Dynamic Tracking of Native Polyclonal Hematopoiesis in Adult Mice

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<u>Abstract</u>

Hematopoietic dysfunction has been associated with a reduction in the number of active precursors.

However, precursor quantification at homeostasis and under diseased conditions is constrained by

the scarcity of available methods. To address this issue, we optimized a method for quantifying a wide

range of hematopoietic precursors. Assuming the random induction of a stable label in precursors

following a binomial distribution, the estimation depends on the inverse correlation between precursor

numbers and the variance of precursor labeling among independent samples. Experimentally

validated to cover the full dynamic range of hematopoietic precursors in mice (1 to 10^5), we utilized

this approach to demonstrate that thousands of precursors, which emerge after modest expansion

during fetal-to-adult transition, contribute to native and perturbed hematopoiesis. We further estimated

the number of precursors in a mouse model of Fanconi Anemia, showcasing how repopulation deficits

can be segregated into autologous (cell proliferation) and non-autologous causes (lack of precursor).

Our results support an accessible and reliable approach for precursor quantification, emphasizing the

contemporary perspective that native hematopoiesis is highly polyclonal.

Introduction

The continuous self-renewal and differentiation of hematopoietic stem and progenitor cells (HSPCs) are fundamental to blood production. Even though rare cases exist where a single HSPC clone supports hematopoiesis, the HSPC population contributing to homeostatic hematopoiesis is usually highly polyclonal [1]. For example, the number of HSPCs actively participating in white blood cell production was estimated to range from 20,000 to 200,000 in humans [2,3]. Conversely, a decline in the number of active hematopoietic precursors has been linked to hematopoietic dysfunctions. For example, in humans older than 70 years of age, hematopoietic dysfunction coincides with an abrupt reduction in the HSPCs population actively contributing to blood production [3].

These observations underscore the association between a low number of active hematopoietic precursors and hematological dysfunction. Yet, few methods are suitable for quantifying active hematopoietic precursors in a native environment [2–7]. In mice, methods utilizing *in situ* barcodes often suffer from barcode homoplasy, preventing total precursor number estimation [8,9]. In humans, estimations based on somatic mutations and computational modeling carry high uncertainty [2,3]. The absence of precise precursor quantification methods in their natural environment hampered the study of precursor numbers across different conditions and their potential usage as predictive functional markers.

In mice, the quantification of developmental hematopoietic precursors has been achieved using mice with a Confetti cassette [10]. This cassette can randomly recombine and express one of four fluorescence proteins (FPs, being RFP, CFP, YFP, or GFP) upon Cre induction, resulting in variability in Confetti expression pattern in mice inversely correlating with precursor numbers (Figure 1A) [10]. While suitable for developmental hematopoiesis, this method has limited linear range (50-2500), impeding its potential application to adult hematopoiesis or clonal hematopoiesis. To measure hematopoietic precursors in various conditions, expansion of the detection range is desired.

Interestingly, the random induction of Confetti colors in HSPCs bears similarity with X-chromosome inactivation (XCI). In XCI, one of the X chromosomes is randomly inactivated in precursor cells, a process that adheres to a binomial distribution. Formula derivation from binomial distribution implied that higher XCI variance among individuals correlates with a smaller number of precursors at the time of inactivation (Figure 1B). As XCI is faithfully maintained in the progeny, this correlation has been used to estimate precursor numbers from mature blood cells in human and mouse, leading to the discovery of the first mutation (Tet2) that contributes to clonal hematopoiesis [6,11–13]. Nonetheless, XCI occurs exclusively in females, limiting its applicability in males. Additionally, XCI takes place during early development when few precursors are present, resulting in low sensitivity in adulthood [14].

Building on insights from Confetti mice and XCI studies, we investigated whether random induction of

Confetti FPs in precursor cells can be similarly modeled by a binomial distribution, thus the variance

of FPs inversely correlates with precursor numbers. Based on this relationship, we asked if we could

broaden the measurable range of precursor number, which would allow us to measure the numbers

outside the existing correlation range.

Examining the premises of binomial distribution and setting some assumptions, we established that

the random induction of FPs among a group of mice or cells could be modeled by a binomial

distribution. Experimental validation established a broader linear range, covering the full spectrum of

precursor numbers ($1\sim10^5$) and overcoming the prior range limit. We leveraged this correlation to

probe the number of hematopoietic precursors at homeostasis, post-myeloablation, during

developmental expansion, and in a mouse model of inherited bone marrow failure.

<u>Results</u>

<u>Binomial distribution underlies the inverse linear correlation between FP% variance and the precursor</u>

<u>numbers</u>

To broaden the limited correlation range, we aimed to elucidate the mathematical relationship between

variance and precursor numbers [10]. Inspired by XCI, we asked how the stochastic induction of

Confetti FPs in precursors may satisfy the premises of binomial distribution [15,16]. We found that

when we assume the number of precursors remain the same for the condition tested, then induction

of FPs meet the required premises : (1) the number of precursors, denoted as *n*, remains constant

within a group of mice or cells; (2) each observation (e.g. each precursor) is independent; (3) each

observation yields one of two outcomes (e.g., it signifies the presence or absence of RFP in a

precursor); (4) the probability of success, represented as *p* (e.g., the likelihood of a precursor

expressing RFP upon Confetti induction), remains consistent across all observations.

Should the induction of a FP adhere to a binomial distribution, precursor numbers can be estimated

using the following equation:

$\lg(\hat{n}) = \lg\left[\frac{(1-\widehat{FP_{mean}}\%)}{\widehat{FP_{mean}}\%}\right] - 2 \times \lg\left(\widehat{CV_{FP}}\right)$ (equation 1, see Methods)

where $\hat{n}$ signifies the estimated number of precursors, $\widehat{FP_{mean}}\%$ represents the estimated probability

of a precursor being one given FP (e.g., $\widehat{RFP_{mean}}\%$), and $\widehat{CV_{FP}}$ denotes the estimated coefficient of

variance (CV) of one given FP (e.g., $\widehat{CV_{RFP}}$). CV is the standard deviation divided by the mean, utilized

by the empirical correlation formula.

The equation proposed a direct linear relationship between the precursor number and the CVs of

individual FP, supporting the empirical correlation formula [10]. Indeed, by analyzing a published

dataset providing precursor numbers and the corresponding distribution of FPs, we found that the

correlation between *n* and individual CVs of FP exhibits superior fit (>0.93), higher than the empirical

correlation equation based on all three CVs (0.75) (Figure S1A) [10]. Moreover, according to equation

1, the y-intercept of correlation between precursor number and CV was influenced by FP_{mean}%, while

the slope was the same (-2) regardless of FP_{mean}%. Given the unequal value of RFP_{mean}%, CFP_{mean}%,

and YFP_{mean}% in the published dataset, their correlation equations showed significantly distinct y-

intercepts (p<0.0001) (Figure S1B). In contrast, the slopes of these correlation equations were similar

(p=0.49, with all 95% confidence intervals encompassing -2, the theoretical value).

In conclusion, the FP induction in precursor cells can be modeled by a binomial distribution with the

assumption that precursor numbers are constant among a group of mice. This sets the mathematical

basis for the inverse linear correlation between variance of FP% and the number of precursors.

<u>A broad range of precursor numbers correlates with variance of FP%</u>

While the correlation formula derived from binomial distribution does not impose any range limitation,

experimental errors may confound the measurement of variance. We next aimed to experimentally

confirm this correlation. To streamline the calculation, we used standard deviation instead of CV to

compute variance. The correlation between standard deviation and precursor numbers is expressed

through the following equation:

$\lg(\hat{n}) = \lg\left[\widehat{FP_{mean}}\% \times (1 - \widehat{FP_{mean}}\%)\right] - 2 \times \lg(\widehat{\sigma_{FP}})$ (equation 2, see Methods)

where $\widehat{\sigma_{FP}}$ denotes the estimated standard deviation of a given FP% (e.g. $\widehat{\sigma_{RFP}}$).

Given this equation, $\lg(\widehat{\sigma_{FP}})$ exhibits a linear and inverse correlation with $\lg(\hat{n})$. Indeed, $\lg(\widehat{\sigma_{RFP}})$,

$\lg(\widehat{\sigma_{CFP}})$ and $\lg(\widehat{\sigma_{YFP}})$ exhibited linear and inverse correlation with $\lg(\hat{n})$ in the published data (Figure

S1C) [10].

To simplify the validation, we used a two-color cell model HL-60 bearing one of the two FPs (BFP and

GFP, GFP represents non-BFP) (Figure 1C). Although the Confetti cassette recombination can

generate one of four colors, our estimation is based on a given FP (e.g. a precursor expresses RFP
or not), making this simplification justifiable.

We first proved that neither BFP nor GFP HL-60 cells had a competitive growth advantage over the
other, ensuring that HL-60 progenies mirrored the seeding population (Figure S1D). We next sorted
one to 10^5 cells into individual wells and allowed them to proliferate for at least three generations
before assessing BFP% at the end of the culture (Figure 1C). We seeded up to 10^5 cells as 1.4×10^5
non-HSC LSK and 5.2×10^3 active HSCs were previously estimated in an adult female mouse, which
corresponded to the highest possible precursor numbers in mice [17]. We considered the survival rate
of cells in wells receiving a single HL-60 cell (53% to 60%), to ensure the accuracy of the number of
cells seeded.

Consistent with the proposed correlation, as seeded number increased, BFP% standard deviation
decreased (Figure 1D and E). The variance of BFP% displayed an inverse correlation with the seeding
numbers ($R^2 = 0.996$) (Figure 1F). After normalizing for cell survival rates, the calculated numbers
closely aligned with the expected numbers (Figure 1G). We had anticipated that measurement of small
variance in high precursor numbers may be confounded by experimental errors. Yet, at 10^5 seeded
cells, the estimated number of precursors did not plateau, suggesting minimal impact from
experimental error even at 10^5 cells.

<u>Experimental practices for accurate precursor number measurement</u>
It is crucial to ensure the accuracy of variance measured because variance of FP% is solely used for
estimation. Based on the data from the two-color HL-60 model, we concluded that there were at least
three experimental practices required for accurate cell estimates: (1) exclusion of outliers; (2) sufficient
flow cytometry recorded events; (3) sufficient sample size per group. Both outliers and insufficient
recorded events inaccurately inflate sample variance, leading to underestimation of precursor
numbers (Figure S1E and F). We found that the minimum recorded events increase in tandem with
the seeded number, in contrary to the reported 500-event threshold (Figure S1F) [10].

While having a small sample size per biological replicate is possible, the variability of estimations
among replicates would be high (Figure S1G). Practically, achieving a large sample size (>20) is
cumbersome in mice. To determine a feasible sample size for relatively accurate estimates, we
resampled datapoints to calculate the amount of variability reduction as sample size increased (Figure
S1H). Our analyses revealed that five samples per group were sufficient to dramatically reduce error,
where additional increase in sample size led to marginal error reduction. We therefore used at least
five mice per biological replicates for our precursor estimations.

Inducible HSC-Scl-CreER^T-targeted Confetti labeling in the active blood precursors at the adult stage

To label hematopoietic precursors and quantify their number *in vivo*, we considered the Cre mouse
line to be crossed with Confetti.   Historically, hematopoietic precursors were thought to be
hematopoietic stem cells (HSCs) capable of long-term repopulation. However, recent studies indicate
that multipotent progenitors (MPPs) also contribute alongside HSCs in maintaining blood production
[18–21]. Hence, both HSCs and MPPs should be labeled.

Among limited Cre mouse lines capable of labeling HSCs and MPPs simultaneously at the adult stage
(such as Rosa26^{CreERT2}, Mx1-Cre, and HSC-Scl-CreER^T), we chose HSC-Scl-CreER^T because of its
preference for labeling HSPCs (immunophenotypically defined as Lin$^-$Sca-1$^+$cKit$^+$ (LSK)) [22]. To
validate the specificity of HSC-Scl-CreER^T, we generated mice possessing a single Confetti allele and
homozygous HSC-Scl-CreER^T alleles. We then examined Confetti expression one day after two-day
tamoxifen administration. We observed Confetti expression in the LSK population, consistent with prior
reports of HSC-Scl-CreER^T activity (Figure 2A, Figure S2A-C). Confetti expression was additionally
detected in T cells, NK cells, CD41$^+$ cells, non-inflammatory monocytes (SSC-A^{low}Ly6C$^-$Ly6G$^-$CD11b$^+$),
common myeloid progenitors (CMP, Lin$^-$Sca-1$^-$cKit$^+$CD16/32$^{-/low}$CD34$^+$), and granulocyte-monocyte
progenitors (GMP, Lin$^-$Sca-1$^-$cKit$^+$CD16/32$^+$CD34$^+$) [20] (Figure S2D-G). Considering the relatively
shorter lifespan of these cells compared to HSC/MPP, the labeling in these cells should minimally
impact precursor calculations, especially with a chase period post-labeling. While the Confetti
induction in the relatively long-lived T cells precludes estimation based on mature T cells, we
concluded that HSC-Scl-CreER^T remained a practical choice for labeling active hematopoietic
precursors.

The stability of Confetti labeling after induction is required for binomial distribution. Therefore, the
absence of background Cre activity after Confetti induction is critical. We determined the background
Cre activity in HSC-Scl-CreER^T/Confetti animals by two approaches: (1) detecting Confetti expression
in non-induced animals; (2) identifying cells co-expressing two Confetti FPs months after tamoxifen
induction due to cassette "flipping" from background Cre activity. We observed no Confetti-expressing
cells without tamoxifen treatment in mice up to 50 weeks old and minimal co-expression of two Confetti
FPs in induced HSC-Scl-CreER^T/Confetti animals (Figure S3A-B). Conversely, Vav-Cre/Confetti
animals constitutively express Cre, resulting in ~15% cells co-expressing two Confetti colors in
peripheral blood (PB) T cells (Figure S3C-D). Both pieces of evidence support minimal Cre
background activity in HSC-Scl-CreER^T and its feasibility to be used as a Cre driver for precursor
number calculation.

<u>Variance of FP% inversely correlates with precursor numbers *in vivo*</u>

To ascertain the inverse correlation between variance of FP% and the number of hematopoietic
precursors *in vivo*, we first investigated the variance of FP% in mice hosting various numbers of
hematopoietic precursors. We generated these mice by non-competitively transplanting 0.25×10^6 or
4×10^6 induced CD45.2$^+$ HSC-Scl-CreER^T/Confetti mouse bone marrow (BM) cells into CD45.1$^+$
recipient mice (Figure 2B). Given the constant frequency of precursors in the donor BM ($1/10^5$) and
the varying doses of donor BM cell received, these mice received approximately 2.5 ("low precursor
number") or 40 ("high precursor number") precursors [15,23–25]. While similar experiments
transplanting defined numbers of precursors have been reported, the estimation for precursor
numbers lower than 50 has not been explored.

Due to the scarcity of hematopoietic precursors in the BM, the precursor number seeded in recipients
followed Poisson distribution instead of being constant, violating premise (1) of binomial distribution.
Nonetheless, our simulations showed that the seeding number variation among recipient mice was
relatively small, therefore the variance resulting from precursor number differences still dominated and
inversely correlated with precursor numbers. A minor problem was that the estimated numbers were
not at one-to-one ratio to expected numbers at low precursor number range (<10, see Methods)
(Figure S4A).

As cells were not 100% labeled with Confetti, for clarity, we use "FP$_{PB}$%" or "FP$_{BM}$%" to represent the
frequency of a given FP in the PB or BM, and use "FP$_{Conf}$%" to represent the frequency of a given FP
in the Confetti$^+$ population. For total precursor number calculations, we normalized the number
calculated from the Confetti-labeled population to Confetti labeling efficiency.

As expected, we observed higher RFP$_{Conf}$% variance across all PB cell types in mice belonging to the
"low precursor number" groups (Figure 2C and Figure S4B). Since variance of any FP following a
binomial distribution inversely correlates with precursor numbers, we also observed higher CFP$_{Conf}$%
and YFP$_{Conf}$% variance in the "low precursor number" groups (Figure S4D-E). Following the same
principle, we observed higher RFP$_{PB}$%, CFP$_{PB}$%, YFP$_{PB}$% and Confetti% variance in "low precursor
number" mice (Figure S4F).

Applying equation 2, we found that the precursor numbers of B cells and myeloid cells were noticeably
higher in the first two months than at four months post-transplantation, suggesting transient progenitor
contribution at early time points (Figure 2D). At four months post-transplantation, the estimated
precursor numbers aligned with the expected values (in myeloid cells, 94 for "high precursor number"
groups, three for "low precursor number" groups, Figure 2D). Again, for any FP following binomial

distribution, variance of FP% inversely correlates with precursor numbers and thus can be leveraged to calculate precursor numbers (Methods). Since different FPs measured the same population of precursors, we expected the estimations from different FPs to be similar. Indeed, precursor numbers derived from RFP_{Conf}% highly correlated with those estimated from YFP_{Conf}% or CFP_{Conf}% ($R^2 = 0.885$ for CFP% in Confetti$^+$, $R^2 = 0.769$ for YFP% in Confetti$^+$), as well as those from RFP_{PB}%, YFP_{PB}% and CFP_{PB}% (Figure 2E, Figure S4G).

Consistent with PB data, higher variance of RFP_{Conf}% was also observed in the BM HSPCs of "low precursor number" mice (Figure S5A). The precursor numbers estimated from various HSPC subpopulations aligned well with each other and were consistent within the same group (Figure 2F). However, in "high precursor number" groups, estimates from BM HSPCs were lower than those derived from PB myeloid cells (27 for BM HSPC, 94 for PB myeloid cells, Figure 2F). This discrepancy may reflect uneven seeding of precursors to the BM throughout the body after transplantation, and the fact that we only sampled a part of the BM (femur, tibia, and pelvis). In summary, we validated the inverse correlation between variance of FP% and precursor numbers *in vivo*. Moreover, the quantification of precursor numbers was feasible even for precursor numbers outside the empirical correlation range.

<u>Cell frequency measurements fail to reflect the differences in precursor numbers</u>

Given that transplantation was performed non-competitively and the recipients were lethally irradiated, we expected to see minimal differences in donor chimerism between two groups, despite drastic differences in donor precursor numbers. Indeed, although we observed substantially lower donor chimerism in mice belonging to "low precursor number" groups during the first two months post-transplantation, the chimerism differences were very small by four months post-transplantation (5.5±2.8%, Figure 2G and Figure S4B). The initial differences in donor precursor numbers did not affect PB cell frequencies except for the first month post-transplantation, when higher PB myeloid frequencies was observed in the "low precursor number" groups (Figure 2H). This suggests myelopoiesis is preferred when very few precursors are available after irradiation-mediated injury (Figure 2C).

In BM, nucleated cell counts, donor chimerism and HSPC frequencies were mostly similar regardless of donor precursor numbers, suggesting that even extremely low numbers of hematopoietic precursors can still repopulate effectively a non-competitive environment (Figure S5B-D). Therefore, in cases where hematopoietic precursors expand or attrite without competition, stem cell frequencies are less informative to study precursor activity. Measuring precursor numbers is more meaningful as low number of active precursors may be hindered by compensatory proliferation.

Thousands of hematopoietic precursors contribute to native hematopoiesis

While the empirical formula, which measures 50-2500 precursors, supports quantification of precursors labeled at fetal stages, it may not be applicable to measure native hematopoiesis as the total precursor number may exceed 2500. Having validated the feasibility for measuring precursor numbers *in vivo* by Confetti animals, we next sought to investigate the number of active precursors contributing to native hematopoiesis. In HSC-Scl-CreER^T/Confetti animals, we opted to label a smaller portion of HSPCs with two-day treatment despite the potential to label up to 60% HSPCs with Confetti (CFP+YFP+RFP) by 14-day tamoxifen treatment (Figure 3A, Figure S6A). This precautionary measure aimed to mitigate potential toxicity arising from prolonged tamoxifen treatment.

Unexpectedly, we observed a successive decline of CFP$_{Conf}$% and CFP$_{PB}$% in some animals, which distorted the distribution of CFP$_{Conf}$% (Figure S6B and C). A similar decline was not observed in RFP$_{Conf}$% or YFP$_{Conf}$%, nor the RFP$_{RPF+YFP}$% (calculated by RFP$_{PB}$%/(RFP$_{PB}$%+YFP$_{PB}$%)) (Figure S6D and E). Declining CFP$_{Conf}$% was also not observed in animals where Confetti was induced during fetal development, suggesting a potential immune response to CFP in adult-induced animals (Figure S6F). Since only ~2% of myeloid cells were labeled with CFP, we reasoned that the decline of CFP in some animals was unlikely to affect the calculation of total precursor numbers based on other FPs. Therefore, we solely relied on the variance of RFP$_{RPF+YFP}$% for adult precursor calculations. In transplantation studies, precursor numbers calculated with variance of RFP$_{RFP+YFP}$% linearly correlated with those calculated with variance of RFP$_{Conf}$% (Figure S4G).

To ensure accuracy, we focused on PB myeloid cells, as (1) Confetti labeling of T cells resulted from direct induction by HSC-Scl-CreER^T, but not differentiation from HSPCs (Figure S2D); (2) Confetti labeling (RFP and YFP) of B cells did not plateau at seven months post-induction (Figure 3B); (3) myeloid cells are of shorter lifespan. In myeloid cells, we focused on precursor numbers after four months post-induction, when their labeling reached a plateau (Figure 3B). We also examined average RFP$_{RFP+YFP}$% (Figure 3C). Its stability suggested an equal contribution of RFP$^+$ and YFP$^+$ cells to the PB, a prerequisite for estimating precursor numbers from FP% distributions in the progenies (PB myeloid cells). The average RFP$_{RFP+YFP}$% in BM HSPC subpopulations mirrored those observed in PB, supporting non-biased differentiation of RFP$^+$ and YFP$^+$ precursor cells (Figure 3C).

Fitting data to equation 2, we estimated an average of 2667 precursors contributing to native myelopoiesis (Figure 3D, average of numbers at five to seven months post-induction after myeloid cell labeling plateaued). This number closely aligns with the clone number estimated by transposon-based barcodes and statistic method in granulocytes (831/30% = 2770) [1].

The average myeloid precursor number at the time of BM analysis (1958) matched the average precursor number calculated from BM myeloid progenitors (MP, Lin⁻Sca-1⁻cKit⁺) and HSPCs (1773 and 1917), but it was five-fold higher than that of LT-HSC (Figure 3E). Although LT-HSCs were conventionally believed to sustain steady-state hematopoiesis, recent studies suggest that ST-HSC and MPPs persist long-term and are the primary contributors to adult hematopoiesis [1,18,20,26]. The observed discrepancy between the number of precursors contributing to LT-HSC and those contributing to MPPs indicates that at least some MPPs were not replaced by progenitors directly differentiated from LT-HSC at seven months post-induction, affirming the long life of MPPs (Figure 3E). Furthermore, the fact that PB myeloid and BM MP precursor numbers were closer to those of MPPs than LT-HSC confirm active and long-term (at least seven months) contributions from MPPs to steady-state myelopoiesis.

Of note, the quantity of precursor does not necessarily correlate with the frequency of cell type in the BM. For example, the cell frequency of MPP2 (Lineage⁻cKit⁺Sca-1⁺Flk2⁻CD48⁺CD150⁺) in BM was very low, but MPP2 comprised a modest number of precursors, making its frequency-to-precursor-number ratio the lowest among the HSPC subtypes (Figure 3F and S6G). The frequency-to-precursor-number ratio reflects how well precursors of a particular type proliferate to expand their absolute cell count. The low frequency-to-precursor-number ratio of MPP2 suggests that it expands more poorly than other HSPC subtypes.

In summary, we detected thousands of hematopoietic precursors contributing to adult hematopoiesis. At the time of BM analysis, the number of PB myeloid precursors was comparable to those observed in BM MP and HSPCs.

Precursor numbers determined by FP% variance confirmed reduced clonality of progenitors after myeloablation

Myeloablation through 5-fluorouracil (5-FU) treatment depletes most actively cycling cells, forcing quiescent stem cells to proliferate [27]. Previous studies have indicated a significant reduction in the number of clones detected within the c-Kit⁺ population (consisting of MP and HSPC) of the BM following a single dose of 5-FU treatment [8]. However, questions arise regarding whether this observed reduction is an artifact stemming from potential under-sampling of the highly expanded c-Kit⁺ population after 5-FU treatment by single-cell sequencing. Given the quantitative estimation of a wide range of precursor numbers through equation 2, we aimed to investigate whether precursor numbers within progenitor populations indeed reduce ten days post-5-FU treatment (Figure 4A).

To compare the precursor changes post-5-FU treatment, we used the animals described in Figure 3A

as untreated (UT) benchmark, which were collected at the same age. At ten days post 5-FU injection,

the efficacy of 5-FU treatment was validated by lower PB myeloid cell frequency post-treatment and

higher frequencies of BM progenitors compared to UT animals (Figure S7A and B). The high

progenitor frequencies in the BM corresponded to the depletion of cycling cells in the BM and the

enhanced proliferation of HSPCs following 5-FU treatment.

Compared to UT animals, the precursor numbers of MP (including CMP, GMP, MEP) and HSPC

significantly decreased, confirming a reduction in clonality in the c-Kit$^+$ population (Figure 4B) [8]. By

contrast, the precursor numbers of LT-HSC did not show a decreasing trend. While transplantation

studies supported unchanged clonality of primitive stem cells after a single dose of 5-FU, similar

investigation in a native environment has not been assessed [28]. Our findings support the notion that

native LT-HSC clonality remains unaltered following one-dose 5-FU treatment. Together,

myeloablation treatment reinforces how the dynamics of precursor numbers can be tracked through

assessing Confetti pattern variations.

<u>Active lifelong precursors expand poorly in development ontogeny</u>

HSCs are thought to undergo substantial expansion in the fetal liver during fetal development [29–31].

However, a recent study employing the empirical formula challenged this notion by quantifying

endogenous lifelong hematopoietic precursors labeled at various developmental stages [32]. It

revealed limited expansion of hematopoietic precursors during the fetal liver stage (from E10 to

E15,1.8- to 2.7-fold) as well as a gradual and moderate increase from the fetal liver to the post-natal

stage (2.4- to 10-fold) [32]. Although intriguing, most post-natal measurements in this study fell outside

the empirical linear range, leaving the genuine degree of post-natal precursor expansion unknown. As

our formula quantitatively assesses a wide range of precursor numbers, we set out to directly compare

the numbers of precursor labeled at various development stages (E11.5 and E14.5, by one dose of

tamoxifen; adult-stage, benchmark shown in in Figure 3A, all analyzed at the same age) (Figure 4C).

For accuracy, we first examined the dynamics of Confetti labeling. In E14.5-induced animals, the

Confetti labeling of T cells almost doubled from one month to four months of age (4.7% at one month,

9.3% at four months), suggesting Confetti$^+$-labeled precursors contributed more to post-natal T cell

than non-labeled precursors (Figure 4D). Therefore, for T cells in E14.5-induced animals, we focused

on precursor numbers generated after three months of age. The Confetti labeling for E11.5-induced

animals remained stable for all timepoints examined (Figure S7C).

Similar to adult-induction, the distribution of average FP% was stable for fetal-induction, supporting
equal proliferation of RFP$^+$ and YFP$^+$ cells (Figure S7D). Unlike adult-induction, fetal-labeled precursor
numbers can be calculated with various combination of FP%, since the expression of CFP introduced
at the fetal stages was stable (Figure S6F). Nonetheless, for consistency, all precursor numbers were
calculated with variance of RFP$_{RFP+YFP}$%, the one employed in adult-induction animals.

The resulting precursor numbers estimated for E11.5 and E14.5-labeled cohorts were similar in all PB
cell types except at two and three months of age, echoing the previous study reporting limited lifelong
precursor expansion in the fetal liver (Figure 4E) [32]. Precursor numbers calculated from BM
subpopulations were also comparable between the two timepoints (Figure 4F). Although E9.5-labeled
clones were reported to seed non-uniformly across bones, the precursor numbers calculated from PB
myeloid cells did not significantly differ from those calculated from BM MP and HSPC in E14.5-labeled
animals (Figure 4F) [8].

For comparison between fetal- and adult-induced animals, we focused on PB myeloid cells, as adult-
induced animals had non-saturated labeling in B cells and non-HSPC-rooted Confetti labeling in T
cells (Figure 3B). We observed a relatively small increase of precursor numbers in adult-induced
animals in PB and BM compared to those labeled at the fetal stage (Figure 4E and F). The increase
between E11.5/E14.5- and adult-labeled precursors were less than ten-fold, a level similar to a
previous report (Figure 4G) [32].

Together, we detected minimal to no expansion of lifelong precursors in the fetal liver stage and minor
expansion of expansion of lifelong precursors from fetal liver to adult stage.

Fancc$^{-/-}$ mice have normal quantity of hematopoietic precursors at steady-state
Our approach provided an opportunity to investigate the number of active HSPC precursor in the native
environment of genetic mouse models. To showcase precursor quantification in genetic mouse models,
we focused on Fanconi Anemia (FA), the most common inherited bone marrow failure syndrome [33].
Current FA mouse models exhibit mostly normal adult hematopoiesis at steady state but demonstrate
reduced repopulation ability upon BM transplantation [34–39]. It remains unclear if their steady-state
blood production is sustained by a reduced number of precursors, predisposing them to repopulation
defects after transplantation.

To quantify precursor number in a mouse model of FA, we generated _Fancc$^{+/+}$_, _Fancc$^{+/-}$_ and _Fancc$^{-/-}$_
mice with a single Confetti allele and homozygous HSC-Scl-CreER^T alleles (Fancc$^{+/+}$Confetti$^{fl/+}$HSC-
Scl-CreER$^{T/T}$, hereafter _Fancc$^{+/+}$;_ Fancc$^{+/-}$Confetti$^{fl/+}$HSC-Scl-CreER$^{T/T}$, hereafter _Fancc$^{+/-}$;_ Fancc$^{-/-}$

Confetti$^{fl/+}$HSC-Scl-CreER$^{T/T}$, hereafter *Fancc$^{-/-}$*) [40]. Consistent with previous literature, we observed

similar PB cell frequencies and blood counts between *Fancc$^{+/+}$* and *Fancc$^{-/-}$* mice (Figure S8A-C) [39].

BM nucleated cell counts, as well as HSPC and MP frequency, were mostly identical, except for ST-

HSC frequencies, which were significantly reduced in *Fancc$^{-/-}$* mice compared to *Fancc$^{+/-}$* mice (Figure

S8D-E).

To label hematopoietic precursors in FA animals, Confetti expression was induced at two months of

age, and FP% was monitored over seven months (Figure 5A). The absence of *Fancc* did not affect

Confetti labeling efficiency, Confetti labeling dynamics or average FP% (RFP$_{RFP+YFP}$%), suggesting

precursor numbers in *Fancc$^{-/-}$* can be similarly calculated (Figure 5B and Figure S8F). As in adult-

induction animals, RFP$_{RFP+YFP}$% was used to estimate precursor numbers. The resulting myeloid and

BM precursor numbers were comparable regardless of *Fancc* genotype, albeit *Fancc$^{-/-}$* mice had a

slight reduction in HSPC precursor numbers (Figure 5C and D). These observations collectively

suggest that a normal number of precursors sustained blood production in *Fancc$^{-/-}$* mice.

<u>The number of precursors remained unchanged in mice transplanted with young donor cells</u>

While *Fancc$^{-/-}$* mice have similar number of precursors with their wildtype counterparts at homeostasis,

it is unknown whether their reduced repopulation ability post-transplantation stems from diminished

precursor numbers, reduced cell expansion, or a combination of both [35]. To understand the

underlying mechanism, we performed competitive transplantation using BM cells from three-month-

old Confetti-induced *Fancc$^{+/+}$*, *Fancc$^{+/-}$* or *Fancc$^{-/-}$* mice along with CD45.1^{+} competitor cells (Figure 5E,

Figure S9A). Consistent with previous studies, recipient mice of *Fancc$^{-/-}$* cells showed significantly

lower donor chimerism in the PB and BM compared to those of *Fancc$^{+/+}$* or *Fancc$^{+/-}$* cells (Figure 5F,

Figure S9B). Despite lower donor chimerism, PB and BM precursor numbers were unaffected in

Fancc$^{-/-}$ recipient mice (Figure 5G, Figure S9C-E). This suggests that reduced *Fancc$^{-/-}$* cell proliferation

instead of fewer active precursors was likely to be the cause for the reduced repopulation capacity

post-transplantation.

<u>Ageing *Fancc$^{-/-}$* mice have reduced number of lymphoid hematopoietic precursors upon transplantation</u>

Ageing *Fancc$^{-/-}$* mice have been reported to develop hematologic neoplasms, resulting in decreased

survival [41]. To determine if aging *Fancc$^{-/-}$* cells also maintain similar precursor numbers post-

transplantation, we competitively transplanted BM cells from nine-month-old Confetti-induced *Fancc$^{+/+}$*

or *Fancc$^{-/-}$* mice with CD45.1^{+} competitor cells, a stage when aging *Fancc$^{-/-}$* mice started to show

decreased survival [41] (Figure S9F).

The diminished repopulation ability of *Fancc*[-/-] cells was reaffirmed by lower PB and BM chimerism

(Figure 5H, Figure S9I). While no differences in precursor numbers were noted during the initial post-

transplantation period, a slight yet consistent reduction in lymphoid precursors was observed in *Fancc*[-]

[/-] recipient mice at three to five months post-transplantation (Figure 5I, Figure S9G and H). Changes

in precursor numbers in the BM of KO recipients were less clear as we observed high variance in

several HSPC subtypes (Figure S9J). Nonetheless, precursor numbers of MEP, LSK and ST-HSC

showed a consistent reduction. In conclusion, aging *Fancc*[-/-] mice showed a modest but consistent

loss of active PB lymphoid precursors post-transplantation, implying decreased lymphoid precursors

additionally contribute to reduced repopulation capacity as *Fancc*[-/-] mice age.

<u>Discussion</u>

The polyclonal nature of endogenous hematopoiesis imposes methodological problems on a robust

dynamic measurement. Inspired by the XCI studies, we successfully employed the correlation formula

informed by binominal distribution to quantify precursor number in native hematopoiesis. Using HSC-

Scl-CreER^T/Confetti animals, we estimate thousands of precursors contributing to adult native

hematopoiesis, a number comparable to the previous report [1]. This number resulted from a moderate

increase during fetal-to-adult transition and responded dynamically to myeloablation by 5-FU. Applying

to a mouse model of inherited bone marrow failure (Fancc$^{-/-}$ mice), we detected normal precursor

numbers at steady state and decreased lymphoid precursors upon transplantation of aging donors. In

sum, the study underscores the flexibility of the approach and the value in analyzing precursor

contributions *in situ*.

Although the linear relationship between variance of FP% and precursor numbers has been described,

its measurable range was limited [10]. Estimates outside this linear range may erroneously fall within

it, making it challenging to distinguish accurate measurements within the range from inaccurate ones

beyond it [10]. Based on binomial distribution, we expanded the measurable range to encompass the

possible dynamic range of hematopoietic precursors in mice, minimizing the likelihood of inaccuracies.

Futhermore, the linear correlation based on binomial distribution can accommodate any labeling

system that follows the underlying premises, alleviating the need for the intricate Confetti cassette.

Having established a quantitative measurement, we were initially surprised to confirm the moderate

number differences between fetal and adult-precursors seen in other report [32]. The maximum

increase observed (seven-fold observed in CMP) was substantially lower than the increase of

competitive repopulation units (CRU) from E12 to E16 determined by limited dilution assays (~38-fold)

(Figure 4G), as well as the escalation from immunophenotype-defined HSPC counts (28- to 47-fold,

from 3-5x10^3 at E14.5 to 1.4x10^5 at two months old) (Figure 4G) [17,30,42,43]. It is possible that we

overestimated E11.5 precursors, as the numbers of lifelong E11.5 precursors were substantially higher
(on average 753 PB myeloid precursor, 512 HSPC precursors) than the one to two repopulation units
estimated with transplantation [44]. Nonetheless, our results were comparable to the ~870
hematopoietic cluster cells observed at E11.5 [45]. On the other hand, we may have underestimated
precursor numbers in adult-labeled animals, as most adult HSCs are quiescent [27]. However, HSC
precursor numbers did not increase after induced proliferation by one dose 5-FU treatment (Figure
4B). In sum, our results suggest that the moderate precursor numbers difference between fetal- and
adult-stage are likely genuine, emphasizing the analysis of hematopoiesis in a native environment.
Future studies using different methods to investigate precursor activity locally is required to validate
this result.

Novel methods tracing clones *in situ* offer new opportunities to study native hematopoiesis, yet most
are challenging to apply to genetic mouse models [1,8,9,46,47]. Since only two mouse lines are
required (sometimes Cre is already used for conditional deletion of alleles), our approach is more
convenient to study hematopoiesis in mouse models of genetic disorders. Applying precursor
measurements to *Fancc$^{-/-}$* mice, we observed a minor decrease in precursor numbers after
transplantation of aging donors. Although donor chimerism differences had been linked to difference
in precursor numbers, reduced proliferation capacity may also contribute to reduced competitive
repopulation capacity [25]. In this case, precursor number analysis is necessary to differentiate these
possibilities.

Currently, concerns regarding clonal restriction in the context of FA gene therapy arises, as the
engraftment of gene-corrected stem cells has yielded marginal clinical benefits [48]. For the first time,
we demonstrated that FA cell maintaining a normal precursor quantity post-transplantation, thereby
disproving clonal attrition, including putative homing deficits post-transplantation as a cause [49].
However, it is imperative to acknowledge that the majority of murine models of FA, including the *Fancc^{-}*
$^{/-}$ model utilized herein, do not manifest spontaneous bone marrow failure, raising uncertainty
regarding the extent to which FA mouse models recapitulate the pathophysiology observed in FA
patients. Future investigations in other FA mice and studies leveraging FA patient-derived materials
will be pivotal in corroborating and validating the findings presented here.

While we implemented careful data processing procedure, one pitfall of our analyses was that we
inferred precursor numbers from their progenies, assuming uniform and linear expansion from
precursor to progenies. A recent study has showed non-uniform precursor clone sizes, although the
level of non-uniformity is low [47]. In cases where the non-uniformity is high, according to mathematical
modeling, we measured the major contributors to hematopoiesis[50]. Another potential caveat is that

the relative contribution of Confetti-labeled precursors to blood production compared to non-labeled

precursors remains unknown. Future studies using different Cre drivers will help validate the precursor

numbers for steady-state hematopoiesis.

In summary, we established a wide applicable range of the correlation between variance of FP% and

the precursor based on binomial distribution. We discovered thousands of precursors contributing to

steady-state adult hematopoiesis and validated that fetal-to-adult precursor expansion were limited.

This analysis highlights active precursor number as an important metric in both normal and genetic

mouse models.

<u>Methods</u>

<u>Variance modeling with two-color HL-60</u>

HL-60 cells were cultured with IMDM containing 20% FBS (Gemini) and 5% Penicillin-Streptomycin

(Gibco) and maintained at 1×10^5 and 1×10^6 cells/ml. Mycoplasma tests (Lonza) were performed

routinely to ensure no mycoplasma contamination. BFP-HL-60 and GFP-HL-60 were generated with

lentivirus transduction of pGK-BFP (Genscript) and LeGO-V2 (a gift from Dr. Stefano Rivella lab).

For seeding of one to 10,000 cells, sorting of HL-60 mixture was performed using a BD FACS Aria

III. For seeding of 100,000 cells, cell counting and dilution was used. After expansion, HL-60 was

fixed with BD fixation buffer before proceeding to Aurora (Cytek) or Cytoflex (Beckman Coulter) for

analysis.

<u>Mice</u>

HSC-Scl-CreER^T mice [22] were crossed with R26R-Confetti mice (B6.129P2-*Gt(ROSA)26Sor*$^{tm1(CAG-}$

$^{Brainbow2.1)Cle}$/J) to generate HSC-Scl-CreER$^{T/T}$Confetti$^{fl/+}$ (HSC-Scl-CreER^T/Confetti) animals. HSC-Scl-

CreER^T/Confetti animals were crossed with *Fancc*$^{+/-}$ mice [40] to generate HSC-Scl-

CreER$^{T/T}$Confetti$^{fl/+}$*Fancc*$^{+/+}$ (*Fancc*$^{+/+}$), HSC-Scl-CreER$^{T/T}$Confetti$^{fl/+}$*Fancc*$^{+/-}$ (*Fancc*$^{+/-}$), and HSC-Scl-

CreER$^{T/T}$Confetti$^{fl/+}$Fancc$^{-/-}$ (*Fancc*$^{-/-}$) mice. Vav-Cre was generously offered by Dr. Wei Tong

(Children's Hospital of Philadelphia). Vav-Cre/Confetti animals were generated by crossing Vav-Cre

with R26R-Confetti mice. All animals are B6 CD45.2 background unless otherwise stated. Six- to

twelve-week-old females were used for timed pregnancies. Eight-week-old mice were used for adult

Confetti induction. Both female and male mice were used indistinctly [10]. Six- to eight-week-old old

female B6 CD45.1 mice (B6.SJL-*Ptprc*^{a} *Pepc*^{b}/BoyJ, Jackson laboratories) were used as competitors

and recipients for transplantation studies. All mice were maintained in the conventional small animal

facility at the Children's Hospital of Philadelphia (CHOP). All procedures involving animals were

approved by the Institutional Animal Care and Use Committee at the Children's Hospital of

Philadelphia.

<u>Animal identification</u>

Tail snip DNA was extracted using KAPA Express Extract Kit (Roche). Genotyping PCR was performed with HotStarTaq Master Mix (Qiagen) according to manufacturer's instruction. Genotyping primers used are summarized in Table S3. To determine the zygosity of HSC-Scl-CreER^T, qPCR was additionally performed with purified tail snip DNA using SYBR™ Green Universal Master Mix (Applied Biosystems).

<u>Animal procedures</u>

For fetal inductions, timed matings of HSC-Scl-CreER$^{T/T}$Confetti$^{fl/fl}$ mice and HSC-Scl-CreER$^{T/T}$ mice were set up. The mice were separated the next morning and the noon of the day of separation was considered E0.5. Tamoxifen was delivered at 100mg/kg to the dam orally at E11.5 or E14.5. Pups were C-sectioned and cross-fostered at E18.5 due to reported delivery difficulties caused by tamoxifen treatment **[51]**. For mice used for defined number of transplantation (Figure 2), tamoxifen was delivered at 70mg/kg orally once per day for 14 days at eight-week-old. For adult induction (Figure 3), tamoxifen was delivered at 70mg/kg orally once per day for two days at eight-week-old. For one dose 5-fluorouracil (5-FU, Sigma) treatment, 5-FU was intraperitoneally injected once ten days before BM harvest (37-week-old) at 150 mg/kg. To obtain peripheral blood for Confetti analysis, mice were anesthetized using isoflurane and retro-orbitally bled or submandibularly bled for *Fancc$^{-/-}$* mice with occasional congenital eye defects**[52]** for 1 capillary of blood (50ul). For peripheral blood counts, blood was collected in EDTA tubes using similar bleeding methods and was analyzed by the Translational Core Lab (Children's Hospital of Philadelphia).

<u>Bone marrow sample processing</u>

To collect BM cells, mice were euthanized by CO_2 inhalation. Tibia, femur and pelvis were dissected, and the bone marrow cells were flushed with 26-gauge needles. The single-cell suspension generated with 18-gauge needles was then filtered through 70µm strainers. Erythrocytes in the bone marrow cells were hemolyzed by RBC lysis buffer before antibody staining or transplantation. For stem cell enrichment, bone marrow cells were further lineage depleted using EasySep™ Mouse Hematopoietic Progenitor Cell Isolation Kit (STEMCELL Technologies).

<u>Flow cytometry analysis</u>

The peripheral blood and bone marrow cells were analyzed using Aurora (Cytek) or BD FACS Aria III and the flow cytometry data was analyzed using FlowJo (Tree Star). The combinations of the following cell surface markers were used to define the peripheral blood populations: myeloid cells: CD11b$^+$ or Gr-1$^+$; T-cell: CD3ε$^+$; B-cell: B220$^+$. The following combinations of cell surface markers were used to

define the bone marrow stem and progenitor cells (Lineage/Lin: CD11b, Gr-1, B220, CD3ε, Ter119):

LTHSC: Lin$^-$c-Kit$^+$Sca1$^+$Flk2$^-$CD150$^+$CD48$^-$; MPP2: Lin$^-$c-Kit$^+$Sca1$^+$Flk2$^-$CD150$^+$CD48$^{+;}$ MPP3: Lin$^-$c-

Kit$^+$Sca1$^+$Flk2$^-$CD150$^-$CD48$^+$;   MPP4:   Lin$^-$c-Kit$^+$Sca1$^+$Flk2$^+$CD150$^-$CD48$^+$;   STHSC: Lin$^-$c-

Kit$^+$Sca1$^+$Flk2$^-$CD150$^-$CD48$^-$;   MEP: Lin$^-$c-Kit$^+$Sca1$^-$CD34$^-$CD16/32$^-$;   CMP: Lin$^-$c-Kit$^+$Sca1$^-$

CD34^{mid}CD16/32^{mid}; GMP: Lin$^-$c-Kit$^+$Sca1$^-$CD34$^+$CD16/32$^+$. For bone marrow stem and progenitor cell

analysis, DAPI (Biolegend) was used to distinguish dead cells. Representative examples of flow

cytometry gating can be found in Figure S2C. The antibodies were used at optimized dilutions listed

in Table S1.

<u>Bone marrow transplantation</u>

The day before transplantation, female CD45.1 recipient mice were lethally irradiated (5.2Gy x 2, 3h

apart) using an X-ray irradiator (Precision). On the day of transplantation, bone marrow cells from

donor mice (CD45.2$^+$) were collected under sterile conditions as described, RBC-lysed and counted

for cell number. For Figure 2 (transplantation of defined number of precursors), donor mice were

induced to express Confetti as described; only donor bone marrow cells were injected into irradiated

recipient mice; each matched high- and low- precursor number group received donor bone marrow

cells pooled from three to five mice. For transplantation of young Fancc mice, donor mice were

induced to express Confetti at E14.5; 1.5 x 10^6 donor bone marrow cells were mixed with 2.5 x 10^5

CD45.1 supporting bone marrow cells and injected into irradiated recipient mice via tail vein. For

transplantation of aging Fancc mice, donor mice were induced to express Confetti at two-month-old;

2 x 10^6 donor bone marrow cells were mixed with 5 x 10^5 CD45.1 supporting bone marrow cells as

donor population.

<u>Derivation of the correlation between variance of confetti fp% and precursor numbers based on</u>

<u>binomial distribution</u>

When a random variable adheres to binomial distribution, studies have established that the following

equation holds true:

635
$$n = \frac{p(1-p)}{\sigma^2}$$

636   where n signifies the number of precursors, p represents the probability of an individual being one of

637   the FPs (e.g., RFP), and σ^2 denotes the variance of specific FP% (e.g. $\sigma_{RFP\%}^2$)[6]. In experiments, p

638   will be estimated with the average FP% in the Confetti$^+$ cells (e.g. RFP$_{mean}$%), and σ^2 will be estimated

639   with the variance of FP% among a group of individual or mice (e.g. $\widehat{\sigma_{RFP}}^2$). Consequently, the

640   estimation of precursors number n is calculated using the following equation:

641
$$\hat{n} = \frac{\widehat{FP_{mean}}\% \times (1 - \widehat{FP_{mean}}\%)}{\widehat{\sigma_{FP}}^2}$$

A logarithmic transformation can then be performed, allowing us to establish a linear relationship

between the variance of FP% and the number of precursors:

$$\lg(\hat{n}) = \lg[\widehat{FP_{mean}}\% \times (1 - \widehat{FP_{mean}}\%)] - 2 \times \lg(\widehat{\sigma_{FP}}) \quad \text{(equation 2)}$$

In a previous study[10],

$$\sigma_{FP} = CV_{FP} \times FP_{mean}\%$$

Therefore,

$$\lg(\hat{n}) = \lg\left[\frac{(1 - \widehat{FP_{mean}}\%)}{\widehat{FP_{mean}}\%}\right] - 2 \times \lg(\widehat{CV_{FP}}) \text{ (equation 1)}$$

<u>Resampling to determine sample size per replicate</u>

FP% data generated from HL-60 was used for re-sampling. For each seeded number, FP% was

resampled for different sample sizes from all the FP% with replacement. Variance of estimated n was

then calculated by the standard deviation of the estimated precursor numbers generated from the

same sample size. Relative error was determined by dividing variance of estimated n with the average

of the estimated n. Refer to "varying well numbers.rmd" for detailed R code.

<u>Simulation to determine the effect of varying FP_{mean}% on correlation between variance of FP% and</u>

<u>precursor numbers</u>

Simulation of binomial distribution of varying FP_{mean}% (probability of being a FP) was performed in R,

generating corresponding FP% values used for variance calculation. The correlation between variance

and precursor number n was then determined by linear correlation, and the slopes and intercepts were

compared to each other. Refer to "Simulation of varying FPmean_percent.rmd" for detailed R code.

<u>Simulation to determine the effect of precursor numbers following Poisson post-transplantation</u>

Precursor numbers in individual samples were simulated to follow a Poisson distribution, where the

mean of precursor numbers is the expected precursor numbers (expected n). The induction of a FP in

each precursor was then simulated by random assignment, where probability of being a FP was set

to be FP_{mean}%. For each expected precursor numbers, variance of FP% among samples was then

calculated, and equation 2 was used to estimate precursor numbers (estimated n). The correlation

between expected n and estimated n was then plotted. Refer to "Double layer binomial simulation.rmd"

for detailed R code.

<u>Data processing and normalization</u>

Peripheral blood cell subset frequencies were normalized to the total % of granulocytes and myeloid

cells, T-cell and B-cell, to avoid an underestimate due to incomplete RBC lysis. For transplant animals,

peripheral blood CD45.1$^+$% and CD45.2$^+$% are normalized to total CD45.1$^+$% and CD45.2$^+$% to avoid

an underestimate due to incomplete RBC lysis. After normalization of cell frequencies, the sum of

Confetti% and FP$_{Conf}$% (diving FP$_{PB}$% or FP$_{BM}$% by sum of Confetti%) were then calculated for each

cohort. At each step, potential outliers were removed based on Tukey method. The variance of FP$_{Conf}$%

and average FP$_{Conf}$% are then fitted into equation 2 to calculate precursor numbers if sample size is

at least five. The precursor number is then used to compared with the minimum flow cytometry

recorded event of samples. If the estimated number is higher than the minimum flow count of samples,

the sample with the minimum flow count will be excluded, and the calculation is performed again on

the rest of samples. After exclusion, the resulting precursor number will be compared again with

minimum flow count of samples, until it is lower than the minimum flow count or sample size is smaller

than five.

<u>Statistical analysis</u>

<u>Statistical significance for precursor number estimates</u>

As the distribution of precursor numbers is not predetermined, to compare mean precursor number

differences between two conditions from a limited number of biological replicates, we employed

permutation test. For unpaired permutation test, if there are three to five biological replicates per

condition, the smallest p values possible are 0.05 ~ 0.004 ($1/\binom{6}{3}$ ~ $1/\binom{5}{10}$). For paired permutation test,

if there are three to five biological replicates per condition, the smallest p value possible are 0.125 ~

0.03125 ($1/2^3$ ~ $1/2^5$). Thus, for three to five biological replicates per condition, even though some of

the p values may not reach the commonly used alpha level (0.05), it still represents substantial number

differences. For those p values that were lowest possible but did not reach the alpha level, we

specifically labeled with a "#" in the figures and legends.

<u>Statistical analyses of other data</u>

All other two-sample statistical analyses were performed using Student's t test, if the sample was

normally distributed, or Welch's t test, when the sample was not normally distributed (F test). For

multiple comparison, one-way ANOVA or two-way ANOVA was used.

<u>Data and code availability</u>

The original Confetti% measured in the peripheral blood and the bone marrow and R code to analyze

and reproduce all the results, numeric and figures can be found at

https://doi.org/10.5281/zenodo.8222789. The re-analyzed data from a previous study can be found in

the online version of papers[10].

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S.L designed, performed experiments, analyzed data and wrote the manuscript. S.A, J.E, S.J, G.J
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Figure Legends

Figure 1. The Principle of Inverse Correlation between Variance of FP% and precursor numbers.

(A) Schematic of the Confetti cassette at the mouse *Rosa26* locus. Sequences of four fluorescence

protein are interspaced by loxP sites. Upon Cre-expression, the Confetti cassette will recombine

and express one of the four fluorescence proteins.

(B) The inverse relationship between variance of markers and precursor numbers. The distribution of

FP% is determined in the progeny to estimate the number of precursors.

(C) Workflow to validate the correlation formula between variance of FP% and precursor numbers by

a two-color cell system.

(D) BFP frequency in wells seeded with 1 to 100,000 HL-60. Each dot represents a well. Two

replicates were shown. Each replicate consists of at least 24 wells per seeded number. Error bars

represent mean ± SD.

(E) The standard deviation of BFP% in wells seeded with various numbers of HL-60. Error bars

represent mean ± SD. N = 2 replicates.

(F) The correlation between seeded numbers and standard deviation of BFP%. N = 2 replicates.

(G) The correlation between seeded numbers and the numbers estimated with standard deviation of

BFP% and equation 2. N = 2 replicates.

For (A-C), images were created with Biorender.

Figure 2. Transplantation of Defined Number of Precursors

(A) Representative flow plots showing Confetti induction by HSC-Scl-CreER^T in the LSK population.

(B) Schematic for transplantation of defined number of precursors. Briefly, 4x10^6 Confetti-induced

WBM for "high precursor number" groups, or 0.25x10^6 WBM for "low precursor number" groups

were transplanted into lethally irradiated recipient mice. Image was created with Biorender.

(C) RFP$_{Conf}$% distribution in PB myeloid cells. Each dot represents one animal. Recipient sample size

= 7-14 mice per replicate, N = 3 replicates per group. Error bars represent mean ± SD.

(D) The precursor numbers of B cells, T cells, and myeloid cells in recipient mice. Dash lines mark the

level of historically estimated transplantable clone numbers. Each dot represents a precursor

number calculated from multiple mice. Error bars represent mean ± SD.

(E) The correlation of precursor numbers calculated from RFP$_{Conf}$%, YFP$_{Conf}$% or CFP$_{Conf}$%. Each dot

represents a precursor number calculated from multiple mice.

(F) The precursor number of BM MP and HSPC in recipient mice. Dash lines represent historically

estimated transplantable clone numbers. Each dot represents a precursor number calculated from

multiple mice. Error bars represent mean ± SD.

(G) The donor chimerism of recipient mice, and the CD45.2$^+$ chimerism differences between high- and

low- precursor number groups. Each solid dot represents one animal. Each outlined dot represents

the average value of a replicate containing multiple animals. Recipient sample size = 7-14 mice

per replicate, N = 3 replicates for each precursor number group. Two-way ANOVA for paired

samples was performed.

(H) The frequency of myeloid cells in the peripheral blood of recipients. Each dot represents one

animal. Recipient sample size = 7-14 mice per replicate, N = 3 replicates for each precursor

number group. Two-way ANOVA for paired samples was performed.

ns, non-significant, * $p < 0.05$, ** $p < 0.001$.

Figure 3 Quantification of active hematopoietic precursor at steady-state

(A) Experiment schematic for Confetti induction in adult-induced animals. Image was created with

Biorender.

(B) Confetti labeling in B cells, myeloid cells and BM HSPCs. Statistics for comparisons between

labeling of month seven and other months were shown. One-way ANOVA for paired samples was

performed.

(C) RFP$_{Conf}$% in PB myeloid cells and BM HSPCs. Statistics for comparisons between labeling of

month seven and other months were shown. One-way ANOVA for paired samples was performed.

(D) The number of myeloid precursors and the average precursor number from month five to seven.

(E) The number of BM precursors. Paired permutation was performed.

(F) The frequency-to-clone ratio of bone marrow HSPCs. Paired permutation test was performed.

Each solid dot represents one animal. Each outlined dot represents the average value of a replicate

containing multiple animals. Error bars represent mean ± SD. PB sample size = 14-22 mice per

replicate; BM sample size = 8-10 mice per replicate; N = 4 replicates. ns, non-significant, * $p < 0.05$,

** $p < 0.001$.

Figure 4. Number of precursor post-5-FU treatment and along developmental ontogeny

(A) Experiment schematic for one-dose 5-FU treatment. Image was created with Biorender.

(B) The number of BM precursors post-5FU treatment. Permutation test was performed. For UT,

sample size = 8-10 mice per replicate; N = 4 replicates; for 5-FU, sample size = 6-8 mice per

replicate; N = 5 replicates.

(C) Experiment schematic for Confetti induction at fetal stages. Image was created with Biorender.

(D) The Confetti labeling of B cells, T cells and myeloid cells. Each dot represents one animal. Each

outlined dot represents the average value of a replicate containing multiple animals. Sample size

= 4-11 mice per replicate; N = 5 replicates.

(E) The number of precursors in PB myeloid cells, B cells and T cells. Permutation test was performed.

For E14.5 and adult-induction, N = 4 replicates; for E11.5, n = 3 replicates.

(F) The number of BM precursors. Each dots represent one precursor number. Permutation test was

performed. For E14.5 and adult-induction, N = 4 replicates; for E11.5, N = 3 replicates.

(G) The relative fold change increase of precursor numbers from fetal- to adult-stage, as well as fold

change of CRU and cell counts.

Error bars represent mean ± SD. ns, non-significant, * $p < 0.05$, ** $p < 0.001$.

Figure 5. Precursor numbers in a mouse model of FA

(A) Experimental workflow to generate $Fancc^{+/+}$ and $Fancc^{-/-}$ mice, induce Confetti expression and

track precursor number dynamics. Image was created with Biorender.

(B) Confetti labeling in PB B cells and myeloid cells. Each dot represents one animal. Each outlined

dot represents the average value of a replicate containing multiple animals. Dashed lines connect

the values for the same replicate.

(C) The number of PB myeloid precursor.

(D) The number of BM precursor.

(E) Experiment schematic for competitive transplantation. Image was created with Biorender.

(F) PB donor chimerism in recipient mice of young FA donors. Each dot represents one animal.

Dashed lines connect the average donor chimerism of three replicates. For $Fancc^{+/+}$ or $Fancc^{+/-}$, n

= 10 per replicate, N = 3 replicates; for $Fancc^{-/-}$, N = 7-9 per replicate, N = 3 replicates. Two-way

ANOVA was performed.

(G) The number of PB myeloid precursor post-transplantation of young FA donors. Dashed lines

connect the average precursor numbers. Permutation test was performed. N = 3 for each group.

(H) PB donor chimerism recipient mice of aging FA donor cells. Each dot represents one animal.

Dashed lines connect the average donor chimerism of three replicates. For both genotype, N = 10

per replicate, N = 3 replicates. Two-way ANOVA was performed.

(I) The number of PB B precursor post-transplantation of aging FA donors. Dashed lines connect the

average precursor numbers. Permutation test was performed. N = 3 for each genotype.

For (A-C), $Fancc^{+/+}$, sample size = 11-17 mice per replicate, N =5 replicates; $Fancc^{+/-}$, sample size =

10-22 mice per replicate, N =7 replicates; $Fancc^{-/-}$, sample size = 9-17 mice per replicate, N =4

replicates. For (D), $Fancc^{+/+}$, sample size = 5-9 mice per replicate, N =5 replicates; $Fancc^{+/-}$, sample

size = 6-12 mice per replicate, N =5 replicates; $Fancc^{-/-}$, sample size = 6-11 mice per replicate, N =4

replicates. Error bars represent mean ± SD. nd, not determined; ns, non-significant; * p < 0.05, ** p <

0.001. # represents lowest p value possible for permutation test.

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<u>Supplementary Figure Legend</u>

Figure S1: Establishing the correlation between variance of FP and precursor numbers.

(A) The correlation between precursor numbers and CVs of individual FP or all CVs (original

correlation formula), replotted from Ganuza et al., 2017.

(B) FP_{mean}% in a previously published dataset, replotted from Ganuza et al., 2017.

(C) The correlation between precursor numbers and standard deviation (σ) of individual FP, replotted

from Ganuza et al., 2017.

(D) The BFP% in HL-60 mix during two weeks of growth.

(E) The effect of outliers on precursor number estimation. Left, BFP% in samples with/without outliers.

Quartiles by Tukey method was shown. The outlier is highlighted by circle. Middle, standard

deviation of BFP% with/without outliers. Right, estimated precursor numbers with/without outliers.

(F) The changes of standard deviation when recorded events varied. "&" denotes the potential

overestimated standard deviation due to low flow recorded events.

(G) The variance of estimated precursor numbers when varying sample sizes were used.

(H) The relative error of estimation as sample sizes increased. Each dot represents data from one

seeded number. Error bars represent mean ± SD.

1020

Figure S2: The induction specificity of HSC-Scl-CreER^T

(A) Experiment schematic to examine the induction specificity of HSC-Scl-CreER^T. N =10 for

induced animals, N =3 for non-induced animals. Image was created with Biorender.

(B) Flow gating strategy for BM HSPC and MPs.

(C) Confetti labeling in BM HSPCs. Error bars represent mean ± SD.

(D) Flow gating strategy for BM lineage cells.

(E) Confetti labeling in BM lineage cells. Error bars represent mean ± SD.

(F) Flow gating strategy to examine spleen cells.

(G) Confetti labeling in spleen cells. Error bars represent mean ± SD.

See Methods and Table S1 for the detailed antibody, fluorophore and marker combination used for

each cell type.

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Figure S3: Analysis of HSC-Scl-CreER^T background Cre activity

(A) Representative flow plot showing Confetti expression in PB T cells in 48-50-week-old HSC-Scl-

CreER^T/Confetti animals compared to an animal induced with one-day tamoxifen treatment.

(B) Quantification of RFP and YFP% in PB T cells. N =8 for non-induced 48-50-week-old animals.

(C) Representative flow plot for Confetti labeling in PB T cells post HSC-Scl-CreER^T-mediated
induction.
(H) The level of CFP$^+$ and RFP$^+$ double positive cells in PB T cells of animals induced by HSC-Scl-
CreER^T compared to Vav-Cre/Confetti animals analyzed at the same age. N =3 for Vav-Cre
animals, N= 8 for HSC-Scl-CreER^T animals.
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Figure S4: Variance of FP% inversely correlates with precursor numbers *in vivo*
(A) The correlation between expected numbers and the estimated numbers calculated from FP% of
precursors following Poisson distribution. See Methods for simulation details.
(B) Flow gating strategy for PB B cell, T cell and myeloid cell. See Methods and Table S1 for the
detailed antibody, fluorophore and marker combination used for each cell type.
(C) RFP$_{Conf}$% distribution in B cell and T cell.
(D) YFP$_{Conf}$% distribution in myeloid cell.
(E) CFP$_{Conf}$% distribution in myeloid cell.
(F) RFP$_{PB}$%, CFP$_{PB}$%, YFP$_{PB}$% and Confetti% distribution in myeloid cells.
(G) Linear correlation of PB precursor numbers calculated from RFP$_{Conf}$%, RFP$_{PB}$%, CFP$_{PB}$%,
YFP$_{PB}$%, Confetti% or RFP$_{RFP+YFP}$%, Each dot represents a precursor number calculated from
multiple mice.
For (C-F), each dot represents one animal. Recipient sample size = 7-14 mice per replicate, N =3
replicates per group. Error bars represent mean ± SD.
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Figure S5 BM analysis for high- and low-precursor number groups
(A) RFP$_{Conf}$% distribution in BM HSPCs. Each dot represents one animal
(B) BM donor chimerism.
(C) BM nucleated cell numbers. N =15 mice for high precursor number group, n= 14 for low precursor
number group. Unpaired t-test was performed.
(D) BM HSPC and MP frequency in donor CD45.2$^+$ population. Two-way ANOVA for paired samples
was performed.
For (B) and (D), each solid dot represents one animal; each outlined dot represents the average value
of a replicate containing multiple animals. Except for (C), recipient sample size = 7-14 mice per
replicate, N =3 replicates per group. For all graphs, Error bars represent mean ± SD. ns, non-significant,
* $p < 0.05$.
1072

Figure S6 Analysis of FP induction in adult-induced animals

(A) Confetti labeling in LSK after 14days of tamoxifen treatment. N =5 animals. Error bars represent

mean ± SD.

(B) CFP_{Conf}% in PB myeloid cells.

(C) CFP_{PB}% in PB myeloid cells.

(D) $RFP_{RFP+YFP}$% in PB myeloid cells.

(E) RFP_{PB}% and YFP_{PB}% in PB myeloid cells.

(F) CFP_{Conf}% in PB myeloid cells of fetal-induced animals

(G) BM HSPC frequencies. Error bars represent mean ± SD.

For (B-E) sample size = 14-22 mice per replicate; for (F), sample size = 4-11 mice per replicate; for

(G) sample size = 8-10 mice per replicate. For (B-E) and (G), N =4 replicates. For (F), N =5 replicates.

1084

Figure S7: Analysis related to 5-FU treatment and Confetti fetal induction

(A) PB myeloid cells frequencies before 5-FU treatment and ten days post-5-FU treatment. Dashed

lines connect the values for the same replicate. Sample size = 6-8 mice per replicate, N =5

replicates for 5-FU. Paired t test was performed.

(B) The BM HSPC frequencies of UT and 5-FU treated groups ten days post-5-FU treatment. For UT,

sample size = 8-10 mice per replicate, N =4 replicates; for 5-FU, sample size = 6-8 mice per

replicate, N =5 replicates.

(C) Confetti% in PB in E11.5-labeled animals. Dashed lines connect values for the same replicate.

(D) The distribution of FPs in Confetti[+] population of B cells. Data showing average from all mice. N

=24-34 for E14.5, N =7-24 for E11.5. Error bars represent mean ± SD.

For (A-C), each solid dot represents one animal; each outlined dot represents the average value of a

replicate containing multiple animals. ns, non-significant; * p < 0.05, ** p < 0.001.

1097

Figure S8 Analysis of FA mice at steady-state

(A) PB myeloid cell frequencies. Each dot represents one animal. Each outlined dot represents the

average value of a replicate containing multiple animals.

(B) PB B cell frequencies. Each dot represents one animal. Each outlined dot represents the average

value of a replicate containing multiple animals.

(C) Complete blood count analysis. Each dot represents one animal. N =13 for *Fancc*[+/+], N =16 for

Fancc[-/-] mice. Unpaired t test was performed.

(D) BM HSPC frequencies. Each dot represents one animal. Each outlined dot represents the average

value of a replicate containing multiple animals. One-way ANOVA was performed.

(E) BM nucleated cells counts. Each dot represents one animal. Unpaired t test was performed. N =12

for *Fancc*[+/+]and *Fancc*[-/-] mice.

(F) RFP$_{RFP+YFP}$% in PB myeloid cells and B cells. Each dot represents one animal. Each outlined dot

represents the average value of a replicate containing multiple animals.

For (A-B) and (F), Fancc$^{+/+}$, sample size = 11-17 mice per replicate, N =5 replicates; Fancc$^{+/-}$, sample

size = 10-22 mice per replicate, N =7 replicates; Fancc$^{-/-}$, sample size = 9-17 mice per replicate, N =4

replicates. For (D), Fancc$^{+/+}$, sample size = 5-9 mice per replicate, N =5 replicates; Fancc$^{+/-}$, sample

size = 6-12 mice per replicate, N =5 replicates; Fancc$^{-/-}$, sample size = 6-11 mice per replicate, N =4

replicates. Error bars represent mean ± SD. ns, non-significant; * $p < 0.05$.

1116

Figure S9 Analysis of FA recipient mice

(A) Young *Fancc* donor BM HSPC frequency analysis. N = 5 for Fancc$^{+/+}$ or Fancc$^{+/-}$, N = 2 for

Fancc$^{-/-}$.

(B) Donor chimerism in recipient BM of young *Fancc* donor. Each dot represents one animal. Each

outlined dot represents the average value of a replicate containing multiple animals. Two-way

ANOVA was performed. Recipient sample size = 7-10 mice per replicate, N =3 replicates per

group.

(C) Number of PB B cell precursor in recipients of young FA donor.

(D) Number of PB T cell precursor in recipients of young FA donor.

(E) Number of BM HSPC precursor in recipients of young FA donor.

(F) Ageing *Fancc* donor BM HSPC frequency analysis. N = 3 for Fancc$^{+/+}$, N = 3 for Fancc$^{-/-}$.

(G) Number of PB myeloid cell precursor in recipients of aging FA donor.

(H) Number of PB T cell precursor in recipients of aging FA donor.

(I) Donor chimerism in recipient BM of ageing *Fancc* donor. Each dot represents one animal. Each

outlined dot represents the average value of a replicate containing multiple animals. Two-way

ANOVA was performed. Recipient sample size = 7-10 mice per replicate, N =3 replicates per

group.

(J) Number of BM HSPC precursor in recipients of aging FA donor.

For (C-E), (G-H) and (J), N =3 replicates per group; permutation test was performed; dashed lines

connect the average precursor numbers. Error bars represent mean ± SD. nd, not determined; ns,

non-significant; * $p < 0.05$, ** $p < 0.001$. # represents lowest p value possible for permutation test.

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Figure 1

Figure 2

Figure 3

Figure 4

Figure 5

Figure S1

Figure S2

Figure S3

Figure S4

Figure S8

Figure S9