Comparison of Lymphocyte Subset Populations in Children From South Africa, US and Europe

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Background: Typically, African healthcare providers use immunological reference intervals adopted from Europe and the United States (US). This may be inappropriate in a setting with many differences including exposure to different environmental stimuli and pathogens. We compared immunological reference intervals for children from Europe and the US with South African children to explore whether healthy children living in settings with high rates of infectious diseases have different baseline immunological parameters.

Methodology: Blood was taken from 381 HIV-uninfected children aged between 2 weeks and 13 years of age from a Child Wellness Clinic in an informal settlement in Cape Town to establish local hematological and lymphocyte reference intervals for South African children. Flow-cytometry quantified percentage and absolute counts of the B-cells, NK-cells, and T-cells including activated, naïve, and memory subsets. These parameters were compared to three separate studies of healthy children in Europe and the US.

Results: Increased activated T-cells, and natural killer cells were seen in the younger age-groups. The main finding across all age-groups was that the ratio of naïve/memory CD4 and CD8 T-cells reached a 1:1 ratio around the first decade of life in healthy South African children, far earlier than in resource-rich countries, where it occurs around the fourth decade of life.

Conclusions: This is the largest data set to date describing healthy children from an African environment. These data have been used to create local reference intervals for South African children. The dramatic decline in the naïve/memory ratio of both CD4 and CD8 T-cells alongside increased activation markers may indicate that South African children are exposed to a wider range of environmental pathogens in early life than in resource-rich countries. These marked differences illustrate that reference intervals should be relevant to the population they serve. The implications for the developing pediatric immune system requires further investigation.

Keywords: lymphocyte, immunophenotype, reference intervals, intervals, resource-limited, South Africa, pediatric
INTRODUCTION

Understanding the immune system in relation to disease pathogenesis and management of infectious diseases is particularly pertinent to children, who are born with an immature immune system that develops rapidly in early life. Understanding what is “normal” in a healthy child should be related to their living environment, and particularly to infectious diseases exposure.

Typically healthcare laboratories, clinical practitioners, and researchers in resource-limited settings use reference intervals adopted from studies that have acquired data from healthy children in resource-rich countries (1–3). These may not be appropriate for assessing health and disease in children from Africa or other distinct environments.

Antiretroviral therapy is now recommended for all patients infected with HIV. Our work has identified that the age and CD4 count at treatment initiation is critical for estimating immune recovery (4), however these CD4 projections are based upon immune reference intervals generated in Europe and USA. Ethnic origin, genetics, climate, altitude, nutrition, and environmental pathogen exposure (5, 6) may influence hematological and immunological parameters, and can vary widely between continents and individual countries. Differences in hematological subsets between resource-rich and resource-limited countries have been reported, including non-genetic neutropenia (7), lower platelet counts (8, 9), and other parameters such as hemoglobin levels, red blood cell counts, haematocrit levels, mean corpuscular volume, and white blood cell counts (10–15).

Differences in lymphocyte parameters, lower CD4 and higher CD8 subsets (12), along with reduced naïve proportions and increased activated CD4 and CD8 T-cells are reported in Ethiopian compared to Dutch adults (16). CD4 percentages are lower in children from Cameroon (17), Kenya (18), Uganda (19), and Malawi (20) compared to European or US reference intervals. There are also differences between African countries in levels of CD8 T-cells, B-cells, and NK cells (21). While local reference intervals in Africa are increasingly being established (11, 12, 17, 18, 22), the range of immunological parameters is limited and statistical comparison across populations has not been explored in detail.

Quantitative differences in immune parameters between children and adults were established relatively recently. For instance, an adult has ~3,000 cells/microlitre of lymphocytes in the peripheral blood, while in children, lymphocyte cell-count/microlitre rises from birth to a maximum of 9,500 between 6 months and one year of age (23, 24), it then follows an exponential decline as the child grows into adulthood. This may be due to several inter-related factors: the progressive involution of the thymus, exposure to antigens, and switch from naïve-to-memory associated with immunological “learning,” change in body size and blood volume associated with growth and the progressive age-related replacement of primary thymic production by peripheral cell division (25). There is currently no agreed standard for the most accurate way to represent these data, however representation of lymphocyte distribution during development is probably best done with mechanistic non-linear modeling rather than using age-categories or empirical methods that represent age as a continuous variable (26).

This study has established how immunological phenotypes change with age in healthy South African children and how this compares with published pediatric data from three resource-rich countries: The Netherlands, Germany and the US (27–29). The potential impact of this work is discussed.

MATERIALS AND METHODS

Participants

Three hundred and eighty-one children aged from 2 weeks to 13 years were recruited from a “Child Wellness Clinic” (CWC) at a community health clinic in an informal settlement of Cape Town, South Africa. The CWC was established primarily as a research clinic, which also aimed to benefit the participants and the wider community for health promotion, education, and screening. Attendance at the CWC was voluntary, and the criteria for recruitment were that the child was well at the time with no chronic medical condition or prescription medications, registered at the health clinic, and attended with their biological mother and hand-held medical record. Maternal HIV-exposure was not excluded. Informed consent was obtained in English or via translator in Afrikaans or Xhosa. The session included clinical history and examination by a pediatrician, plotted anthropometry, assessment of vaccination status (with catch-up as needed), and provision of nutritional supplements and a food voucher. Each participant had phlebotomy of 2–3 mls of blood used for rapid HIV-antibody analysis (Alere Determine®, 4th Generation), full blood count and basic immunophenotyping. HIV-infected children were not included.

Stellenbosch University granted ethical approval (M12/01/005) and permission for the study was given by Cape Town Department of Health.

Laboratory Testing

Blood samples were taken between 9 a.m. and 1 p.m., with 500 μL of the original samples in EDTA couriered at room temperature by air to Johannesburg and processed the following morning at the National Health Laboratory Service, Charlotte Maxeke Johannesburg Academic Hospital (SANAS M0109). Immunophenotypic analysis was performed at the Johannesburg flow cytometry laboratory according to standard operating procedures. Directly labeled antibodies CD3 APC, CD3 FITC, CD16 PE, CD19 FITC, CD45 PerCP, CD45RO PE, CD45RA FITC, HLA Dr APC (Becton Dickinson Immunocytometry Systems (BDIS), San Hose, CA), CD4 FITC, CD8 PE, and CD56 PE (Beckman Coulter, Inc. Miami, Florida) were added in pre-titrated manufacturer optimized concentrations to tubes with 50 μL of well-mixed whole blood. Stained samples were vortexed once and incubated for 30 min. Red blood cells were then lysed using FACS Lysing Solution (BDIS, San Hose, CA). All samples were run on a Becton Dickinson FACS Calibur™ and acquired, and analyzed, using CellQuest™ Pro software. Prior to analysis,
basic daily flow cytometer set-up included assessment of Calibrite™ 3 and Calibrite™ APC beads (BDIS, San Hose, CA) to monitor laser, optics, fluidic alignment, linearity, and instrument performance for the FACSCalibur™, according to the manufacturers’ standards. Listmode data was stored for retrospective analysis. External CD4 Quality Assessment for CD4 testing was performed through the U.K. NEQAS Immune Monitoring scheme and the NHLS CD4 African Regional External Quality Assessment Scheme (30). Lymphocyte subsets were expressed as a proportion of total lymphocytes determined using bright CD45 expression and side scatter. Specific lymphoid subsets assessed included: total CD3, CD3+/CD4+, CD3+/CD8+, CD3-/CD56+, CD16+/56+, CD3+/HLA DR+, CD3+/CD4+/HLA DR+, CD3+/CD8+/HLA DR+, CD3+/CD4+/45RA+, CD3+/CD4+/45RO+, CD3+/CD8+/45RA+, CD3+/CD8+/45RO+, and CD19+ lymphocytes (Supplement 1A: Gating Strategy to analyse lymphocyte subsets; Supplements 1B, C: Examples of age-related distributions of populations of CD8 and CD4). Absolute cell counts were obtained using a dual platform method; total lymphocyte counts on all samples were obtained on a Beckman Coulter LH750 hematology analyser. All laboratory work and data analysis performed were blinded.

**Studies for Comparison**

Three independent studies were used to compare the lymphocyte subsets from our population of healthy South African children with those from the US and Europe: Shearer et al. [US (29)], Comans-Bitters et al. [The Netherlands (27)], and Huenecce et al. [Germany (28)]. Shearer et al. and Comans-Bitters et al. used age-categorization (albeit different choices of age-groups), whereas Huenecce et al. presented their data using single exponential regression analysis. We therefore compared these lymphocyte populations according to the presentation of data in each publication. The US and European studies were selected for comparison because the first two are currently being used for reference intervals in South Africa and the latter enabled comparison of the populations using exponential regression techniques. The immunological parameters available for comparison are listed in Table 1.

**Statistical Methods**

Subgroups of CWC children with clinical characteristics of interest were compared using Wilcoxon Rank Sum tests to determine whether their lymphocyte subsets differed from the rest of the CWC participants. Comparisons were made between CWC individual age-groups matched to Shearer et al. (29) and

### Table 1: Corresponding immunological parameters for comparison between the healthy South African cohort and three published studies: Comans et al. (29), Shearer et al. (28), and Huenecce et al. (30).

| Subset                          | CWC          | Comans et al. | Shearer et al. | Huenecce et al. |
|--------------------------------|--------------|---------------|----------------|-----------------|
| No. children                    | 381          | 358           | 851            | 80              |
| Age range                       | 2 wks−13 yrs | 1 wk−16 yrs   | Birth−18 yrs   | 2 mths−18 yrs   |
| Country                         | South Africa | Netherlands   | USA            | Germany         |
| Total lymphocytes               | SSC/CD45 and FSC/SSC | FSC/SSC | FSC/SSC | FSC/SSC |
| CD3+ T-cells                    | CD3+         | CD3+          | CD3+           | CD3+            |
| CD4+ T-helper                   | CD3+CD4+     | CD3+CD4+      | CD4+           | CD4+CD4+        |
| CD8+ Cytotoxic                  | CD3+CD8+     | CD3+CD8+      | CD8+           | CD3+CD8+        |
| CD4/CD8 ratio                  | CD3+CD4+/CD3+CD8+ | NA           | CD3+CD4+/CD3+CD8+ | CD3+CD4+/CD3+CD8+ |
| Naive CD4+                      | CD4+CD45RA+  | NA            | CD4+CD45RA+    | CD3+CD4+CD45RA+ |
| Memory CD4+                     | CD4+CD45RO+  | NA            | CD3+CD4+CD45RO+ | CD3+CD4+CD45RO+ |
| Naive/Memory ratio of CD4+      | CD4+CD45RA+/CD4+CD45RO+ | NA         | CD3+CD4+CD45RA+/CD28+ | CD3+CD8+CD45RA+/CD28+ |
| Naive CD8+ T-cells              | CD8+CD45RA+  | NA            | CD8+CD45RA+    | CD3+CD4+CD45RO+ |
| Memory CD8+                     | CD8+CD45RO+  | NA            | CD3+CD4+CD45RO+ | CD3+CD8+CD45RO+ |
| Naive/Memory ratio of CD8+ T-cells | CD8+CD45RA+/CD8+CD45RO+ | NA    | CD3+CD8+CD45RA+ /CD28+ | CD3+CD8+CD45RO+ /CD28+ |
| Activated T-cells               | CD3+HLADR+   | CD3+HLADR+    | CD4+HLADR+     | CD3+CD4+HLADR+ |
| Activated CD4+                  | CD3+CD4+HLADR+ | NA          | CD8+HLADR+     | CD3+CD8+HLADR+ |
| Activated CD8+                  | CD3+CD8+HLADR+ | NA         | CD19+           | CD19+           |
| CD19+ B-cells                  | CD19+HLA-DR+ | CD19+         | CD16+CD56+     | CD16+CD56+      |
| Natural Killer cells            | CD3-CD56+    | NA            | NA             | CD3-CD56+       |
|                                | CD16+CD56+   | NA            |                |                 |
|                                | CD3-CD56+    | NA            |                |                 |

Both cell count per ul and percentage is available for each parameter.
Comans-Bitter et al. (27) data (median and 10/90th or 5/95th centiles, respectively) using the Chi-squared goodness-of-fit test, and for differences between the fitted median distributions of each cell marker group using the non-parametric Wilcoxon Rank Sum test.

Comparisons were also made using the results of regression analyses. For this purpose single and double exponential models were fitted to the data. For the single exponential model a similar technique was employed to that used by Huenecke et al. (28), [i.e., the cell counts or percentages of each lymphocyte subgroup were regressed against subject age using a three parameter exponential model: $f(t) = \beta_0 + \beta_1 \left(1 - e^{-\beta_2 t}\right)$, where $t$ is age (in weeks) and the betas ($\beta_0, \beta_1, \beta_2$, etc.) are the constants in the equation describing each lymphocyte subset]. Best-fit (median) and 95% CI parameters were determined by minimizing the sum of the square of residuals using MS-Excel’s Generalized Reduction Gradient non-linear solver function and a constraint precision of 0.0001. This was also done in R version 3.5.x using packages nlme and nlra. To account for the non-normal distribution of the residuals, upper and lower 95% confidence intervals (CIs) were fitted to the data independently.

The double exponential model was formulated from the single exponential model as described above. This model is defined as $f(t) = \frac{\beta_0 + \beta_1 \exp(-\beta_2 t)}{1 + \exp(-\beta_3 t)}$, where $t$ is the age. The dependent, (i.e., CD marker, variables in this model allows for growth at smaller values of $t$). For larger values of $t$, the denominator $1 + \exp(-\beta_3 t)$ approaches 1 and thus, the model reduces to the single exponential model. The average of $\beta_1$ and $\beta_2$ approximates the counts of the cell-markers at birth. The $\beta_3$ parameter determines the shape of the function. The double exponential model was sufficiently flexible to model either a simple asymptotic reduction, or an initial rise and then fall in CD markers over time, as has been seen in prior mechanistic studies (23, 31).

### RESULTS

All 381 children recruited from the CWC were included in this study and population characteristics are described in Table 2. Lymphocyte distributions of these South African children are represented in Figure 1 using three different presentations of the data: age-category histograms [using the age-categories from Comans-Bitter et al. (29) and Shearer et al. (28)], and single and double exponential regression lines. The double exponential regression lines appear to follow the data and histogram more closely than the single exponential regression.

### What Is “Immunologically Normal?”

Should children with clinical conditions that could theoretically alter the immunophenotype be included as “normal” for this population? Figure 2 illustrates the distribution of CD4 and CD8 (cells/μL) from the CWC recruits including subgroups of clinical conditions common in this population, that might affect the child’s developing immune system and thereby influence the spread of data and reference intervals derived. These conditions include (a) past history of a serious childhood illness [e.g., TB, meningitis ($n = 11$)]; (b) acute recent illness within the past month but more than a week ago [e.g., upper respiratory tract infections, gastroenteritis ($n = 69$)]; (c) maternal infections during pregnancy [e.g., TB, HIV, syphilis ($n = 28$)]; and (d) prematurity <32 weeks ($n = 13$). The exact exponential fit of the regression line for each lymphocyte phenotype examined (as per Table 1) did not appreciably change when these four clinical subsets were in turn removed from the analysis, therefore justifying the inclusion of these children.

“Maternal conditions during pregnancy” was further divided to explore the association of maternal HIV on the lymphocyte subsets ($n = 14$). Differences were detected using the Wilcoxon rank sum test between these 14 children and the rest of the cohort with lower B-cells (CD19+HLADR+) and lower memory CD4 T-cell (CD3+CD4+CD45RO+) in the HIV-exposed children. However, removing them from the entire dataset did not affect the exponential regression curves, therefore these children were included as part of this “healthy” population.
Comparison of Single Exponential Regression Curves of Lymphocyte Populations Between South African and German (28) Children

Exponential fits were obtained for all lymphocyte subsets listed in Table 1 and a selection are illustrated in Figure 3 using the single exponential fit for purpose of comparison. Absolute cell count curves for the lymphocyte subsets either initially increased or simply descended asymptotically with age. Best-fit, 5 and 95% confidence interval exponential regression curves were fitted for the CWC cohort (black lines) and the Huenecke data from healthy German children for comparison (red lines). A trend toward higher absolute counts of CD8 T-cells and B-cells were seen in South African compared to German children, however significantly higher for NK-cells ($p = 0.002$) and activated CD8
FIGURE 2 | Distribution of CD4 and CD8 (cells/μL) in healthy children from the CWC including clinical subsets of conditions that might be presumed to influence the spread of data. Red circles = past history of a serious childhood illness [e.g., TB, meningitis (n = 11)]; Blue circles = acute recent illness within the past month but more than a week ago e.g., upper respiratory tract infections, gastroenteritis (n = 69); Purple circles = maternal conditions during pregnancy e.g., TB, HIV , syphilis (n = 28); Yellow circles = prematurity <32 weeks (n = 13). Black lines represent the best-fit double exponential regression curves with 5 and 95% confidence intervals.

T-cells (p = 0.001, Figure 3). Increased CD4 or CD8 activation was associated with decreasing naïve/memory ratio in the South African cohort (cor −0.15, 95% CI [−0.25 −0.05], p = 0.003 and cor −0.23, 95% CI [−0.32 −0.13], p < 0.0001 respectively).

Marked differences were also seen for both CD4 and CD8 memory populations particularly within the first 3 years of life, as illustrated by the change in naïve/memory ratios with age in Figure 4 (respectively p = 0.07 and 0.01 overall). While Huenecke et al.’s data suggests naïve/memory ratios do not reach a 1:1 status (28) until around the third decade of life, it is apparent that this occurs within the first decade in our South African cohort.

Age-Categorized Data Between South African and US or Dutch Children

Consistent with the above results, the most significant finding when examining the distribution across all age-groups (denoted by the overall p-value in Table 3), were the increased proportions of CD4 and CD8 T-cells memory subsets across the entire age range of South African children compared to the US and European counterparts. These children also had lower CD4 and CD8 naïve subsets, particularly at <1 year of age. As illustrated by Figure 4, the data in Table 3 shows that both CD4 and CD8 T-cell naïve/memory ratios differ dramatically at less than a year of age, but not significantly so thereafter.

Table 3 illustrates a significantly higher CD3 HLA-DR+ % in South African children aged 1 week to 15 months, and 10–16 years compared to children from the Netherlands; and significantly higher CD8 HLA-DR+ % in South African children aged 2–5 months compared to children from the US. Additional differences were identified between lymphocyte subset distributions within individual age-categories in South African children vs. those from the US and The Netherlands including: lower B-cells in children aged from 15 months to 13 years; and higher percentage of CD8 T-cells in 10–16 years olds.

DISCUSSION

The dearth of local pediatric reference range data in South Africa (32) prompted this study to establish a relevant local set laboratory reference values to ensure that health care, treatment, and monitoring is appropriate for the population of children being cared-for (33). The data we have generated may be representative of lymphocyte subsets in children living in resource-limited communities who are more likely to be exposed to significant diseases, such as TB and HIV, than their counterparts in resource-rich countries.

Our main finding was the dramatic difference in naïve/memory ratios of T-cell populations in South African compared to US and European children. Parity between these populations of cells was reached some three decades earlier than observed in the German population (30). This has been noted before (31, 34), and could be explained by a reduction in thymic output with depletion of the naïve T-cell pool and/or accompanied by expansion of memory cells as naïve T-cells encounter antigen and memory populations proliferate (35).

Until now, characterization of this transition throughout the first decade of life has not been described, nor compared across continents where genetics, nutrition, and environmental antigenic exposure differ extensively.

This increased rate of decline with age in naïve/memory ratio of T-cell populations seems most likely to be due to the induction of immune-activation by increased exposure to environmental pathogens as seen in the South African study population. This is reflected by the increased proportions of CD8 T-cells, natural killer cells and activated T-cells were
demonstrated in healthy South African children compared to their US or European counterparts, and increased CD4 or CD8 activation was indeed associated with decreasing naïve/memory ratio. Environmental exposure to common pathogens such as herpesvirus, cytomegalovirus, and Epstein-Barr virus may drive the switch from naïve to memory T-cells; however exposure per se
FIGURE 4 | Comparison of naïve:memory ratios of CD4 and CD8 T-cells between South African and German children (30). (A) Ratio of naïve/memory CD4+ T-cells in cells/μL. (B) Ratio of naïve/memory CD8+ T-cells in cells/μL. Black dots represent individual data-points from the South African cohort. Black lines represent the best-fit single exponential regression curves with 5 and 95% confidence intervals for the South African cohort. Red lines indicate the same but for the data of Huenecke et al. (30).

may not be the only factor and the abundance of environmental pathogens (36), poor nutritional status and high levels of microbial translocation (37) may also drive the immune response to such pathogens (38, 39). Background immune-activation may be particularly relevant in the current climate of the COVID19 pandemic, whereby pre-existing immune activation might predispose the individual to a more inflammatory response to a new pathogen than a child with an unactivated immune system (40).

It is not possible in our study to determine whether differences seen are related to environmental exposure or genetics. The European studies for comparison do not describe ethnicity, and although the US study reports the majority of their cohort to be of African-American race (58%), the genetics are likely to differ from an African population. While our cohort broadly represents the general population of South Africa in terms of socio-economic background, it does not represent ~25% of the South African population that are relatively wealthier with well-equipped and sanitized home and school environments (41), and thereby might have comparatively less disease exposure and potentially different “normal” immunological phenotypes compared to the participants of the CWC.

The inclusion of the children with histories of significant illnesses, maternal infections during pregnancy, recent illnesses, and prematurity <32 weeks gestational age in the CWC healthy cohort might be a source of debate. However, since the prevalence in the CWC cohort is similar to the study population and there is no clear biological evidence to implicate the effect of these conditions on the child’s developing immune system we considered it acceptable to include these conditions. The sub-analyses performed on the 14 children who were born to HIV-infected mothers did not affect the overall regression curves in this study due to the small number and age distribution of the group. A larger study is warranted to explore these potential differences in more detail, especially since only 3.7% of children in our cohort were HIV-exposed compared to recent estimates of 30% of infants born in the public sector in South Africa (42, 43). This low rate of HIV-exposure might be explained by the fact that the clinic was promoted as a “healthy child” clinic where a HIV test would be done on all children, and this may have deterred HIV-infected mothers.

There are limitations to the outcomes of the comparison of lymphocyte subsets between South African children and the three other studies in children from the US and Europe. These three cohorts come from contrasting environments, and the data is generated from studies that used non-identical methodologies. There are multiple practical factors that might influence data derived from flow cytometric studies including sample transport, storage, and preparation, choice of fluorochrome-conjugated antibodies and immunological markers to define subsets of interest. These factors make direct comparison between such studies challenging. In an attempt to minimize the effect of these factors, exponential regression curves were used to compare the changes in immunological parameters across the age-range examined and this approach should help to reduce the influence of confounding factors.

When multiple statistical comparisons are performed, as done in Table 3, there is potential that significant differences detected may be due to chance rather than biological plausibility. These calculations were not adjusted for multiple comparisons because the covariate data from the other studies was not available, and adjustment would have been unlikely to add additional information of value (44). A combination of statistical approaches have been applied to the analysis of these data-sets, however regardless of the statistical approach...
TABLE 3 | Differences in cell marker measurements within specific age-groups between South African children and North- American (Shearer et al.) or Dutch children (Coman et al.).

| Age groups | Overall p-value |
|------------|-----------------|
| **PART A: B-cells** | |
| CD19% |  |
| CWC | 1 wk-2 mths | 2-5 mths | 5-9 mths | 9-15 mths | 15-24 mths | 2-5 yrs | 5-10 yrs | 10-16 yrs |
| Shearer | 16 (9–23) | 17 (11–25) | 18 (13–26) | 18 (12–28) | 20 (11–32) | |
| Coman | 13 (8–21.1) | 19.1 (9.7–31.4) | 19.6 (12.6–38.3) | 18.5 (10.6–29.1) | 20.1 (10.6–32.6)* | |
| Payne et al. | 20.1 (10.6–32.6)* | 15.8 (9.9–30.2)* | 13 (9–18.4)* | 11.8 (7.8–16.6) | 0.05 |
| **PART B: Activation markers and cytotoxic T-cells** | |
| CD8% | 1 wk-2 mths | 2-5 mths | 5-9 mths | 9-15 mths | 15-24 mths | 2-5 yrs | 5-10 yrs | 10-16 yrs |
| Shearer | 18.1 (10.7–28.9) | 19.4 (11.8–28.6) | 22.5 (12.1–31.7) | 24.3 (13.3–34.6) | 24.2 (13.8–34.9) | |
| Coman | 4.7 (2.6–12.1)** | 6 (2.6–20.1)** | 8.5 (3.2–19.3)** | 9.5 (3.7–19.2)** | 9 (2.9–18.5) | |
| Payne et al. | 5 (1–38) | 3 (1–9) | 3 (1–7) | 4 (2–8) | 6 (3–12) | 6 (3–13) |
| Shearer | 6.9 (1–9.9) | 1.0 (0.6–1.8) | 1.5 (0.9–2.2) | 1.3 (0.8–2.0) | 1.1 (0.6–1.6) | 0.3 (0.2–0.6) |
| **PART C: Naïve and memory subsets** | |
| CD4+CD45RA% | 0–3 mths | 3–6 mths | 6–12 mths |
| Shearer | 72.9 (51–83.8)* | 78.4 (62.4–84.4) | 73.9 (63.8–81.7) |
| Coman | 4.7 (2–20.6) | 9.6 (2.5–31.7)** | 16.9 (6–69.3) |
| Payne et al. | 72 (62.6–81.6) | 63.8 (47–76.4) | 54.7 (40.2–67.8) |
| Shearer | 14.4 (5.5–28.5) | 12.5 (6–23.5) | 8.8 (5.5–21.4) |
| **(Continued)**
employed, the main findings of rapid and early transition of naïve CD4 and CD8 T-cells to their respective memory populations in the CWC, was concordant across all three of the compared international studies. Although our study has several limitations, it provides a starting point for exploring differences in immunological phenotypes and the optimal way to characterize the lymphocyte distributions as they change with age. We have illustrated a double exponential model that account for the possibility that cell counts may increase and peak during the 1st year of life with subsequent decline.

The purpose of the CWC was to collect immunological data from healthy South African children to establish local reference intervals (33), since in South Africa clinicians and laboratories had been using a combination of the reference intervals published by Comans-Bitter et al. and Shearer et al. from the US and Europe. A number of important differences between the CWC and these international studies were found, highlighting the value of having contextually appropriate reference intervals available. Although no gross difference was identified in the numbers of lymphocyte subsets most commonly used in clinical practice such as CD4, CD8, and CD4/CD8 ratios, a dramatic and significant difference was demonstrated in the rapid early decline of the naïve/memory ratios of both CD4 and CD8 T-cells alongside increased lymphocyte activation in this pediatric population. While providing valuable insight into the developing pediatric immune system within an African context, the long-term health implications of these findings require further investigation.

**DATA AVAILABILITY STATEMENT**

The datasets generated for this study are available on request to the corresponding author.

**ETHICS STATEMENT**

The studies involving human participants were reviewed and approved by Stellenbosch University granted ethical approval (M12/01/005) and permission for the study was given by Cape Town Department of Health. Written informed consent to participate in this study was provided by the participants’ legal guardian/next of kin.

**AUTHOR CONTRIBUTIONS**

HP and DG conceived the study. HP organized, conducted the CWC, and prepared the manuscript. DL processed and analyzed the specimens. MN and HP performed the statistical analysis. NK, DMG, AB, MC, DG, and Robin Callard, contributed to the study design and data interpretation. All authors have contributed to the writing of the manuscript and approved the final draft for submission.

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### TABLE 3

| Age groups | Overall p-value |
|------------|-----------------|
|            |                 |

Continued...

| CWC | Shearer | CWC | Shearer | CWC | Shearer | CWC | Shearer |
|-----|---------|-----|---------|-----|---------|-----|---------|
| CD8+CD45RO+ cells/ul x 10^3 | CD8+CD45RO+ cells/ul x 10^3 | CD8+CD45RO+ cells/ul x 10^3 | CD8+CD45RO+ cells/ul x 10^3 |
| Naïve/memory CD8 ratio cells/ul | Naïve/memory CD8 ratio cells/ul | Naïve/memory CD8 ratio cells/ul | Naïve/memory CD8 ratio cells/ul |
| Numbers | Numbers | Numbers | Numbers |
| CWC, n | Shearer, n | CWC, n | Shearer, n | CWC, n | Shearer, n | CWC, n | Shearer, n |

(A) Differences seen in B-cells; (B) Activation markers and cytotoxic T-cells; (C) Naïve-memory subsets. Information is displayed with in the context of the reference ranges of each data set. Measured data from the Child Wellness Clinic (CWC) has been organized to match the age groups of the comparison studies. Reference ranges from Shearer et al. (28) are medians with 10-90th centiles, and for Comans-Bitter et al. (29) are medians with 5-95th centiles. Categories with significant differences are highlighted in red. Age-category comparisons were performed using the CHI-squared test, * indicates significance $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$.
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SUPPLEMENTARY MATERIAL
The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fped.2020.00406/full#supplementary-material
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**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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