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Fetal growth trajectories of babies born large-for-gestational-age in the LIFECODES Fetal Growth Study

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Abstract

Introduction: Babies born large-for-gestational age (LGA) have an increased risk of adverse health outcomes, including birth injuries, childhood obesity, and cardiometabolic disorders. However, little work has been done to characterize patterns of fetal growth among LGA births, which may further elucidate high- and low-risk subgroups.

Objectives: Identify subgroups of LGA births based on trajectories of fetal growth derived from prenatal ultrasound measurements, and explore differences in sociodemographic, pregnancy, and birth outcome characteristics across subgroups.

Study Design: We identified and described trajectories of fetal growth among LGA births $(n = 235)$ in the LIFECODES Fetal Growth Study. Ultrasound measurements of fetal growth in mid- to late pregnancy were abstracted from health records. We applied group-based multitrajectory modeling to measurements of head circumference (HC), abdominal circumference (AC), and femur length (FL) z-scores to identify multivariate trajectories of fetal growth. We then summarized sociodemographic variables, pregnancy characteristics, and birth outcomes based on trajectory membership.

Results: We identified four multivariate trajectories of fetal growth among LGA births: Catch-up Growth $(n = 28)$, Proportional AC:FL Growth $(n = 67)$, Disproportional AC:FL Growth $(n = 96)$,

CONFLICTS OF INTEREST:

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and Consistently Large $(n = 44)$. Fetuses in the "Catch-up Growth" group exhibited small relative size in mid-pregnancy (i.e., below average HC, AC, and FL z-scores) and large relative size in late-pregnancy. Growth among these births was driven by increases in relative AC and HC size. Participants who delivered births assigned to this group were less likely to have normal glucose control (40% vs. $65 - 75%$) and more likely to have pre-gestational diabetes (36% vs. $10 - 17%$) compared to other LGA subgroups. The babies in this trajectory group were also more likely to have macrosomia (86% vs. $67 - 73\%$) and to be admitted to the NICU (32% vs. $14 - 21\%$) compared to other LGA subgroups. On the other hand, babies in the "Consistently Large" group had the largest relative size for all growth parameters throughout gestation and experienced lower risk of adverse birth outcomes than other LGA subgroups.

Conclusions: We characterized several trajectories of fetal growth among LGA births, which were related to different pregnancy characteristics and distribution of adverse birth outcomes. While the numbers of individuals within some trajectories was small, we identified a subgroup that exhibited a catch-up growth phenotype during gestation, which may be uniquely associated with exposure to pre-gestational diabetes and a higher risk of admission to the NICU. These results highlight that the risk of adverse outcomes may not be evenly distributed across all LGA births.

Keywords

Large-for-gestational age; group-based trajectory models; latent class trajectory analysis; fetal growth trajectories

INTRODUCTION

Abnormal fetal size is linked to adverse health outcomes, such as neurodevelopmental, cardiovascular, and metabolic disorders.¹⁻⁴ Past work has highlighted the difficulties in identifying babies that are born abnormally small (i.e., growth restricted), $5-7$ including that reliance on percentile cut-offs, such as small-for-gestational age (SGA; < 10 percentilefor-gestational age), does not reliably distinguish between babies with pathological versus constitutional growth profiles.^{7–9} The identification of abnormally large babies, or large-forgestational age (LGA; $> 90th$ percentile weight-for-gestational age), likely suffers from the same shortfalls, but to our knowledge, has received less attention in the peer-reviewed literature.

The incorporation of ultrasound measurements of fetal growth to characterize the overall growth trajectory of a fetus has been suggested as a mechanism to better distinguish pathological deviations in growth from normal variability. Specifically, growth mixture models have been used to characterize fetal growth trajectories, with a particular focus on characterizing small fetuses or isolating subgroups consistent with fetal growth restriction.^{10–13} To our knowledge, these same methods have not been used to characterize growth trajectories among large babies (e.g., LGA) for whom health risks may also differ depending on the trajectory of prenatal growth. These methods assume that identified growth trajectories represent subgroups of individuals with shared risk factors or etiologies¹⁴ and may be more informative in future studies than a single definition of LGA when examining how either exposures or outcomes relate to fetal growth.

In this analysis, we characterized growth trajectories among LGA births using ultrasound measurements of fetal growth collected from mid- to late pregnancy in the LIFECODES Fetal Growth Study. Specifically, we applied group-based multi-trajectory modeling to identify subgroups of individuals with shared profiles of multiple growth indicators, namely, head circumference (HC), abdominal circumference (AC) and femur length (FL) z-scores. We described the distribution of demographic and pregnancy characteristics across trajectory groups and examined how these fetal growth trajectories relate to birth outcomes.

METHODS

Study population

The LIFECODES cohort is an ongoing, prospective pregnancy cohort that began in 2006 at Brigham and Women's Hospital in Boston, MA.15 The study has few eligibility criteria: participants must be at least 18 years of age, initiate prenatal care prior to 15 weeks gestation, and plan on delivering at Brigham and Women's Hospital. As a part of the study, participants attend a total of three study visits occurring at approximately 10-, 26-, and 35 weeks gestation. At the first study visit, participants provide written and informed consent. The LIFECODES study was approved by the Institutional Review Board of Brigham and Women's Hospital and the present analyses were deemed exempt by the National Institute of Environmental Health Sciences.

This analysis utilizes the LIFECODES Fetal Growth Study, which is nested within the LIFECODES cohort and has been described elsewhere.¹² Briefly, this case-cohort study included pregnancies in the LIFECODES cohort that resulted in singleton deliveries between 2008 – 2018 and that had birthweight recorded at delivery. Prior to selection, we determined the birthweight z-score using a standard from the underlying Brigham and Women's Hospital clinic population, which provides a mean and standard deviations of birthweight according to nearest completed weeks gestation.^{16,17} Using the calculated *z*-scores, we identified all SGA (i.e., $< 10th$ percentile-for-gestational age), appropriate-for-gestational age (AGA; i.e., 10^{th} – 90^{th} percentile-for-gestational age), and LGA (i.e., $> 90^{th}$ percentilefor-gestational age) births in the underlying LIFECODES cohort. Selection subsequently occurred in two stages. First, a subcohort of 504 participants was randomly selected from within this time frame. Second, enrichment sets of SGA ($N = 199$) and LGA ($N = 199$) 198) births were randomly selected from identified cases in LIFECODES. This selection mechanism resulted in 249 SGA, 411 AGA, and 241 LGA births. For the purposes of this analysis, we focus on LGA cases and present AGA controls for initial comparisons only.

Ultrasound measurements of fetal growth

We abstracted standard measurements of fetal growth, namely HC, AC, and FL, from ultrasound records. All measurements collected after 15 weeks gestation were converted to gestational age-specific z-scores using an internal growth standard.16,17 As described above, the internal growth standard used in this study provided mean and standard deviations for each ultrasound measurement according to nearest completed gestational week. For a subset of participants, no HC measure was available $(n = 47)$. Instead, we abstracted biparietal diameter (BPD) and substituted their BPD z-score for the HC z-score. As a summary

measure of growth, we also calculated estimated fetal weight (EFW) using Hadlock's Formula #3 based on HC, AC, and FL. For participants missing HC measurements, EFW was calculated using Hadlock's Formula #2, which uses BPD as a measure of head size instead of HC.¹⁸

Participant characteristics and birth outcomes

At the first study visit, participants completed detailed questionnaires to assess demographics (e.g., self-reported race and ethnicity, educational attainment), behavioral characteristics (e.g., smoking and alcohol consumption), and medical history (e.g., previous pregnancies and pregnancy complications). Clinical measurements (e.g., weight and blood pressure) were collected during all study visits. Medical diagnoses occurring during pregnancy were validated by two Maternal-Fetal Medicine specialists. We examined glucose homeostasis on a spectrum with a categorical variable derived from information on glucose test results and information related to gestational, Type-1 or Type-2 diabetes mellitus in the medical record. The categories were as follows: (1) normal glucose metabolism, including participants without gestational or pre-gestational diabetes and normal results (<140 mg/dL) on the standard 1-hour glucose loading test (GLT); (2) mild carbohydrate intolerance, including participants who met or exceeded 140 mg/dL glucose on the GLT, but who did not meet criteria for gestational diabetes on a subsequent 3-hour oral glucose tolerance test and/or did not receive a diagnosis of gestational diabetes in the medical record; (3) gestational diabetes diagnosis in the medical record; and (4) pre-gestational diabetes in the medical record.

Birth outcomes were abstracted from medical records after delivery. The outcomes considered in this analysis included: birthweight z-scores calculated from an internal growth standard¹⁶, birthweight, birthweight-length ratio,¹⁹ gestational age at delivery, delivery via C-section (including scheduled vs. unscheduled), admission to the neonatal intensive care unit (NICU), and length of NICU stay.

Statistical analysis

We summarized demographics, pregnancy characteristics, and birth outcomes of parentchild pairs with either an LGA or AGA birth by calculating the median (interquartile range [IQR]) or n (%). Demographics, pregnancy characteristics, and birth outcomes observed among parent-child pairs selected into the LIFECODES Fetal Growth Study were also compared to those in the underlying LIFECODES cohort from which they were sampled. We displayed the distributions of ultrasound measurements of fetal growth using scatter plots.

We identified the number and shape of growth trajectories among LGA births using groupbased multi-trajectory modeling, as described in Appendix A. Briefly, we optimized the number of multivariate trajectories of HC, AC, and FL z-scores using criteria described by Nagen and colleagues.²⁰ After optimizing the number of trajectories, individual participants were assigned to the trajectory group that corresponded to their highest posterior inclusion probability. We visualized growth trajectories by displaying the predicted mean z-scores (95% CI) for each parameter within each of the identified groups and provided descriptive

names to each trajectory group based on visual inspection of these trajectories rather than the calculation or examination of any additional characteristics.

We summarized the median (IQR) and n (%) of demographics, pregnancy characteristics, and birth outcomes among multivariate trajectory groups. In addition, we examined how the identified trajectory groupings compared to another method of identifying pathologically large fetuses, namely macrosomia (defined as birthweight > 4000 g^{24}).

One limitation of latent class trajectory modeling is that the assignment of group membership contains uncertainty. While individuals were assigned to the group corresponding to their highest posterior inclusion probability, there is some probability of belonging to any of the groups identified. Not accounting for this uncertainty may result in misclassification of the group assignments. To address this, we replicated our descriptive statistics using a case-weight approach, where all results were produced using the posterior inclusion probabilities of group membership as inverse probability weights, rather than assigning fixed trajectory group membership.²⁵

Several previous analyses applying trajectory modeling to longitudinal measurements of fetal growth have analyzed EFW as a summary measure of growth.10–12 In order to compare our results to this literature, we applied univariate group-based trajectory modeling¹⁴ to identify the shape and number of EFW trajectories, and we present these results supplementally. The shape and number of univariate growth trajectories was optimized using the criteria defined above.^{21–23} Demographics, pregnancy characteristics, and birth outcomes were summarized across univariate groups as described previously. In addition, we compared the group assignments for participants between the multivariate and univariate trajectory modeling approaches.

We present this analysis as descriptive in nature and have not included p-values or statistical testing in presentation of our results. Instead, in accordance with recommendations from the American Statistical Association²⁶ and the guidelines for Strengthening Reporting of Observational Studies in Epidemiology $(STROBE)^{27}$, we focus primarily on the direction and magnitude of differences observed between groups under investigation in this work.

RESULTS

Study characteristics

This study comprises 241 parent-child pairs with an LGA birth and 411 with an AGA birth (Table 1). Compared to those with an AGA birth, participants who gave birth to an LGA baby were more likely to have a bachelor's degree or higher education level, to self-identify as non-Hispanic white, have a higher pre-pregnancy BMI, have pre-gestational diabetes, and to give birth to a baby identified as male at birth. They were also less likely to report consuming alcohol or smoking during pregnancy compared to those with an AGA birth. There were similar rates of preeclampsia and gestational diabetes in these two groups. With respect to birth outcomes, LGA babies had similar gestational ages at delivery but were more likely to be delivered via C-section and to be admitted to the NICU at delivery compared to AGA babies. In addition, characteristics observed within our sample of AGA

and LGA births were consistent with those observed among all AGA and LGA births in the underlying LIFECODES cohort (Supplemental Tables 1 and 2).

Fetal growth characteristics

Distributions of ultrasound measurements of fetal growth are shown in Figure 1. In general, babies that were identified as LGA at birth had fetal growth measurements that were larger than those reported for AGA babies. Nevertheless, there was still overlap between measurements in the two groups, especially for FL and HC. There was an average of 4.3 and 4.1 ultrasounds per parent-child pair with an LGA and AGA birth, respectively. A total of 235 parent-child pairs with an LGA birth had at least one HC, AC, and FL z-score and were used in subsequent trajectory modeling.

Growth trajectories among LGA births

Using a multivariate trajectory modeling approach, a 4-group solution was deemed optimal to describe trajectories of fetal growth (Figure 2; Table 2; Supplemental Table 3). The "Catch-up Growth" group ($n = 28$ [12%]) had the smallest relative size in mid-pregnancy with growth driven primarily by increases in relative HC and AC into late pregnancy. There were two groups, termed "Proportional AC:FL Growth" $(n = 67 \, [29\%])$ and "Disproportional AC:FL Growth" ($n = 96$ [41%]), where fetuses were average in size in mid-pregnancy but large in late pregnancy. These groups also had increasing relative AC across gestation, though they differed with respect to the other growth parameters, particularly FL. Relative FL in the "Proportional AC:FL" group increased steadily across pregnancy, as with AC, but FL in the "Disproportional AC:FL" group remained average across gestation. Last, we identified a "Consistently Large" group ($n = 44$ [19%]) with the largest z-scores for all growth measurements throughout gestation.

We observed that participants assigned to the "Catch-up Growth" group were least likely to have normal glucose metabolism and most likely to have pre-gestational diabetes compared to other LGA groups (Table 3; Figure 3). The "Disproportional AC:FL Growth" group was characterized by older participants, the highest proportion of participants with a bachelor's degree or greater, and were more likely to self-identifiy as non-Hispanic white, have private health insurance, and to have used assisted reproductive technologies to conceive relative to the other LGA groups. The "Proportional AC:FL Growth" group had higher pre-pregnancy BMI relative to participants in the other LGA groupings. The "Consistently Large" group were most likely to have normal glucose metabolism relative to the other LGA groups.

With respect to birth outcomes, the "Catch-up Growth" group had the highest proportion of babies admitted to the NICU relative to other LGA groups (Figure 4; Table 3). However, the distributions of other birth outcome measurements overlapped substantially with those observed among the other groups. Conclusions were consistent using a caseweight approach rather than assigning fixed trajectory group membership (Supplemental Table 4).

Some differences in the distribution of macrosomia diagnoses by multivariate trajectory group membership were observed (Supplemental Table 5). Notably, the "Catch-up Growth"

group had the highest proportion of infants with macrosomia (86%) compared to the other LGA groups (67–73%).

In sensitivity analyses, we identified three univariate EFW trajectory groups among babies born LGA (Supplemental Figure 1). There were some similarities between the univariate and multivariate trajectories. For example, we identified a group of "Consistently Large" babies, with the largest EFW z-scores throughout gestation, and a "Catch-up Growth" group where EFW was relatively small (i.e., below the 10th percentile) in mid-pregnancy and large during late pregnancy. There were also similarities with respect to trends in sociodemographic variables and birth outcomes observed in the identified trajectory groups (Supplemental Table 6). For example, participants in the "Catch-up Growth" group had the highest proportion of NICU admissions relative to other trajectory groups. There was concordance between parent-child pairs assigned to the "Consistently Large" group in both the univariate and multivariate approach (Supplemental Table 7). Similarly, the univariate "Catch-up Growth" group was inclusive of nearly all individuals in the multivariate "Catchup Growth" group.

COMMENT

Principal findings

In this study, we identified and described phenotypes of fetal growth among LGA births in the LIFECODES Fetal Growth Study. Using repeated ultrasound measurements collected from mid- to late pregnancy, we identified four multivariate trajectories based on shared profiles of HC, AC, and FL z-scores. These trajectory groups differed in their associations with birth outcomes, including NICU admission. Specifically, fetuses exhibiting the largest relative increase in AC and HC across pregnancy, which we defined as a prenatal "Catch-up Growth" group, had the highest proportion of births admitted to the NICU.

Results in the context of what is known

To our knowledge, no other studies have characterized growth phenotypes among LGA births using repeated ultrasound measurements of growth across gestation. However, we have previously conducted similar analyses among SGA infants. In these studies, we reported 3 univariate trajectories based on EFW and 4 multivariate growth trajectories based on measurements of HC, AC, and FL z -scores.^{10,12} In this context, we observed that babies with the smallest size across gestation had higher risk of adverse birth or childhood neurodevelopmental outcomes relative to those with transient growth profiles. This is contrary to our current findings among LGA births, where we have found that fetuses with transient growth (e.g., "Catch-up Growth") had the highest proportion of infants admitted to the NICU compared to other groups. Specifically, the "Catch-up Growth" trajectory group was characterized by smaller relative size during mid-pregnancy and large increases in relative HC and AC. A key characteristic of the "Catch-up Growth" group was that it contained the highest proportion of individuals with pre-gestational diabetes (primarily Type 1 diabetes) and the lowest proportion of individuals with normal glucose control. Previous studies examining growth profiles among pregnant persons with diabetes that have described fetal growth trajectories featuring both mid-pregnancy growth delays and rapid increases in

AC from mid- to late-pregnancy.^{28–32} Others have hypothesized that mid-pregnancy growth delay among pregnant persons with diabetes could be due to the influence of poor glycemic control on placentation in early pregnancy.33 On the other hand, increased growth rate in late pregnancy may be due to fetal hyperinsulinemia triggered by maternal hyperglycemia, which may result in greater adipose development. $33,34$ Together, these findings suggest that this growth phenotype could be a characteristic of prenatal exposure to diabetes.

Clinical implications

Latent class methods have shown promise in their ability to identify hidden subgroups of diseased individuals with shared clinical profiles or etiologies, including in the analysis of fetal growth.8,35 While these methods have been primarily applied to growth at the small end of the birthweight spectrum, we have demonstrated that it may also distinguish between clinically meaningful subgroups of LGA babies. Further research is warranted to continue interrogation of fetal growth trajectories among LGA babies.

Research implications

Latent class trajectory modeling is used to derive relatively homogenous subgroups of individuals hypothesized to share etiology or mechanisms of disordered growth.14 From this perspective, these subgroups may be more meaningful in the context of future epidemiologic studies seeking to examine how risk factors or outcomes relate to fetal growth. Thus, this characterization of LGA phenotypes may contribute to further understanding the determinants and consequences of abnormally large babies.

Strengths and limitations

This analysis was not without limitations. First, this study population was recruited from a high-risk tertiary care facility. While the number of ultrasounds collected on these participants was relatively large, their measurements may not reflect "normal" growth among LGA births within the general population, representing possible selection bias. Thus, the shape and distribution of growth phenotypes may not generalize to other populations. However, the enrichment for high-risk pregnancies may improve our ability to characterize pathological growth phenotypes, such as those characterized by individuals with diabetes. Second, although we standardized our growth measures for gestational age using an internal standard, it only provided information to characterize growth according to the nearest completed week and did not consider individual gestational days.16,17 This may result in some misclassification, particularly for measures occurring at either the beginning or the end of a gestational week. Third, we limited our analysis to participants with LGA births. One consequence of this approach is that we were not able to capture fetuses that may still have exceeded their true growth potential, but who were not classified as LGA at the time of delivery. Yet, we expect that a large proportion of babies experiencing pathological overgrowth would be captured among LGA births. Fourth, we had a limited number of birth and childhood follow-up outcomes that could be included in this analysis. Last, while this study population had a large number of LGA births, we observed small sizes within subgroups. Thus, further research is needed to replicate our findings with larger sample sizes.

BOMMARITO et al. Page 9

Nevertheless, this study had several strengths. To our knowledge, this is the first analysis to characterize phenotypes of fetal growth among LGA births using longitudinal ultrasound measurements. Given the large number of ultrasounds in this population (on average > 4 per participant), we were able to identify multiple trajectory groupings in our approach. In addition, we were able to apply group-based multi-trajectory modeling to examine profiles of growth according to HC, AC, and FL measurements. Relying solely on EFW as a summary measure of growth has been shown to have several issues $36-38$ and we expect the ability to consider fetal body composition to provide a more comprehensive analysis of growth.

Conclusions

We identified phenotypes of fetal growth among LGA births in the LIFECODES Fetal Growth Study, which differed in their profiles of demographics, pregnancy characteristics, and adverse health outcomes. Specifically, we observed that a small subgroup of babies following a prenatal "Catch-up Growth" phenotype during gestation were characterized by potentially higher risk of adverse birth outcomes, namely NICU admission, than other babies born LGA. These findings suggest that considering the overall trajectory of fetal growth may provide important information about health risks compared to relying on birthweight alone. Given the exploratory nature of this study, further work is needed to replicate these findings.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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AJOG AT A GLANCE:

Why was this study conducted?

This exploratory study sought to characterize fetal growth profiles of babies born largefor-gestational age using latent class multi-trajectory modeling applied to ultrasound measurements of fetal growth from mid- to late pregnancy

What are the key findings?

We observed that the subgroup of fetuses with small relative size in mid-pregnancy and rapid growth into late pregnancy had the highest risk of admission to the neonatal intensive care unit at the time of delivery relative to other subgroups of large-forgestational age births. In addition, this subgroup was characterized by the greatest exposure to pre-gestational diabetes relative to others.

What does this add to what is known?

To our knowledge, this is the first study to characterize multivariate fetal growth trajectories among large-for-gestational age births using latent class trajectory modeling approaches.

CONDENSATION:

This study identifies subgroups of fetal growth using biometry measurements abstracted from repeated ultrasounds taken from mid- to late pregnancy among large-for-gestational age births.

Figure 1.

Distribution of ultrasound measures of fetal growth by case status in the LIFECODES Fetal Growth Study. Measures of fetal growth abstracted from ultrasounds (i.e., estimated fetal weight, head circumference, abdominal circumference, and femur length) are shown according to case status. AGA = appropriate-for-gestational age, SGA = small-forgestational age.

Figure 2.

Predicted mean HC, AC, and FL z-scores (95% CI) of multivariate trajectory groups identified among babies born LGA in the LIFECODES fetal growth case-cohort. Red dashed lines indicate lines representing LGA (z-score = 1.28) and SGA (z-score = −1.28) cut points. Black dashed line indicates average growth (z -score = 0).

BOMMARITO et al. Page 16

Figure 3.

Categories of glucose control for multivariate fetal growth trajectory groups among parentchild pairs with an LGA birth in the LIFECODES Fetal Growth Study. Data underlying this figure can be found in Table 3.

Figure 4.

Proportion or median (IQR) of select birth outcomes in univariate trajectory groups among LGA births in the LIFECODES Fetal Growth Study. NICU = neonatal intensive care unit. Data underlying this figure can be found in Table 3.

Table 1.

Demographics, pregnancy characteristics, and birth outcomes among parent-child pairs with an LGA or AGA birth in the LIFECODES Fetal Growth Study

Abbreviations: AGA = appropriate-for-gestational age, ART = assisted reproductive technologies, BMI = body mass index, LGA = large-forgestational age, NICU = neonatal intensive care unit

Notes: $n = 9$ missing education, $n = 8$ missing health insurance, $n = 9$ missing pre-pregnancy BMI, $n = 6$ missing alcohol consumption, $n = 43$ missing measures of glucose control, n = 36 missing birthweight-length ratio, n = 5 missing delivery via C-section, n = 6 missing unscheduled C-section; $n = 8$ missing NICU admission. Other race and ethnicity includes Asian (n = 31), More than one race (n = 20) and Other (n = 12). Public insurance category includes those using Medicaid/Mass Health/SSI ($n = 162$), Self pay ($n = 12$) and no health insurance ($n = 1$). Type of pre-gestational diabetes only displayed for individuals diagnosed with pre-gestational diabetes (n = 47). Unscheduled C-section only displayed among individuals with a C-section delivery $(n = 291)$. Length of NICU stay only calculated among individuals with a known NICU admission $(n = 100)$ 92).

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Table 2.

Descriptions of fetal growth phenotypes identified by univariate and multivariate latent class trajectory models in the LIFECODES Fetal Growth Study. Descriptions of fetal growth phenotypes identified by univariate and multivariate latent class trajectory models in the LIFECODES Fetal Growth Study.

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Table 3.

Demographics, pregnancy characteristics, and birth outcomes for multivariate fetal growth trajectory groups among parent-child pairs with an LGA birth Demographics, pregnancy characteristics, and birth outcomes for multivariate fetal growth trajectory groups among parent-child pairs with an LGA birth in the LIFECODES Fetal Growth Study in the LIFECODES Fetal Growth Study

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for individuals diagnosed with pregestational diabetes (n = 36). Unscheduled C-section only displayed among individuals with a C-section delivery (n = 131). Length of NICU stay only calculated among for individuals diagnosed with pregestational diabetes (n = 36). Unscheduled C-section only displayed among individuals with a C-section delivery (n = 131). Length of NICU stay only calculated among Note: $n = 4$ missing education, $n = 2$ missing health insurance, $n = 2$ missing pre-pregnancy BMI, $n = 1$ missing alcohol consumption during pregnancy, $n = 14$ missing measures of glucose control, $n = 1$ Note: $n = 4$ missing education, $n = 2$ missing health insurance, $n = 2$ missing pre-pregnancy BMI, $n = 1$ missing pregnancy, n $n = 14$ missing measures of glucose control, $n = 14$ 5), More than one race $(n = 5)$ and Other $(n = 2)$. Public insurance category includes those using Medicaid/Mass Health/SSI $(n = 61)$ and Self pay $(n = 1)$. Type of pregestational diabetes only displayed 5), More than one race (n = 5) and Other (n = 2). Public insurance category includes those using Medicaid/Mass Health/SSI (n = 61) and Self pay (n = 1). Type of pregestational diabetes only displayed 8 missing birthweight-length ratio, $n = 1$ missing delivery via C-section, $n = 5$ missing delivery via unscheduled C-section, $n = 5$ missing NICU admission. Other race and ethnicity includes Asian ($n =$ 8 missing birthweight-length ratio, n = 1 missing delivery via C-section, n = 5 missing delivery via unscheduled C-section, n = 5 missing NICU admission. Other race and ethnicity includes Asian (n = individuals with a known NICU admission ($n = 46$). individuals with a known NICU admission $(n = 46)$.