scatter, $\sigma_\Gamma$, and fit it to the MF probability distribution functions (PDFs) of the individual clusters. We also use this model to search for possible dependencies of $\Gamma$ on cluster mass, age, and size. We verify this approach with an extensive set of artificial clusters that were inserted into PHAT images, processed, and analyzed identically to the real clusters.

This paper is organized as follows. In Section 2, we define our cluster sample and briefly describe the PHAT photometry, ASTs, and artificial cluster sample. We outline our method of modeling the cluster CMDs in Section 3 and present the results for the ensembles of real and artificial clusters in Section 4. Finally, in Section 5, we discuss our findings in the context of current and historical views on the high-mass IMF and illustrate the implications of our results.

2. DATA

2.1. Cluster Sample

Clusters in PHAT were identified as part of the Andromeda Project, a collaborative effort between PHAT and the Zooniverse citizen science platform. As described in Johnson et al. (2015), cluster candidates were assigned probabilities of actually being a cluster based on citizen scientist ranking. These probabilities were verified using synthetic cluster results and comparison with the preliminary catalog, which was assembled by professional astronomers (Johnson et al. 2012). Extensive details of the cluster identification methodology can be found in Johnson et al. (2015).

For the present study, we have limited our analysis to the 85 clusters with CMD-based ages $\lesssim 25$ Myr and masses $\gtrsim 10^5 M_\odot$ (L. C. Beerman 2015, in preparation), and half-light radii $\gtrsim 2$ pc (Johnson et al. 2015). We focus on these young, more massive clusters in PHAT to avoid the very sparse sampling in the low-mass clusters; to limit the impact of stellar crowding in the densest clusters; to facilitate the extrapolation of the observable MF to an IMF; and to remain in the regime of high completeness (>90% for our sample; see Figure 10 in Johnson et al. 2015). We will explore each of these issues in detail in our comprehensive study of PHAT cluster MFs in a future paper.

2.2. Cluster Photometry

Photometry of PHAT clusters was performed using DOL-PHOT, a version of HSTPHOT (Dolphin 2000) with HST specific modules, following the description in Williams et al. (2014). For reasons discussed below, we made slight changes to the survey-wide photometry in that we re-reduced the entire survey using only the optical images (F475W, similar to SDSS g; and F814W, similar to I-band). The optical image and CMD

15 http://www.andromedaproject.org
16 https://www.zooniverse.org
17 http://americanodolphinism.com
18 http://archive.stsci.edu/missions/hlsp/phat/
of a typical cluster used in this study are shown in Figures 1 and 2.

For the purposes of this paper, we only use optical photometry, which contains the most information about the IMF slope, particularly when modulated by observational and computational considerations. In principle, the full spectral energy distribution (SED) of each star can increase the degree of precision to which we know its mass, and, in turn provide improved constraints on the IMF. However, the UV and IR photometry in PHAT clusters typically contains fewer than ~50% the number of stars detected in the optical, reducing their statistical utility for constraining the MF slope (see Weisz et al. 2013). PHAT UV imaging is significantly shallower (~2 mag) than the optical, even along the main sequence, resulting in many fewer sources detected. The IR also yields far fewer main sequence stars as a result of significant crowding due to the lower angular resolution of the WFC3/IR camera.

There are also subtle systematics associated with non-optical bands. For example, the UV and IR bolometric corrections are not as certain as those in the optical, particularly for massive, metal-rich stars, which can introduce systematics into our analysis.

Finally, the modest gain (at best) in precision in the MF slope of each cluster must be balanced against the computational cost of processing >10 million additional ASTs. Thus, by only reducing and analyzing the optical data, we retain virtually all statistical leverage on measuring the MF slope, but save considerably on computational costs.

However, analysis of the full stellar SEDs is valuable for searching for the most massive stars in clusters. In addition to the slope, the maximum stellar mass plays a fundamental role in defining the high-mass IMF, but is unconstrained by the approach we use to measure the MF slope (e.g., Weisz et al. 2013). An investigation of the most massive stars in the PHAT area is the subject of an ongoing study and will be presented in a future paper.

For each cluster, we ran ∼50,000 ASTs by adding stars of known luminosities one at a time (to avoid artificially enhancing crowding effects) into the PHAT optical images of each cluster. The input ASTs were uniformly distributed over each cluster’s CMD and spatially distributed approximately according to each cluster’s light profile. Extensive testing showed that such a spatial scheme was necessary to accurately characterize the completeness and photometric uncertainty profiles of the clusters, as a simple uniform spatial scheme results in overly optimistic completeness and photometric uncertainty profiles. Similarly, our adoption of a uniform distribution of ASTs in CMD space (as opposed to only matching a given cluster’s CMD) is necessary to evaluate all possible cluster models (i.e., all combinations of age, mass, metallicity, membership, IMF). In contrast, only using ASTs that roughly follow a cluster’s observed CMD would result in a poor noise model for less obvious possible combinations of cluster properties (e.g., stochastic sampling of the IMF can be degenerate with cluster age), inhibiting a comprehensive search of likelihood space. Finally, we tested the effects of the number of ASTs (ranging from 10,000 to >100,000) used on the recovery of cluster parameters. We found that 50,000 ASTs provided an optimal balance between stability of the solutions and computational efficiency.

2.3. Artificial Clusters

We use extensive sets of artificial clusters to verify the accuracy of each component of our MF determinations, including photometry, ASTs, and CMD modeling. The artificial clusters were designed to match the physical properties of our real cluster sample and span a range of physical parameters (age, mass, size, extinction) that encompass the real clusters. The artificial clusters were inserted into images at a range of galactocentric radii and environments to capture the full range of background densities in PHAT. Overall, the artificial cluster tests confirmed that we can accurately recover
the true MF from clusters in PHAT imaging, along with all other cluster physical properties.

3. METHODOLOGY

3.1. Measuring the Stellar Mass Function of a Cluster

To measure the MF of a cluster we construct synthetic models of its optical CMD using MATCH, a CMD fitting program described in Dolphin (2002). Briefly, for a given stellar evolution library, binary fraction, IMF, age, metallicity, extinction, and distance, MATCH constructs a simple stellar population (SSP) in the desired filter combinations. This synthetic SSP is then convolved with completeness and photometric biases and uncertainties as determined by the ASTs. The resulting synthetic CMD is then linearly combined with a suitable CMD of foreground/background populations, with the relative contribution of each controlled by a scale parameter. For a given set of these parameters, the synthetic Hess diagram, the binned number density of stars in the CMD, is compared to the observed CMD using a Poisson likelihood function, which is necessary to account for the effects of sparse sampling. The procedure is repeated over a large grid of parameters in order to map all likelihood space. For this analysis, we adopted the Padova stellar evolution models (Girardi et al. 2002, 2010; Marigo et al. 2008) and we have listed the other model parameters and their ranges in Table 1.

Of these parameters, we limited the metallicity to a fairly small range near Z⊙. This was done because (a) the cluster CMDs are too sparsely populated to provide robust metallicity constraints and (b) the current metallicity of M31 is known to be approximately solar, with a very weak gradient based on Hα region and supergiant abundances (e.g., Venn et al. 2000; Trundle et al. 2002; Sanders et al. 2012; Zurita & Bresolin 2012).

Similarly, we adopted a fixed binary fraction of 0.35, where the mass ratio of the stars is drawn from a uniform distribution. Historically, stellar multiplicity has been problematic for high-mass IMF determinations from luminosity functions (e.g., Maíz Apellániz & Ubeda 2005; Maíz Apellániz 2008). However, for CMD analysis, it is much less of an issue for two reasons. First, the addition of color information provides clear separation between the single and binary star sequences, minimizing confusion between the two. In contrast, when only luminosity information is available, the binary fraction and IMF slope are essentially degenerate parameters. Second, stellar multiplicity is observationally less important for high-mass stars compared to sub-solar mass stars (e.g., Weidner et al. 2009). For stars above 1 M⊙, the majority of companion stars are likely to be significantly less massive and less luminous than the primary (see Kobulnicky et al. 2014), resulting in minimal change to the CMDs. Recent work on binary mass transfer suggests that the precise details of interactions between binary (and multiple) companions may influence IMF determinations (e.g., de Mink et al. 2014; Schneider et al. 2015), but full stellar evolution models that include binary mass transfer are not yet available.

Within PHAT, we verified that our choice in binary fraction does not change the results. For example, in the case of AP 94 (Figures 2 and 3), we modeled the CMD with binary fractions of 0.35 and 0.70 and found differences in the resulting IMF slope constraints to be <0.02 dex. Tests including different binary fractions on artificial clusters yielded similarly small changes.

However, we did find that extreme binary fraction values, i.e., near 0 or 1, resulted in significantly worse CMD fits. In these two cases, the model either entirely populated the single star sequence (binary fraction = 0) or only sparsely populated it (binary fraction = 1), which substantially reduced the resemblance of the model to the observed CMD. Thus, for the present work, we adopted a fixed binary fraction of 0.35, which appears to be a reasonable value from LMC cluster studies (e.g., Elson et al. 1998).

To model background populations, we construct an empirical CMD from an annulus around the cluster that is ten times its size. The large area is necessary to accurately constrain the background scaling parameter.

Given a set of cluster model parameters (age, extinction, metallicity, and MF slope) the resulting model CMD, and the empirically derived CMD for fore/background sources are compared to the observed CMD in a likelihood sense (Dolphin 2002). The resulting posterior probabilities, p(t, A_v, \Gamma) are illustrated in Figure 3 for a typical cluster in the sample. The prior on metallicity tightly brackets solar metallicity and the cluster mass emerges as a normalization factor, rather than a fit parameter. The smoothness of the p(t, A_v, \Gamma) of each cluster allows us to interpolate in order to increase the resolution of our PDFs.

Finally, we emphasize that our analysis does not account for all sources of systematic uncertainty. Perhaps the more important systematic is the assumed stellar evolution models. Different stellar models are known to produce CMDs that can look quite different (e.g., Gallart et al. 2005), particularly for evolved phases of evolution. In principle, measuring each cluster’s MF should be done with multiple stellar models, to assess the robustness of the MF to stellar evolution effects. In practice, this is currently not possible. Aside from the Padova models, at this time, there are no other stellar libraries that

| Parameter | Range | Grid Resolution | Notes |
|-----------|-------|----------------|-------|
| \( \Gamma \) | \((-1.2, 5)\) | 0.2 dex | MF slope for stars \( \geq 2 M_\odot \) |
| Binary Fraction | 0.35 | | Fraction of Stars with a Companion |
| \( \text{Log}(\text{Age}) \) | \((6.60, 9.0)\) | 0.1 dex | Fraction of Stars with a Companion |
| \( A_v \) | \((0.0, 3.0)\) | 0.05 dex | Cluster Age |
| \( m=M_0 \) | 24.47 | | Line of Sight Extinction |
| \( \log(Z) \) | \((-0.2, 0.1)\) | 0.1 dex | Adopted Distance Modulus to M31 |
| Metallicity | | | Metallicity |

Note. Parameters and their ranges used for modeling M31 cluster CMDs. The adopted distance modulus is from the tip of the red giant branch measurement made by McConnachie et al. (2005).
include the range of ages and phases of stellar evolution necessary to analyze this diverse cluster population. If such models become available in the future, we will revisit our analysis to assess the impact of systematics. In addition to stellar models, other sources of systematics include the binary nature of clusters, where we simply assume that the MFs of falsely presumed clusters have a different $N(\Gamma_{\text{false}}, \sigma_{\Gamma_{\text{false}}})$. This mixture model mitigates the impact of erroneously identified clusters, without forcing a binary decision on which clusters have been identified with sufficient confidence.

More explicitly, the mixture model can be written as

$$p(\Gamma | \theta, \mathbf{Q}_k) = Q_k \exp\left(-\frac{(\Gamma - \Gamma_l)^2}{2\sigma^2_k}\right) + (1 - Q_k) \exp\left(-\frac{(\Gamma_{\text{false}} - \Gamma_l)^2}{2\sigma^2_{\text{false}}}\right)$$

where $\mathbf{Q}_k$ is the marginalized PDF for the MF of the $k$th object and $Q_k$ is the probability that the $k$th object is a cluster. $\Gamma$ and $\sigma^2_k$ are the mean and variance of the Gaussian of interest for the clusters, and $\Gamma_{\text{false}}$ and $\sigma^2_{\text{false}}$ are nuisance parameters for possible contaminating sources. Simply put, with this model, the more probable an object is a cluster the more it contributes to the MF parameters of interest.

This same framework also readily allows us to explore whether the mean MF slope depends on any of the other cluster properties such as age, mass, and size. Specifically, we can generalize the model to

$$\Gamma(t_k, M_k, r_k) = a_m \log M_k - M_c + a_r \log r_k + a_r \log r_c$$

where $\{t_k, M_k, r_k\}$ are the most likely age, mass, and size for the $k$th object and $\{t_c, M_c, r_c\}$ are the mean of our cluster sample in each case. We assume priors that are flat in $\Gamma$, $\log M_k$, $a_m$, $a_r$, $a_r$, over the ranges indicated in Table 1, and sample these parameters’ PDFs with the affine invariant ensemble Markov chain Monte Carlo program emcee (Foreman-Mackey et al. 2013).

4. RESULTS

The analysis laid out in Section 3 proceeds in two steps, the first leading to the MF slope measurements for individual clusters, $p(\Gamma)$, the second to the characterization of the intrinsic, i.e., error corrected, distribution of MF slopes in M31 clusters.

The result of the first step, the marginalized PDF, $p(\Gamma)$, for all 85 sample clusters is illustrated in Figure 4, showing the MF slope and its uncertainty for each cluster versus its mass. Points in the top panel are color-coded by age and those in the bottom panels by their probability of being a cluster. In both cases, the point size is proportional to the cluster’s half-light radius. Clusters with larger half-light radii have lower crowding and thus more accurate MF determination. From a physical standpoint, the size of a cluster may also reflect the environmental conditions in which it was born.

While we have used a different methodology here than Weisz et al. (2013), the error bars on the MF are similar in the two approaches, limited in both cases by the number of well-detected cluster members and by the range of stellar masses they span. For example, consider AP 94, whose CMD and

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19 https://github.com/dfm/emcee
derived physical property PDFs are shown in Figures 2 and 3. Effectively, there are ~120 stars (after background correction) between ~3 and 15 \( M_\odot \) on the upper MS that are used for MF determination. Given these numbers, from Figure 7 in Weisz et al. (2013), we expect a theoretical uncertainty of \( \pm 0.3 \) on its MF slope. In comparison, the 1-\( \sigma \) uncertainty from the marginalized MF PDF in Figure 3 is \( \pm 0.32 \), which is very close to theoretical expectation. Uncertainties in cluster membership, the exact mass range on the main sequence, etc., can explain the small difference in these two values. In general, the cluster MF slopes are generally less precise for older and lower mass clusters, and follow the expected precision relationship derived in Weisz et al. (2013).

Figure 4 also gives a qualitative impression of the ensemble properties of MF measurements. The majority of clusters have MF slopes that are near the Kroupa (1.30) or Salpeter (1.35) slopes; \( \sim 70\% \) of the clusters MF slopes are within \( \sim 1-\sigma \) of Kroupa and \( \sim 92\% \) are within 2-\( \sigma \). Remarkably, in this large sample of young clusters we find that none have extremely steep or flat MF slopes. At least for the parameter space covered by our cluster sample, extreme MF slope outliers (\( >3\sigma \)) are quite rare. The plots also show the lack of strong correlations between cluster MF slope and their physical properties. From visual inspection, it is clear that the MF slopes do not show significant trends as a function of age, mass, size, or probability of being a cluster. We list the MF slopes for all clusters in the Appendix.

These impressions from Figure 4 are quantified in the second step of our analysis, where we propose and constrain a simple model for the cluster’s distribution of MFs (see Section 3.2), and constrain its parameters through the comparison with all 85 \( p(\Gamma) \).

Figure 5 illustrates the result for the five parameters in our model: the mean MF slope, \( \Gamma \), and its intrinsic dispersion \( \sigma_\Gamma \), and the three coefficients that represent linear trends between the MF slope and cluster physical age, mass, and size. This plot shows that the mean MF slope (\( \Gamma = 1.46_{-0.07}^{+0.04} \)) is steeper than a Kroupa IMF, that the scatter (\( \sigma_\Gamma = 0.03_{-0.01}^{+0.01} \)) is consistent with expectations from a universal IMF as determined by similar analysis of artificial clusters, and that there are no significant trends between the MF slope and age, mass, or size of the clusters.

The small degree of scatter is particularly interesting, giving the visual impression of large variation between single clusters Figure 4. The typical cluster has an uncertainty on its MF slope of \( \sim 0.5 \) dex, but that ensemble scatter scales roughly as 1/\( \sim (N_{\text{clusters}}) \), indicating that a large and homogenous sample of clusters is needed to statistically identify a universal IMF. Finally, the limit on \( a_\sigma \) implies that there is no evidence that the MF slope changes between 4 and 25 Myr. We list summary statistics for each distribution in Table 2 along with the same statistics for our analysis of the population of artificial clusters.
Table 2

| Parameter | Artificial Clusters | Real Clusters | Real Clusters |
|-----------|---------------------|---------------|---------------|
| $\Gamma_1$ | 1.29$^{+0.05}_{-0.04}$ | 1.46$^{+0.07}_{-0.07}$ | 1.45$^{+0.03}_{-0.06}$ |
| $\sigma_1$ | 0.03$^{+0.02}_{-0.01}$ | 0.03$^{+0.01}_{-0.01}$ | ... |
| $\mu_{0.01}$ | 0.01$^{+0.06}_{-0.06}$ | 0.06$^{+0.10}_{-0.10}$ | ... |
| $\mu_{0.05}$ | 0.05$^{+0.08}_{-0.08}$ | 0.06$^{+0.12}_{-0.12}$ | ... |
| $\mu_{0.10}$ | -0.02$^{+0.10}_{-0.12}$ | -0.05$^{+0.11}_{-0.11}$ | ... |

Note. Column (2) values of the five parameter model for the ensemble IMF from Equations (1) and (2). In the case of $\sigma_1$, we quote the mode of its marginalized PDF, along with the upper 68% confidence interval, as the distribution is highly non-Gaussian. Column (3) value of the IMF slope by fixing all other parameters to zero, i.e., the assumption that the IMF is “universal.” The listed values reflect the median and surrounding 68% confidence interval.

There are two reasons why the artificial clusters results are not an exact delta function. First, the finite grid resolution in $p(\Gamma_1)$ sets a floor of $\sigma \sim 0.03$; second, as $\sigma_1$ is a positive definite quantity, its most likely estimate from noisy data must be significantly positive. However, this effect is quite small, as the lower limit on the scatter for the artificial clusters is consistent with the minimum possible value, given the resolution of our individual cluster PDFs.

Given that the scatter is consistent with expectations for a universal IMF, we can consider a model in which all the clusters have the same underlying MF i.e., $\sigma \to 0$. We find an ensemble-wide value of $\Gamma = 1.45^{+0.03}_{-0.05}$, which is consistent with results from the hierarchical model. For comparison, the same analysis applied to the artificial clusters yields a value of $\Gamma = 1.30^{+0.04}_{-0.04}$. Uncertainties on the two mean values begin to overlap at the $\sim 2$-$\sigma$ level. We plot the results of the “universal” MF model in Figure 6, and adopt this as the measured high-mass IMF slope of M31 for the remainder of this paper. As discussed in Section 3.1, the quoted uncertainties are only statistical in nature.

5. DISCUSSION

The quality of the PHAT data has enabled important progress on three issues: (1) quantifying the extent to which an MF measurement reflects an initial MF (IMF) estimate; (2) accurately determining what the mean IMF slope is in typical solar-metallicity (cluster) populations of a large disk galaxy; and (3) assessing the extent to which the IMF is “universal,” in the sense that it shows very little scatter in the slope, and no trends of the slope with cluster properties. The fundamental conceptual limitation of the present approach is that it only applies to clusters that remain grouped for longer than 4 Myr, i.e., they have not (yet) disrupted. Our conclusions are based on measurements of the high-mass IMF that combine a principled and systematic constraint of each cluster’s MF slope, from CMD analysis, with a probabilistic modeling framework to analyze the distribution of MF slopes.

Overall, we find that the high-mass IMF for clusters in M31 is consistent with a universal IMF. That is, as discussed in Section 4 and illustrated in Figures 5 and 6, the favored high-mass IMF model has a steeper mean slope than Salpeter/Kroupa with a small scatter that is inline with expectations for a universal IMF. Importantly, we also show that the mean MF slope in clusters of different ages changes by $\Delta \Gamma \lesssim 0.1$ per decade of age. This suggests that the MF slope is unchanged over the first $\sim 25$ Myrs, and provides direct observational evidence that the measured MF represents the IMF. In this sense, our result represents an important reference point for understanding the potential evolution of the cluster MF with cluster physical parameters (age, mass, size) and local star-forming conditions, both of which will be explored in D. R. Weisz (2015, in preparation).

Of these results, the most robust determination is the mean IMF slope. Even if we assume that all of the clusters come from the same underlying IMF (i.e., no scatter) the steep slope remains. The lack of correlations between mean IMF slope and cluster physical properties suggests that other potential sources of bias (e.g., mass segregation) contribute at a negligible level to our measurement of the mean.

In contrast, it is more challenging to interpret the significance of the upper bound on the scatter. The results of our inference suggest that a universal IMF is most probable. However, there is a tail to the distribution that extends to $\gtrsim 0.1$ dex, which exceeds the tail of the same distribution for the artificial clusters. We believe that this scatter is likely due to sources of uncertainty not included in our modeling that may artificially enhance the scatter relative to expectations from the artificial clusters. For example, mismatches between the stellar models and real stars are not captured by the artificial cluster tests. However, the degree to which model-data mismatches contribute to the scatter is hard to quantify.

Differential effects may also contribute to the observed scatter. For example, significant cluster-to-cluster variations in mass segregation or binarity could contribute to the scatter in ways not captured by the artificial clusters or our model. Similarly, we impose a single binary fraction for all masses, which may not be accurate, given that the most massive stars may be more likely to have equal mass companions than intermediate mass stars (e.g., Sana et al. 2013; Kobulnicky et al. 2014). Furthermore, dynamical effects and ejections may be coupled to the stellar binarity of a given cluster, which could...
also introduce scatter into the measure MFs (e.g., Banerjee & Kroupa 2012; Oh et al. 2015b). At the same time, given the modest size of the upper bound, we expect these effects are, at most, fairly small.

Overall, it is remarkable that the IMF slope is consistent with the universal in this mass, metallicity, and cluster density regime. As summarized in Kennicutt (1998a), quantifying the degree of scatter is one of the most critical aspects of understanding the IMF, and ultimately how stars form out of dense gas. The intrinsic scatter we find suggests that individual clusters should only minimally deviate from the fiducial value, and that extreme outliers (e.g., $\geq 3\sigma$) are quite rare. The identification of such objects may represent insight into a mode of star-formation that truly differs from what we observe in this regime.

5.1. Comparison with Other High-mass IMF Slopes in the Local Group (LG)

It is important to place our results into context with other high-mass IMF studies in the LG. Among the strongest evidence for the universality of the high-mass IMF in the LG comes from a series of papers that combine spectroscopy and photometry of resolved, massive stars in clusters and OB associations in the LMC and MW (Massey et al. 1995a, 1995b; Massey 1998, 2003).

To summarize, these studies found (a) no drastic differences between the IMF slopes of very young, $<5$ Myr star-forming regions in the MW and LMC, despite large differences in metallicity; and (b) that the resulting IMF slopes are broadly consistent with Salpeter/Kroupa. Combined, these studies represented a major milestone in our understanding of the high-mass IMF and its (in-)sensitivity to environment.

Although re-analyzing these studies is beyond the scope of this paper, we can quantify what the ensemble high-mass IMF slopes are in the LMC and MW using data from Massey (2003). The purpose is to give a general sense of how different our M31 result is from other LG environments. To perform this analysis, we extracted the necessary data from Massey (2003) and have listed it in Table 3. We simply use the model from Equation (1) with $Q_i = 1$ for each object, i.e., we know they are all bone fide clusters, to infer the mean and variance for the MW and LMC clusters. We do not model possible age, mass, size dependencies.

From this exercise we find mean values of $\Gamma_{MW} = 1.16_{-0.10}^{+0.12}$ and $\Gamma_{LMC} = 1.29_{-0.11}^{+0.11}$. In both cases, $\sigma_\Gamma \sim 0.3$–$0.4$. The large scatter in the posterior distributions is likely a reflection of underestimated uncertainties and/or systematics, as opposed to true physical variation.

Massey (2003) emphasizes that the reported uncertainties are only statistical in nature. However, even these are likely underestimated, as discussed in Weisz et al. (2013). Unfortunately, not all of these studies in the literature list both the number of stars and mass ranges needed to estimate degree of underestimation of the random uncertainties. There are also issues of systematic uncertainties in stellar models that can significantly change the masses of the most massive stars used in these studies (e.g., Massey 2011).

However, even from this cursory exercise, we see that the MW, LMC, and M31 do not appear to have drastically different IMF slopes. While this is quite remarkable given the diversity of environments, it should be tempered by the heterogenous nature of this comparison. If high-mass IMF variations are subtle, then homogenous and principled analyses are absolutely necessary to uncover them. At present, only our analysis in M31 is the result of a systematic study of the high-mass IMF over a large number of young clusters. Any comparison of the high-mass IMF slopes of LG galaxies must come from similarly homogenous data and principled analysis in order to minimize the systematics that have been persistent in IMF studies for the better part of six decades.

5.2. Practical Usage of High-mass IMF

Given the analysis presented in this paper and comparison with LG literature, we do not find strong evidence that a universal Salpeter/Kroupa IMF is the best representation of the high-mass IMF slope in the LG. Although we only rule it out at a $\sim 2\sigma$ level, there appears to be little basis for this canonical value being correct in the first place.20 Instead, for practical purposes the high-mass IMF can be represented by a single-sloped power-law with $1.45_{-0.06}^{+0.03}$ where the reported uncertainties are statistical only (see Section 3.1). Although this is only 0.1–0.15 dex steeper than Salpeter/Kroupa it does make a

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Table 3: Select Literature High-mass IMF Slopes

| Cluster       | $\Gamma_{LMC}$ | $M_{upper}$ |
|---------------|----------------|-------------|
| LH 6          | 1.7 ± 0.4      | 85          |
| LH 9          | 1.7 ± 0.4      | 55          |
| LH 10         | 1.7 ± 0.4      | 90          |
| LH 38         | 1.7 ± 0.4      | 85          |
| LH 47/48      | 1.3 ± 0.2      | 50          |
| LH 58         | 1.4 ± 0.2      | 50          |
| LH 73         | 1.3 ± 0.4      | 65          |
| LH 83         | 1.3 ± 0.5      | 50          |
| LH 114        | 1.0 ± 0.1      | 50          |
| LH 117/118    | 1.6 ± 0.2      | >120        |
| R136          | 1.3 ± 0.1      | >120        |

| Cluster       | $\Gamma_{MW}$ | $M_{upper}$ |
|---------------|----------------|-------------|
| NGC 6823      | 1.3 ± 0.4      | 40          |
| NGC 6871      | 0.9 ± 0.4      | 40          |
| NGC 6913      | 1.1 ± 0.6      | 40          |
| Berkeley 86   | 1.7 ± 0.4      | 40          |
| NGC 7280      | 1.7 ± 0.3      | 65          |
| Cep OB5       | 2.1 ± 0.6      | 30          |
| IC 1805       | 1.3 ± 0.2      | 100         |
| NGC 1893      | 1.6 ± 0.3      | 65          |
| NGC 2244      | 0.8 ± 0.3      | 70          |
| NGC 6611      | 0.7 ± 0.2      | 75          |
| Cyg OB2       | 0.9 ± 0.2      | 110         |
| Tr 14/16      | 1.0 ± 0.2      | >120        |
| h & χ Per     | 1.3 ± 0.2      | 120         |

Note. High-mass IMF slopes from the massive star studies in the MW and LMC clusters as listed in Massey (1998) and Massey (2003). The dynamic mass range of the clusters extend from $>1 M_\odot$ to the mass listed in column (3).

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20 Salpeter (1955) has had remarkable staying power, despite the fact that the power-law index of $\Gamma = 1.35$ was derived assuming the MW was 6 Gyr old. A more realistic 13 Gyr age of the MW yields a value of $\Gamma = 1.05$, which is considered an extreme outlier by modern standards (e.g., Salpeter 2005; Zinnecker 2005).
difference in the number of high-mass stars formed, and related quantities such as commonly used star formation rate (SFR) indicators. We discuss these implications further in Section 5.3.

For general usage, we recommend the adoption of the following forms of the IMF over all stellar masses. With the M31 high-mass IMF, a modified Chabrier IMF Chabrier (2003) has the form

$$\xi(m)\Delta(m) = 0.158 \frac{(1/m)}{\exp^{-0.6 \log(m) - 0.3}}$$

$$\xi(m) = cm^{-(\Gamma+1)}$$

(3)

and a modified Kroupa (2001) IMF has the form

$$\xi(m) = cm^{-(\Gamma+1)}$$

(4)

Here we have made two extrapolations to match conventional IMF definitions in the literature. First, we have extrapolated the M31 IMF down to 1 $M_\odot$. For most practical purposes, this extension from 2 $M_\odot$ to 1 $M_\odot$ affects the resulting stellar mass spectrum at the level of a few percent.

Second, we have extrapolated the M31 IMF up to 100 $M_\odot$. While our data do not provide good constraints above ~25 $M_\odot$ (see Section 1.2), there is little other information to go on for mass spectra of the most massive stars. Thus, as is commonplace in the literature, we suggest extrapolating the M31 IMF up to the highest masses, but recognize that systematic spectroscopic searches for the massive stars are absolutely critical for understanding the IMF of the most massive stars (e.g., Massey et al. 1995b; Massey. 2003).

Finally, we have assumed that the sub-solar IMF in M31 is similar to that of the Galaxy. Although not confirmed by direct star counts, Conroy & van Dokkum (2012) have shown that spectral features sensitive to the low-mass IMF yield a result that is similar to the low-mass Galactic IMF.

### 5.3. Broader Implications

In Figure 7 we illustrate the mass spectra of high-mass stars predicted by various commonly used IMF models. The plot shows the number of stars expected relative to a Kroupa IMF ($\Gamma = 1.3$) as a function of mass. For fair comparison, the slopes have all been normalized at 2 $M_\odot$, the lower main sequence mass limit of the PHAT data, and extended to 100 $M_\odot$ on the upper end. We chose this normalization, as opposed to 1 $M_\odot$ to discuss our ability to distinguish different forms of the IMF in the literature over the dynamic mass range of the PHAT data, which is indicate by the vertical dashed lines. As previously discussed, the indicated upper limit (25 $M_\odot$) is likely a lower bound due to the inability of optical-only photometry to accurately characterize the most massive stars.

The M31 high-mass IMF predicts the formation of fewer stars than Kroupa at all masses. The red solid line presents our median IMF model and shows that the fractional deficit varies from ~0.9 at ~10 $M_\odot$ to ~0.7 at 100 $M_\odot$. The red shaded region reflects the 68% confidence interval in the number of predicted high-mass stars due to uncertainty on the mean high-mass IMF slope in M31. At most masses, the uncertainty deviates from the median by ±5%, but increases to ±15% for stars >50 $M_\odot$.

The M31 high-mass IMF is similar to other high-mass IMF models in the literature. It is particularly close to the Kennicutt IMF ($\Gamma = 1.5$; Kennicutt 1983). It is also the same as the IMF of Miller & Scalo (1979) over the 1–10 $M_\odot$ range. The steepness of the Miller & Scalo (1979) IMF above 10 $M_\odot$, is known to be too extreme (e.g., Kennicutt 1983; Kennicutt et al. 1994). Further, the high-mass IMF from Scalo (1986) has a similar slope ($\Gamma = 1.6$) above ~4 $M_\odot$, but a much steeper slope below it. Finally, the M31 high-mass IMF predicts fewer high-mass stars than a Salpeter IMF for masses >2 $M_\odot$.

We further quantify differences in these IMF models in Table 4. Here we have computed the expected number of stars per $10^6$ $M_\odot$ formed at selected stellar masses assuming a Kroupa (2001) IMF below 2 $M_\odot$ and the listed IMFs above 2 $M_\odot$. This table solidifies many of the above points regarding differences in commonly used high-mass IMF models. It also highlights an important implication for core collapse supernovae, whose progenitors are believed to have masses >8 $M_\odot$ (e.g., Smartt 2009). For example, the median M31 high-mass IMF model predicts ~25% fewer stars with masses >8 $M_\odot$ compared to a Kroupa IMF.

Finally, we consider the implications for commonly used SFR indicators (e.g., Kennicutt 1998b; Kennicutt & Evans 2012). Using the Flexible Stellar Population Synthesis code (Conroy et al. 2009, 2010), we compute the luminosity to SFR conversion coefficients for GALEX far- and near-UV luminosities, along with H$\alpha$, which we list in Table 5. To do this, we have assumed a constant SFH over the last 1 Gyr, the indicated IMF, mass limits from 0.08 to 100 $M_\odot$, and case B recombination to convert the number of ionizing photons to H$\alpha$ luminosity. From Kennicutt & Evans (2012) the canonical conversion from luminosity to SFR can be written as

$$\log M_\odot(\text{M}_\odot \text{yr}^{-1}) = \log L_{H\alpha} - \log C_x$$

(5)
| IMF           | $N_\geq 2M_\odot$ | $N_\geq 8M_\odot$ | $N_\geq 20M_\odot$ | $N_\geq 50M_\odot$ | $N_\geq 90M_\odot$ | Frac. $\geq 2M_\odot$ | Frac. $\geq 8M_\odot$ | Frac. $\geq 20M_\odot$ | Frac. $\geq 50M_\odot$ | Frac. $\geq 90M_\odot$ |
|---------------|-------------------|-------------------|--------------------|-------------------|-------------------|---------------------|---------------------|---------------------|---------------------|---------------------|
| Kroupa        | 6.9               | 1.1               | 3.0                | 6.2               | 6.3               | 1                   | 1                   | 1                   | 1                   | 1                   |
| Salpeter      | 6.8               | 1.0               | 2.7                | 5.4               | 5.3               | 0.97                | 0.90                | 0.87                | 0.84                | 0.83                |
| Weisz et al.  | $6.7^{+0.1}_{-0.0}$ | $0.88^{+0.08}_{-0.04}$ | $2.2^{+0.1}_{-0.2}$ | $4.0^{+0.8}_{-0.4}$ | $3.8^{+0.9}_{-0.4}$ | $0.90^{+0.04}_{-0.01}$ | $0.74^{+0.09}_{-0.04}$ | $0.66^{+0.11}_{-0.06}$ | $0.60^{+0.13}_{-0.07}$ | $0.56^{+0.14}_{-0.06}$ |
| Kennicutt     | 6.7               | 0.82              | 1.9                | 3.5               | 3.2               | 0.87                | 0.67                | 0.57                | 0.50                | 0.46                |
| Miller & Scalo| 6.8               | 0.66              | 0.82               | 0.81              | 0.56              | 0.85                | 0.51                | 0.23                | 0.11                | 0.08                |
| Scalo         | 5.6               | 0.20              | 0.43               | 0.72              | 0.65              | 0.54                | 0.14                | 0.10                | 0.08                | 0.07                |

**Note.** The number of expected stars above the specified mass limit per $10^6 M_\odot$. For these calculations, we have adopted a Kroupa (2001) IMF below 2 $M_\odot$, the specified IMF above 2 $M_\odot$, and assumed lower and upper mass bounds for the total IMF to be 0.08 $M_\odot$ and 100 $M_\odot$. Columns (7)–(11) give the fraction of stars formed above the listed mass, relative to the standard Kroupa IMF.
where $\dot{M}$ is the SFR, $L_x$ is the luminosity over the bandpass $x$, and $C_x$ is the conversion constant over the same wavelength range. We have provided updated values for $C_x$ in Table 5.

From column (5) of Table 5, we see modest differences in the SFR indicators due to an M31 high-mass IMF. The variations are such that for a fixed luminosity, using the M31 high-mass IMF results in SFRs that are a factor $\sim 1.3$–1.5 larger.

Although the change in absolute calibration is modest, the IMF measurement in M31 provides a significant increase in the precision of the SFR. Including uncertainties the high-mass Kroupa IMF slope is $\Gamma = 1.3 \pm 0.7$. Propagating this uncertainty through Equation (5) results in an order of magnitude uncertainty in H$\alpha$ SFRs. In contrast, the precision of our IMF measurement in M31 reduces this uncertainty to be of the order unity, representing a substantial improvement in our knowledge of galaxy SFRs (under the assumption of a universal IMF).

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**APPENDIX**

In this Appendix, we have tabulated the MF slopes for each cluster used in this study. The APID numbers listed in Table 6 can be used to cross-correlate the MF slopes with other cluster properties as listed in L. Beerman et al. (2015, in preparation).

| Band | $L_x$ Units | $\log\left(C_x\right)$ (Kroupa) | $\log\left(C_x\right)$ (M31) | $M_\odot$ (M31)/$M_\odot$ (Kroupa) |
|------|-------------|---------------------------------|-----------------------------|----------------------------------|
| FUV  | erg s$^{-1}$ \(L_x\) | 43.29 \(\pm\) 0.05 | 43.14 \(\pm\) 0.04 | 1.28 \(\pm\) 0.12 |
| NUV  | erg s$^{-1}$ \(L_x\) | 43.14 \(\pm\) 0.04 | 40.91 \(\pm\) 0.07 | 1.52 \(\pm\) 0.13 |
| H$\alpha$ | erg s$^{-1}$ | 40.91 \(\pm\) 0.07 | 40.91 \(\pm\) 0.07 | 1.52 \(\pm\) 0.13 |

**Table 6** MF Slopes for Each Cluster

| APID | Median $\Gamma$ | Upper 68% Uncertainty | Lower 68% Uncertainty |
|------|----------------|------------------------|-----------------------|
| 94   | 1.37          | 0.31                   | 0.32                  |
| 98   | 2.21          | 0.37                   | 0.32                  |
| 111  | 0.01          | 1.48                   | 1.55                  |
| 249  | 1.36          | 0.67                   | 0.8                   |
| 266  | 1.4           | 0.27                   | 0.28                  |
| 285  | 1.26          | 0.25                   | 0.24                  |
| 323  | 1.05          | 0.39                   | 0.42                  |
| 390  | 1.89          | 0.26                   | 0.32                  |
| 403  | 1.39          | 0.19                   | 0.23                  |
| 431  | 1.45          | 0.38                   | 0.41                  |
| 447  | 1.27          | 0.72                   | 1.24                  |
| 452  | 2.39          | 0.41                   | 0.41                  |
| 462  | 0.09          | 0.84                   | 0.82                  |
| 522  | 1.36          | 0.29                   | 0.29                  |
| 552  | 1.46          | 0.17                   | 0.17                  |

**Note.** The cluster MFs used in this study. Values in column (2) represent the median of the marginalized PDF for each cluster’s MF slope. Columns (3) and (4) indicate the upper and lower 68% confidence intervals.

We plan to publish a larger study of cluster MFs in M31, including 1000 star clusters. In this paper, we will include a full table of MFs and cluster age/mass/extinction/size, etc.

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