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Bi-Directional Associations of Affective States and Diet among Low-Income Hispanic Pregnant Women using Ecological Momentary Assessment

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Abstract: Background: Affective states play a role in dietary behaviors. Yet, little research has studied within-subjects associations between affect and diet during pregnancy. We examined the acute bidirectional relationships between affect and food intake and moderation by pre-pregnancy body mass index (BMI) in low-income, Hispanic pregnant women using ecological momentary assessment (EMA). Methods: Women (N=57) completed four days of EMA during their first trimester. Women responded to five random prompts per day about their current affect and past two-hour food intake. Results: Higher positive affect (PA) or lower negative affect (NA) predicted greater likelihood of fruit/vegetable consumption in the next two hours in women with lower pre-pregnancy BMI and lower likelihood in women with higher pre-pregnancy BMI. Higher PA predicted less likelihood of fast food consumption in the next two hours in women with lower pre-pregnancy BMI had higher likelihood in women with higher pre-pregnancy BMI. Women with lower pre-pregnancy BMI had higher PA when they reported consuming chips/fries in the past two hours, and women with higher pre-pregnancy BMI had lower PA when they reported consumption of chips/fries in the past two hours. Conclusions: Results showed differential relationships between affect and food intake as a function of pre-pregnancy BMI.

Keywords: affect; diet; pregnancy; food intake; body mass index; ecological momentary assessment

1. Introduction

Excessive gestational weight gain (GWG) is associated with adverse short- and long-term maternal and child outcomes such as hypertensive disorders of pregnancy and obstetric interventions (Truong et al., 2015), and maternal and child obesity, cardiovascular disease risk, and diabetes (Fraser et al., 2010; Hedderson et al., 2010; Houghton et al., 2016). Low income, Hispanic women are at particular risk for excessive GWG during pregnancy (Headen et al., 2012; Rosal et al., 2016), which underscores the need for prevention efforts targeting unhealthy GWG in this population. A promising approach to reducing excessive GWG in pregnancy includes improving diet (Rogozinska et al., 2017). However, psychoeducational interventions that seek to modify dietary behaviors to prevent excessive GWG have produced mixed results (Gardner et al., 2011), and they do not specifically target low income, Hispanic women (Chasan-Taber, 2012). Therefore, greater understanding of determinants of low income, Hispanic women's dietary intake choices during pregnancy is needed to develop tailored prevention and interventions.

A growing body of research suggests that emotional factors may play a role in dietary intake behaviors. Specifically, in a qualitative study of Hispanic women, negative emotions were identified as a barrier to eating healthy (Fletcher et al., 2018). It is possible that

negative emotions impair ability to prepare healthy meals or may increase desire to consume comfort foods. In general, research suggests that positive affect promotes healthy food intake (e.g., consumption of fruits and vegetables) because it may bolster individuals' ability to make healthier decisions (Isen, 2001). Also, individuals may choose healthy foods to maintain positive affective states, as consumption of healthy foods may lead to feelings of joy or pride (Wansink et al., 2003). Furthermore, consumption of healthy foods may predict more positive affect as individuals may feel good about engaging in healthy behaviors (Wahl et al., 2017). There is also evidence that both positive and negative affect can predict unhealthy dietary intake. The affect regulation model suggests that individuals consume unhealthy foods, often in large quantities, to cope with feelings of negative affect, such as anger or sadness (Polivy & Herman, 1993). Although, other research has shown positive affect to be a predictor of increased unhealthy eating behaviors, particularly in subgroups at-risk for unhealthy eating (e.g., individuals high in emotional eating or with low self-control; Bongers et al., 2013; Smith et al. 2018). Because pregnancy is a unique time where emotions may shift more than usual due to hormones, stress, and body changes (Bjelica et al., 2018) and women experience more hunger and cravings (Groth et al., 2021), it may be particularly relevant to study emotions as antecedents and consequences of dietary behaviors in pregnant women. However, research in this area is scant.

It has been recognized that dietary intake varies both between and within individu-A burgeoning array of studies have used ecological momentary assessment (EMA) methods to examine positive and negative affective states as antecedents and consequences of dietary behaviors. EMA collects repeated measurements of experiences and behaviors across the day for a short duration (e.g., a week) in individuals' natural environment (Shiffman et al., 2008). Thereby EMA methods allow for the examination of how behaviors at one moment in time predict subsequent states. Additional strengths of EMA include minimizing recall biases and maximizing ecological validity. EMA studies have found positive bi-directional within-subject associations between positive affect and fruit/vegetable intake (Liao et al., 2019; Mason et al., 2019). In addition, college students with greater positive affect across EMA reported more fruit consumption (Jeffers et al., 2019). Also, among college students, reporting of at least one positive affective state was concurrently related to greater likelihood intake of sweets and less likelihood of intake of fast food, but there were no relations with negative affect (Ashurst et al., 2018). In a separate study, within-subject differences in affective states did not predict subsequent sugary food intake, yet sugary food intake was associated with increased negative affect during the same 2-hr window (Jeffers et al., 2019). Further, between-subjects differences in negative affect were related to sugary food and sweet consumption, such that those with greater negative affect consumed more sugary foods and sweets (Jeffers et al., 2019; Mason et al., 2019). The extent to which these study findings on affect and food intake extend to low-income, Hispanic pregnant women has not been tested.

While differences in the type of study sample may to some degree explain mixed findings, it is also possible that associations between affect and diet may differ based on previously unexplored moderators such as weight status (Geliebter & Aversa, 2003; Reichenberger et al., 2018). Body mass index (BMI) has been shown to be positively associated with the tendency to eat in response to negative emotions and inversely associated with the tendency to eat in response to positive emotions (Nolan et al., 2010). Further, longitudinal research has found that emotional eating is associated with increases in BMI (Shriver et al., 2019). In addition, studies have found positive associations between BMI and reward responsivity to food suggesting that higher weight individuals may derive more pleasure from food (Luo et al., 2013; Price, Higgs, & Lee, 2015); this could manifest as increased positive affect and lower negative affect following eating unhealthy food. Finally, BMI is positively associated with body dissatisfaction (Fitzgibbon et al., 2000; Mason & Lewis, 2015), and thus, women with higher weights may feel more negative affect after eating unhealthy foods or more positive affect after healthy foods given their possible effect on weight gain.

The current analyses examined acute within-subject bidirectional relationships between affective states and dietary intake in low-income Hispanic pregnant women using EMA. It was hypothesized that greater than usual positive affect would be associated with increased likelihood of fruit/vegetable intake in the next two hours, and some fruit/vegetable intake in the previous two hours would predict greater positive affect at the end of that time interval. Also, it was hypothesized that greater than usual negative affect would be associated with increased likelihood of no sweets, fried foods, and fast food intake in the next two hours, and intake of these foods in the past two hours would predict greater negative affect at the end of that time interval. Pre-pregnancy BMI was explored as a moderator of these associations. Fruits/vegetables, sweets, chips/fries, and fast food were selected at the target dietary intake behaviors given their contributions to weight gain and caloric intake (Guallar-Castillón et al., 2007; Nour et al., 2018; Rosenheck, 2008; Te Morenga et al., 2013).

2. Methods

2.1. Participants and Procedure

A subsample of low income, Hispanic pregnant women from the Maternal and Developmental Risks from Environmental and Social Stressors (MADRES) cohort completed four days of EMA during their first trimester. They responded to five random EMA prompts per day about their current affective states and past two-hour food intake. Women were recruited from several designated healthcare facilities (e.g., community clinic, academic medical center) in urban Los Angeles. Inclusion criteria for the larger cohort were: <30 weeks since the date of mother's last menstrual period at the time of enrollment; 18 years of age or older; and singleton pregnancy. Exclusion criteria were HIV positive status; physical, mental, or cognitive disabilities that prevent participation; and current incarceration. For the EMA study, women had to be Hispanic and <20 weeks gestation at enrollment. Sixty-five women completed the consent process and enrolled in the study, but only 59 women completed EMA at first trimester. Finally, two participants were removed because did not report affect data. As a result, the analytical sample size was 57.

Participants were provided an Android smartphone loaded with the MovisensXS application for each four-day EMA burst, which always included two weekend days. EMA surveys were prompted five times per day at a random time within during pre-set sampling windows (i.e., wake-up–10 am; 11 am–1 pm; 2 pm–4 pm; 5 pm–7 pm, and 8 pm–bed-time). Sleep and wake times were customized for each participant. Each EMA survey took approximately 2-3 minutes to complete. Participants had the option to delay surveys if they were prompted during incompatible times. There were up to two reminder prompts sent within 10 minutes of the initial prompt, after which time the EMA survey was closed. Each survey question was presented on a separate screen. EMA data from smartphones was wirelessly uploaded after each entry and stored on a cloud server or pushed to the server by staff after data collection, where it was viewed by research staff to monitor compliance.

2.2. Measures

Affect. Negative affect was measured with three items: "Right before the phone went off, how (1) angry, (2) sad/depressed, and (3) stressed were you feeling?", and positive affect was measured with two items: "Right before the phone went off, how (1) happy and (2) calm/relaxed were you feeling?" Response options consisted of a scale ranging from 1 (not at all) to 4 (extremely). Items were combined to create a negative affect and positive affect summary score, which was created by averaging the responses together. In addition, Responses to the two EMA negative affect and positive affect items were internally consistent (Coefficient Omega (ω s) > 0.85). This affect measure has been used in previous EMA research (Mason et al., 2019).

Food intake. Women reported their recent dietary intake through the following EMA question: "Over the last 2 hours, which of these things have you done?" They indicated whether they had consumed the following: (a) pastries, sweets, or pan dulce, (b) chips or fries, (c) fast food, and (d) fruits or vegetables. Participants were instructed to select all that applied. This dietary intake checklist has been used in previous EMA research (Mason et al., 2019) and has shown concordance with 24hr recalls of dietary intake (O'Connor et al., 2018).

Pre-pregnancy BMI. Women self-reported their pre-pregnancy weight and height, which was used to calculate pre-pregnancy BMI (kg/m²). Continuous BMI was used.

Demographic and participant characteristics. Recruitment site, maternal acculturation (foreign-born versus not), participant age at study entry, highest attained education level. Participants were asked "Right before the phone went off, how NAUSEOUS were you feeling?" on a 4-point scale from 1 (not at all) to 4 (extremely); these scores were averaged for each day to create a daily nausea score Lastly, the EMA program recorded whether each day was a weekday or a weekend day.

2.3. Statistical Analysis

Multilevel models (MLMs) tested the bi-directional within- and between-subjects associations (1) between affective states at one prompt and past two-hour dietary intake reported at the next prompt and (2) between past two-hour dietary intake and current affective state both reported at the same prompt. MLMs estimate models that are nested within-persons and can handle time-varying data (e.g., within-subjects variance at prompt; Garson, 2013).

The first set of MLMs used binary logistic regression to test affective states as predictors of dietary intake. Affect variables were disaggregated into between-subjects (i.e., aggregate level of positive or negative affect across EMA) and within-subjects (i.e., deviation from one's own average positive or negative affect level at an individual prompt) effects, and within-subjects effects were time-lagged (prompt-1). Models also included pre-pregnancy BMI, and the cross-level interaction between within-subjects time-lagged affect and pre-pregnancy BMI were included as predictors of intake of each dietary outcome within the past two hours.

The second set of MLMs examined past two-hour dietary intake as predictors of current affect. Models included within-subjects dietary intake, between-subjects dietary intake, BMI, and the cross-level interaction between within-subjects dietary intake and prepregnancy BMI as predictors of positive and negative affect. Due to non-normality, a zero-truncated Poisson function was applied to the negative affect model. For both sets of models, positive and negative affect and each dietary behavior were tested in different models. All models included the following covariates: maternal age, education level, language preference, recruitment site, EMA nausea (disaggregated into between- and within-subject nausea scores), and weekday vs. weekend day.

3. Results

Participant demographic characteristics are displayed in Table 1. Across all participants, there were 1,086 EMA prompts received, of which 914 were responded to for an overall compliance rate of 84% (range from 33% to 100%). Women reported consuming fruit or vegetables in 22.9% of prompts; chips or fries in 4.7% of prompts; pastries or sweets in 8.3% of prompts; and fast food in 6.4% of prompts. EMA affect items had within-subjects internal consistency reliabilities (ω s) of .90 (positive affect) and .86 (negative affect), and between-subjects ω s were .97 (positive affect) and .89 (negative affect). Mean level of negative affect across prompts was 1.24 (SD=0.40) and positive affect was 2.71 (SD=0.74). The mean level of daily nausea was 1.51 (SD=0.79).

Table 1. Demographic Descriptive Statistics.

	Percent	Mean	Std	Min	Max
Pregnancy BMI		29.57	6.73	19.07	51.97
Age		28.77	6.09	18.31	45.42
Education					
Less than 12th grade (did not finish high school)	35.6%				
Completed grade 12 (high school)	28.8%				
Some college or technical school	23.7%				
Completed four years of college	8.5%				
Some graduate training after college	3.4%				
Language Preference					
English	54.2%				
Spanish	45.8%				
Citizenship					
US-Born	44.1%				
Foreign-Born	55.9%				

3.1. Affect Predicting Dietary Intake

Results of models testing affect as a predictor of dietary intake are presented in Table 2. Mothers with higher education or who were foreign born reported greater fruit/vegetable intake, and chips/fries were more likely to be consumed on weekends. There were interactions between pre-pregnancy BMI and positive and negative affect predicting fruit/vegetable intake. When women with lower pre-pregnancy BMI had higher positive affect or lower negative affect than their average, they were subsequently more likely to consume fruits/vegetables in the next two hours. Oppositely, when women with higher pre-pregnancy BMI had higher positive affect or lower negative affect than their average, they were subsequently less likely to consume fruits/vegetables in the next two hours (see Figure 1 and 2). For fast food, there was an interaction between pre-pregnancy BMI and within-subjects positive affect (Figure 3). When women with lower pre-pregnancy BMI had higher positive affect than their average, they were subsequently less likely to consume fast food in the next two hours. Higher positive affect slightly increased the likelihood of fast food consumption in the next two hours for women with higher pre-pregnancy BMI, but in general, women with higher pre-pregnancy BMI had lower levels of fast food consumption. There were no associations between affect and consumption of sweets/pastries or chips/fries.

Table 2. Results of Multilevel Models Testing Affect Predicting Subsequent Food intake.

Regressor / Outcome (1=intake)	Fruits/vegetables		Sweets/pastries		Chips/fried foods		Fast food	
	OR	p	OR	p	OR	p	OR	p
Positive affect (PA)								
Intercept	0.14	0.124	0.64	0.734	0.19	0.301	0.72	0.831
<u>Level 1</u> $(n = 589)$								
$\overline{\text{Weekend}}$ (Weekend = 1)	0.99	0.958	1.46	0.197	2.10	0.041	1.16	0.613
Within-subjects PA (Lag1)	10.94	0.010	8.05	0.121	0.91	0.950	0.06	0.020
Within-subjects nauseous (Lag1)	1.19	0.330	0.78	0.327	1.13	0.682	1.00	0.995
Level 2 $(n = 57)$								
Between-subjects PA	0.81	0.523	0.52	0.073	1.33	0.526	1.35	0.486
Between-subjects nauseous	0.86	0.631	1.71	0.072	1.18	0.685	1.12	0.774
BMI	0.98	0.363	0.94	0.050	0.96	0.348	0.98	0.614
Maternal baseline age (Unit: 10yrs)	1.19	0.553	1.01	0.966	0.77	0.585	0.79	0.571
Maternal education level	1.58	0.016	0.96	0.848	1.08	0.765	0.71	0.177
Site of data collection	1.09	0.837	0.50	0.095	1.71	0.367	0.96	0.937
Maternal acculturation (Foreign-Born)	2.36	0.029	1.49	0.308	0.48	0.158	0.53	0.206
<u>Interaction</u>								

BMI x WS PA (Lag1)	0.92	0.010	0.95	0.208	1.02	0.737	1.09	0.037
Negative affect								
Intercept	0.14	0.133	1.02	0.988	0.14	0.229	0.76	0.860
$\underline{\text{Level 1}} \ (n = 589)$								
Weekend (Weekend $= 1$)	1.01	0.956	1.41	0.234	2.06	0.045	1.15	0.643
Within-subjects NA (Lag1)	0.07	0.032	0.08	0.268	10.15	0.291	17.52	0.172
Within-subjects nauseous (Lag1)	1.20	0.297	0.70	0.157	1.08	0.800	1.03	0.923
$\underline{\text{Level 2}} \ (n = 57)$								
Between-subjects NA	0.96	0.965	0.56	0.548	1.59	0.717	0.59	0.642
Between-subjects nauseous	0.91	0.763	2.18	0.011	1.02	0.966	1.07	0.851
BMI	0.98	0.412	0.94	0.047	0.97	0.381	0.98	0.567
Maternal baseline age	1.17	0.593	1.00	0.997	0.79	0.605	0.81	0.602
Maternal education level	1.55	0.023	0.89	0.579	1.11	0.665	0.72	0.178
Site of data collection	1.06	0.885	0.44	0.065	1.89	0.291	0.93	0.896
Maternal acculturation (Foreign-Born)	2.37	0.027	1.40	0.400	0.48	0.148	0.51	0.169
<u>Interaction</u>								
BMI x WS NA (Lag1)	1.09	0.033	1.06	0.498	0.93	0.342	0.91	0.192

Note. OR=Odds Ratio; PA=positive affect; BMI=body mass index; NA=negative affect.

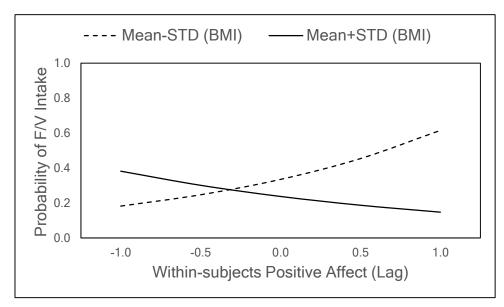


Figure 1. Two-way interactions of lagged within-subjects positive affect and pre-pregnancy BMI predicting intake of fruits and vegetables.

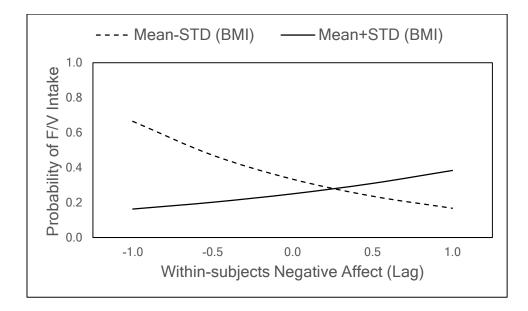


Figure **2.** Two-way interactions of lagged within-subjects negative affect and pre-pregnancy BMI predicting intake of fruits and vegetables.

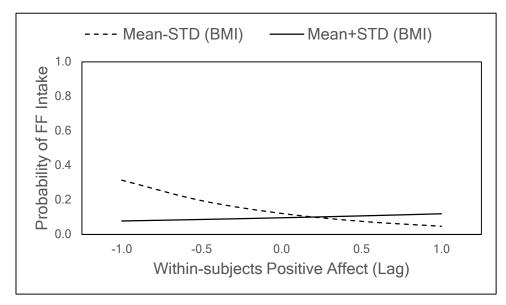


Figure 3. Two-way interaction of lagged within-subjects positive affect and pre-pregnancy BMI predicting fast food intake.

3.2. Dietary Intake Predicting Affect

Results of models of dietary intake predicting affect are displayed in Table 3. Mothers reporting more nausea (both WS and BS nausea effects) had lower positive affect and higher negative affect. Mothers with higher education reported lower negative affect, and positive affect was lower on weekend days. There was a main affect for the association of within-subjects sweets/pastry intake and subsequent negative affect, such that intake of sweets/pastries during a two-hour interval was associated with lower negative affect reported at the end of that interval. Pre-pregnancy BMI interacted with within-subjects chips/fries intake in predicting positive affect. Shown in Figure 4, women with lower pre-pregnancy BMI had higher positive affect when they reported consuming chips or fries in the past two hours compared to when they did not. Conversely, women with higher pre-pregnancy BMI had lower positive affect when they reported consumption of chips or fries in the past two hours.

 Table 3. Results of Multilevel Models of Food Intake Predicting Affect.

		come e Affect	Outcome Negative Affect	
	Beta	p	Beta	p p
Fruits/vegetables	Вети	P	Вен	P
Intercept	2.282	<.0001	1.990	0.106
Level 1 $(n = 589)$			2.550	0.100
Weekend (Weekend = 1)	-0.117	0.010	0.283	0.140
Intake (Yes = 1)	0.200	0.461	-1.002	0.378
Within-subjects nauseous	-0.234	<.0001	0.632	0.001
Level 2 $(n = 57)$	-0.234	<.0001	0.032	0.001
Between-subjects nauseous	-0.269	0.031	0.686	0.015
BMI	0.008	0.508	-0.046	0.013
Maternal baseline age	0.049	0.693	0.278	0.341
Maternal education level	0.062	0.435	-0.453	0.014
Site of data collection	0.064	0.710	-1.221	0.003
Maternal acculturation (Foreign-Born)	-0.138	0.407	-0.374	0.326
<u>Interaction</u>				
BMI x Intake	-0.005	0.593	0.028	0.478
Sweets/pastries				
Intercept	2.286	<.0001	2.223	0.067
$\underline{\text{Level 1}} \ (n = 589)$				
Weekend (Weekend $= 1$)	-0.116	0.011	0.268	0.166
Intake $(Yes = 1)$	0.227	0.606	-3.697	0.045
Within-subjects nauseous	-0.230	<.0001	0.607	0.001
Level 2 $(n = 57)$				
Between-subjects nauseous	-0.275	0.028	0.737	0.009
BMI	0.007	0.547	-0.047	0.070
Maternal baseline age	0.056	0.651	0.223	0.439
Maternal education level	0.066	0.405	-0.464	0.010
Site of data collection	0.065	0.707	-1.250	0.002
Maternal acculturation (Foreign-Born)	-0.132	0.428	-0.384	0.306
Interaction	-0.132	0.426	-0.364	0.500
BMI x Intake	-0.007	0.668	0.115	0.078
Chips/fried foods	-0.007	0.000	0.113	0.070
Intercept	2.209	<.0001	1.855	0.123
Level 1 $(n = 589)$	2.209	<.0001	1.655	0.123
	0.100	0.017	0.277	0.150
Weekend (Weekend = 1)	-0.108	0.017	0.277	0.150
Intake (Yes $= 1$)	1.152	0.007	-0.298	0.865
Within-subjects nauseous	-0.232	<.0001	0.621	0.001
$\underline{\text{Level 2}} \ (n = 57)$				
Between-subjects nauseous	-0.265	0.034	0.695	0.012
BMI	0.008	0.452	-0.039	0.129
Maternal baseline age	0.058	0.640	0.248	0.385
Maternal education level	0.073	0.359	-0.465	0.010
Site of data collection	0.067	0.700	-1.208	0.002
Maternal acculturation (Foreign-Born)	-0.122	0.463	-0.410	0.272
Interaction				
BMI x Intake	-0.040	0.007	0.008	0.895
Fast food				
Intercept	2.303	<.0001	1.798	0.135
Level 1 $(n = 589)$	2.505		1.770	0.133
Weekend (Weekend = 1)	-0.116	0.011	0.274	0.153
Intake (Yes = 1)	-0.101	0.784	0.373	0.814
Within-subjects nauseous	-0.233	<.0001	0.621	0.001
T and 1 4 (EP)				
$\underline{\text{Level 2}} \ (n = 57)$	0.272	0.000	0.60=	0.045
Between-subjects nauseous	-0.272	0.030	0.697	
	-0.272 0.006 0.055	0.030 0.604 0.653	0.697 -0.037 0.250	0.012 0.148 0.382

Maternal education level Site of data collection	0.069 0.064	0.388 0.710	-0.464 -1.212	0.010 0.002
Maternal acculturation (Foreign-Born)	-0.128	0.442	-0.402	0.283
Interaction				
BMI x Intake	0.007	0.542	-0.015	0.776

Note. PA=positive affect; BMI=body mass index; NA=negative affect.

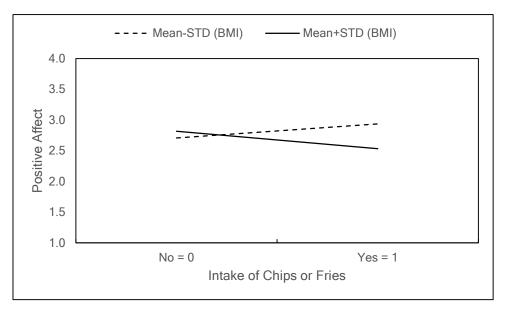


Figure 4. Two-way interaction of within-subjects chips or fries intake and pre-pregnancy BMI predicting positive affect.

4. Discussion

This study investigated bi-directional momentary associations between affective states and dietary intake in low-income Hispanic pregnant women, and whether associations differed by pre-pregnancy BMI. To our knowledge, this is the first study to examine affect and food intake using EMA during pregnancy. Hypotheses for main effects were generally not supported as there were no within-subjects effects of affect predicting subsequent dietary intake and only one main effect for intake predicting affect. However, exploratory moderation analyses revealed several interactions with pre-pregnancy BMI.

For fruit and vegetable intake, women with lower pre-pregnancy BMI followed patterns consistent with theory and empirical research (Isen, 2001; Liao et al., 2019) such that higher positive affect was associated with greater fruit and vegetable intake. In addition, higher negative affect was associated with a lower likelihood of fruit and vegetable intake in the next two hours. Most research has found no relation between negative affect and fruit and vegetable intake (e.g., Jeffers et al., 2020; Mason et al., 2019), but it appears during pregnancy negative affect may have a more relevant role. Oppositely, when women with higher pre-pregnancy BMI had lower positive affect and higher negative affect than their average, they were subsequently more likely to consume fruits/vegetables in the next two hours. It is possible that women with higher pre-pregnancy BMI consume food for comfort or affect regulation but choose fruit and vegetables given the importance of eating healthy during pregnancy. For instance, a previous systematic review showed that women tend to increase fruit and vegetable intake and decrease fried and fast food intake during pregnancy (Hillier & Olander, 2017).

For fast food, when women with lower pre-pregnancy BMI had higher positive affect than their average, they were subsequently less likely to consume fast food in the next two hours. In non-pregnant populations, EMA research found no associations between positive affect and unhealthy food intake behaviors (e.g., Jeffers et al., 2020; Mason et al., 2020).

However, theory suggests that positive affect helps promote healthy habits and goal-directed behaviors (Isen, 2001). Therefore, it is possible that when pregnant women with lower pre-pregnancy BMI have higher than average positive affect, they are more motivated to put in the effort required to cook at home.

With regard to the associations between dietary intake and subsequent affect, there was a main effect of sweet food intake on greater negative affect and an interaction between chips/fries intake by weight status on positive affect. Consumption of sugary foods or pastries may be associated with greater negative affect due to guilt about eating (Kuijer & Boyce, 2014), particularly during pregnancy when women may have body image issues and/or are told to eat healthier (Hodgkinson et al., 2014; Institute of Medicine, 2009). Consumption of chips/fries was associated with higher positive affect among women with lower pre-pregnancy BMI and lower positive affect among women with higher pre-pregnancy BMI. It is possible that chips/fries are differentially rewarding for pregnant women depending on pre-pregnancy BMI, but this pattern requires further investigation.

Strengths of the current study include use of EMA in an understudied population of pregnant women and assessment of intake of a range of food categories. There are several limitations to note. This study only examined positive and negative affective states, which was done to reduce the burden on participants during EMA. However, there are other self-conscious emotions that may be relevant to examine in future research on affect and eating during pregnancy (e.g., shame, guilt, anxiety). Furthermore, dietary intake was recalled across a two-hour interval, but relationships among affect and food intake may occur across shorter time scales, which would be missed with the current sampling schedule. Future studies may benefit from having women initiate eating event recordings when they occur (i.e., event-contingent recording). This study also did not assess food portion size, which could be related to affective states. For example, studies have shown affect and binge eating to be highly linked (Engel et al., 2016). Further, while this study uncovered associations between affect and food intake that differed based on pre-pregnancy BMI, we can only speculate on the mechanisms that drive these associations. Future research should study other trait variables related to eating behavior and BMI that may be more salient moderators of affect-behavior relationships. Such factors include emotional eating, hedonic hunger, or dietary restraint, which have all been shown to be positively correlated with BMI.

Overall, this study elucidated bi-directional associations between affect and food intake in low income, Hispanic pregnant women. Results showed that associations differed depending on BMI status. Therefore, future EMA research on momentary determinants of health behaviors in pregnant women should include large samples of various weight groups. Further, future EMA research should examine other factors influencing food availability and choice (e.g., poverty, lack of time, difficulty with accessing and purchasing fresh fruits and vegetables) and how these may influence the association between affective states and food intake in low-income populations. Finally, prevention, intervention, and clinical implications may differ by weight status during pregnancy.

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