1	Association of alpha-aminoadipic acid (2-AAA) with cardiometabolic risk factors in healthy
2	and high-risk individuals
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24 ABSTRACT

25 Plasma levels of the metabolite alpha-aminoadipic acid (2-AAA) have been associated with risk 26 of type 2 diabetes (T2D) and atherosclerosis. However, little is known about the relationship of 27 2-AAA to other cardiometabolic risk markers in pre-disease states, or in the setting of comorbid 28 disease. We measured circulating 2-AAA using two methods in 1) a sample of 261 healthy 29 individuals (2-AAA Study), and 2) in a sample of 134 persons comprising 110 individuals with 30 treated HIV, with or without T2D, a population at high risk of metabolic disease and 31 cardiovascular events despite suppression of circulating virus, and 24 individuals with T2D 32 without HIV (HATIM Study). We examined associations between plasma 2-AAA and markers 33 of cardiometabolic health within each cohort. We observed differences in 2-AAA by sex and 34 race in both cohorts, with higher levels observed in men compared with women, and in Asian 35 compared with Black or white individuals (P<0.05). There was no significant difference in 2-36 AAA by HIV status within individuals with T2D in the HATIM Study. We confirmed 37 associations between 2-AAA and dyslipidemia in both cohorts where high 2-AAA associated 38 with low HDL cholesterol (P<0.001) and high triglycerides (P<0.05). As expected, within the 39 cohort of people with HIV, 2-AAA was higher in the setting of T2D compared to pre-diabetes or 40 normoglycemia (P<0.001). 2-AAA was positively associated with body mass index (BMI) in the 41 2-AAA Study, and with waist circumference and measures of visceral fat volume in HATIM (all 42 P<0.05). Further, 2-AAA associated with increased liver fat in persons with HIV (P<0.001). Our 43 study confirms 2-AAA as a marker of cardiometabolic risk in both healthy individuals and those 44 at high cardiometabolic risk, reveals relationships with adiposity and hepatic steatosis, and 45 highlights important differences by sex and race. Further studies are warranted to establish 46 molecular mechanisms linking 2-AAA to disease in other high-risk populations.

47 **INTRODUCTION**

48 Cardiometabolic diseases, including diabetes and cardiovascular disease (CVD) are 49 increasing in prevalence globally and represent a major contributor to mortality (Tsao et al., 50 2022). Known risk factors include obesity, dyslipidemia, dysregulated glucose metabolism and 51 inflammation (Shah et al., 2018). However, after accounting for these risk factors there remains a 52 high degree of variability in disease susceptibility, and a clear need for more refined biomarkers 53 of cardiometabolic risk to improve our understanding of the underlying disease mechanisms and 54 to improve prediction and treatment of at-risk individuals. 55 Cardiometabolic diseases are characterized by changes in metabolism that may contribute 56 to disease pathophysiology, or may act as biomarkers of disease progression (Upadhyay, 2015). 57 Circulating metabolites that associate with disease states can shed light on underlying disease 58 etiology, biological mechanisms, and may have clinical utility for prediction (Chu et al., 2021). 59 Strategies to identify individuals at high cardiometabolic risk and to modulate disease processes 60 in these individuals before onset of overt disease, would have significant impact in reducing 61 mortality, morbidity, and healthcare costs. For this approach to be successful, early biomarkers 62 of disease that predict at-risk individuals are required, as well as discovering novel pathways for 63 therapeutic targeting. To this end, studying both healthy individuals, as well as individuals with 64 conditions that place them at higher risk of cardiometabolic diseases, may provide an important 65 model to identify novel physiologic relationships. 66 The metabolite alpha-aminoadipic acid (2-AAA) is associated with the development of

67 type 2 diabetes (T2D) (Wang et al., 2013) and atherosclerosis (Saremi et al., 2017), potentially

68 identifying at-risk individuals before development of other known risk markers (Lee et al.,

69 2019). Relatively little is known about the function of 2-AAA, or potential mechanisms linking

70	2-AAA to disease. 2-AAA is derived from the breakdown of the essential amino acid lysine, and
71	is primarily metabolized within mitochondria, with potential involvement in oxidative stress
72	(Estaras et al., 2020; Luna et al., 2021). Elevated 2-AAA is associated with increased insulin
73	secretion, obesity, and dysregulated mitochondrial metabolism (Wang et al., 2013, 2021; Wu et
74	al., 2014; Ho et al., 2016; Plubell et al., 2018; Lee et al., 2019). This makes 2-AAA an
75	interesting novel candidate in cardiometabolic disease biology. However, the relationships
76	between 2-AAA and other cardiometabolic risk markers have not been well-described.
77	The purpose of this study was to characterize the association between 2-AAA and other
78	demographic and circulating markers in a sample of healthy individuals, as well individuals at
79	high risk of metabolic and cardiovascular disease. As chronic viral infections, including treated
80	human immunodeficiency virus (HIV), predispose individuals to a higher incidence of
81	cardiometabolic disease and earlier onset, these conditions can serve as an models of exaggerated
82	or accelerated risk to further identify important physiologic relationships (Barale et al., 2022;
83	Gooden et al., 2022; Rivera et al., 2022; Spieler et al., 2022). Here, we assess the relationship of
84	2-AAA with range of cardiometabolic disease conditions and risk factors among healthy
85	individuals and those with treated HIV infection.
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88	MATERIALS AND METHODS
89	Study Populations: Samples and data from two independent studies are included here.

90 Participants of both studies were recruited from the same geographic area (Nashville, TN, and

91 surrounding areas), and study procedures completed at Vanderbilt University Medical Center.

92

93 Determinants of 2-AAA: Screening Study (2-AAA Study)

94 Healthy adults (non-pregnant and non-lactating women and men, age 18-45 years) were recruited 95 to complete a single study visit as part of a cross-sectional study at Vanderbilt University 96 Medical Center between November 2018 and June 2021. Exclusion criteria included body mass index (BMI) >30 kg/m², active use of tobacco products, active use of prescription medications 97 98 (apart from hormonal birth control), and diagnosis of diabetes mellitus, cardiovascular disease, 99 renal disease, liver disease, or bleeding disorders. Data for 261 individuals who completed study 100 procedures (vital signs, anthropometric measurements), provided a fasting blood sample, and had 101 sufficient plasma available for 2-AAA measurement are included in the current analysis. All 102 participants provided written, informed consent, and the study was approved by the Vanderbilt 103 University Institutional Review Board.

104

105 The HIV, Adipose Tissue Immunology, and Metabolism Study (HATIM) Study

106 Adults with human immunodeficiency virus (HIV, N=112) were recruited from the Vanderbilt 107 Comprehensive Care Clinic between August 2017 and November 2019. Participants were on 108 combination antiretroviral therapy (ART) for ≥ 18 months, with a minimum of 12 months of 109 sustained suppression of plasma viremia at enrollment and had no known inflammatory or 110 rheumatologic conditions. Exclusion criteria were self-reported heavy alcohol use (>11 111 drinks/week), known cirrhosis, active hepatitis B or C, cocaine or amphetamine use, and use of 112 corticosteroids or growth hormones. By design and to enrich for the presence of cardiometabolic 113 disease, the cohort enrolled approximately equal numbers of individuals who were 114 normoglycemic (HbA1c < 5.7 or fasting blood glucose (FBG) < 100 mg/dL); pre-diabetes 115 (HbA1c 5.7%-6.4% and/or FBG 100-126 mg/dL); and diabetes (HbA1c \geq 6.4%, and/or FBG \geq

116 126 mg/dL or on diabetes medication). To allow for direct comparison of 2-AAA levels with
HIV-negative individuals, the study also recruited individuals with diabetes but without HIV
(N=24). Participants provided written, informed consent, and the study was approved by the
Vanderbilt University Institutional Review Board (ClinicalTrials.gov Identifier: NCT04451980).

121 Measurement of 2-AAA

122 In the 2-AAA Study, plasma levels of 2-AAA were quantified by liquid chromatography mass

123 spectrometry (LCMS) at the Vanderbilt Mass Spectrometry Core. Samples were spiked with

124 internal standard (Arginine-15N4, Sigma Aldrich), extracted with methanol, and derivatized with

125 dansyl chloride (Sigma Aldrich) prior to analysis. The dansyl derivative of 2-AAA ([M+H]+

126 395.1271) was measured by targeted selected ion monitoring (SIM) using a Vanquish ultrahigh

127 performance liquid chromatography (UHPLC) system interfaced to a QExactive HF

128 quadrupole/orbitrap mass spectrometer (Thermo Fisher Scientific). Data acquisition and

129 quantitative spectral analysis were conducted using Thermo-Finnigan Xcaliber version 4.1 and

130 Thermo-Finnigan LCQuan version 2.7, respectively. Calibration curves were constructed by

131 plotting peak area ratios (2-AAA / Arg-15N4) against analyte concentrations for a series of 2-

132 AAA standards. Electrospray ionization source parameters were tuned and optimized using an

133 authentic 2-AAA reference standard (Sigma Aldrich) derivatized with dansyl chloride and

134 desalted by solid phase extraction prior to direct liquid infusion.

In the *HATIM Study*, plasma 2-AAA was measured as part of a metabolomics panel, at the Southeast Center for Integrated Metabolomics (SECIM) at the University of Florida, using previously described methods (O'Kell et al., 2017, 2019). Briefly, plasma samples were spiked with internal standards solution. Proteins were precipitated using 8:1:1 Acetonitrile: Methanol:

139 Acetone (Fisher Scientific, San Jose, CA), and the supernatant dried under a gentle stream of 140 nitrogen at 30°C (Organomation Associates, Inc., Berlin, MA). Samples were reconstituted with 141 injection standards solution. LC-MS untargeted metabolomics was performed on a Thermo Q-142 Exactive Orbitrap mass spectrometer equipped with a Dionex UPLC system (Thermo, San Jose, 143 CA). Percent relative standard deviation of internal standard peak areas were calculated to 144 evaluate extraction and injection reproducibility. Mzmine 2 was used to identify features, 145 deisotope, align features and perform gap filling. The data was searched against SECIM internal 146 retention time metabolite library. All adducts and complexes were identified and removed from 147 the data set. Ion counts from features mapping to alpha-aminoadipic acid in positive ion mode 148 were summed for analysis. Because measurement of 2-AAA was conducted at different sites, 149 studies were analyzed separately.

150

151 Lipid and Biomarker Measurement

152 In the 2-AAA Study, serum lipids were profiled at the Vanderbilt Lipid Laboratory. Briefly, total 153 cholesterol and triglycerides (TG) were measured by standard enzymatic assays. High-density 154 lipoprotein (HDL) was measured with the enzymatic method after precipitation of VLDL and 155 LDL using polyethylene glycol reagent (PEG). LDL cholesterol was calculated using the 156 Friedewald equation (Friedewald et al., 1972). In the HATIM Study, fasting plasma HDL, LDL, 157 and TG were measured using the selective enzyme hydrolysis method (Abbott, Chicago, IL). In 158 the 2-AAA Study, fasting glucose was measured at the study visit by finger prick (AimStrip Plus 159 Blood Glucose Meter, Germaine Laboratories Inc., San Antonio TX). In the HATIM Study, insulin was measured by radioimmunoassay (Millipore Cat. # PI-13K). The assay utilizes ¹²⁵I -160 161 labeled insulin and a double antibody/PEG technique to determine serum insulin levels. The

assay was modified by the Vanderbilt Hormone and Analytical Services Core to improve the

163 sensitivity to 1uU/ml(0.04ng/ml). Glucose and hemoglobin A1c (HbA1c) were measured in

164 fasting blood samples at the Vanderbilt Clinical Chemistry Laboratory.

165

166 Body Composition Analysis

167 In the HATIM Study, individuals underwent computed tomography (CT) imaging using a

168 Siemens Somatom Force multidetector scanner (Erlangen, Germany) to acquire chest, abdominal

and liver images, as described (Gabriel et al., 2021; Bailin et al., 2022). Briefly, separate non-

170 contrast electrocardiogram-gated thorax (top of the aortic arch through the lung base) and

abdominal (diaphragm to lumbosacral junction) scans were performed using a scanning protocol

and image interpretation approach previously described (Carr et al., 2005; VanWagner et al.,

173 2014; Terry et al., 2017). Abdominal subcutaneous adipose tissue (SAT) and visceral adipose

174 tissue (VAT) volumes were measured within a 10-mm block of images consisting of eight

175 images, 1.25-mm thick, at the L4-5 vertebrae using Osirix software. Pericardial adipose tissue

176 (PAT) volume was measured within a 45-mm block of images spanning 15 mm above and 30

177 mm below the superior extent of the left main coronary artery, which includes the adipose tissue

178 located around the epicardial coronary arteries (left main coronary, left anterior descending, right

179 coronary, and circumflex arteries) as well as the epicardial and PAT around the coronary arteries

180 (Alman et al., 2016; Miljkovic et al., 2020). Images at T12-L1 were used to identify the liver

181 below the right diaphragm corresponding to superior aspects of the right and medial lobes or

182 hepatic segments 4a, 7, and 8 using the Couinaud classification system. Three regions of interest

183 within homogenous portions of the liver at three levels were identified and liver density was

averaged from the nine total regions. Tissue radiodensity was quantified using the Hounsfield
Units scale where water has a value of 0 HU and air has a value of -1000 HU.

186

187 Statistical Analysis

188 Plasma 2-AAA was assessed for normality of distribution through visualization, and testing for 189 skewness and kurtosis, and was found to follow a normal distribution in both the 2-AAA and 190 HATIM studies. Two individuals were considered outliers for 2-AAA in HATIM (>3 SD from 191 the mean) and were removed prior to analysis. Associations between 2-AAA and continuous 192 variables were analyzed using linear regression models. Analyses between 2-AAA and discrete 193 variables were analyzed by T-test or ANOVA. Models were adjusted for sex and race in both 194 studies and for additional covariates in HATIM (smoking, diabetes group). Models were further 195 adjusted for other risk factors as indicated in the corresponding results sections, including BMI, 196 cholesterol, HDL, LDL, TG, fasting glucose. P<0.05 was considered statistically significant, and 197 Bonferroni P<0.05 considered statistically significant for post hoc multiple testing correction. 198 Analyses were completed and results visualized using IBM SPSS Statistics version 28 (IBM, 199 Armonk NY) and GraphPad Prism version 9.4.1 (GraphPad Software, San Diego, CA). 200

201

202 **RESULTS**

The characteristics of the participants of the 2-AAA Study are shown in **Table 1**. Characteristics of the participants of the HATIM Study are shown in **Table 2**. Participants of the 2-AAA study were 72% female, and 74% white, with an average age of 28 years. Participants of the HATIM study were 67% male, and 54% white, with an average age of 48 years. Plasma 2-AAA in

persons with HIV (PWH) with diabetes (ion count $312 \times 10^4 \pm 75 \times 10^4$) was slightly higher than that in HIV-negative with diabetes (ion count $271 \times 10^4 \pm 74 \times 10^4$), but the difference was not statistically significant (P=0.08).

210

211 Plasma 2-AAA levels are higher in men than in women, and higher in Asian individuals

212 There was a significant difference in plasma 2-AAA by sex in the 2-AAA Study, with higher

213 levels in men than in women (plasma 2-AAA 95.99±33.7 vs. 68.43±27.7 ng/ml, P<0.0001;

Figure 1A). A similar difference by sex was observed in the HATIM Study samples, with higher

levels in men than women (plasma 2-AAA ion count $281 \times 10^4 \pm 73 \times 10^4 \text{ vs.}$ $242 \times 10^4 \pm 65 \times 10^4$

216 ion count, P=0.004; Figure 1C). Because other risk factors also differ by sex, we performed

217 stepwise linear regression models including risk factors (BMI, fasting glucose, cholesterol, HDL,

LDL, TG), and found that the associations with sex remained significant (P<0.001 2-AAA Study,

219 P<0.02 HATIM Study). We observed a significant difference by self-reported race in the 2-AAA

220 Study (Overall P=0.002; Figure 1B), with individuals self-identifying as Asian having

borderline significantly higher plasma 2-AAA ($95.68 \pm 35.5 \text{ ng/ml}$) compared with individuals

self-identifying as Black or African American (72.26 \pm 30.0 ng/ml, P=0.05), or white (72.73 \pm

223 30.7 ng/ml, P=0.007). This was not attributable to differences in sex distribution or risk factors

between groups. In fact, Asian individuals in the 2-AAA Study had significantly lower BMI

225 (P=0.018) and systolic blood pressure (P=0.005) than other individuals. Interestingly, there was

also an overall difference by self-reported race in the HATIM sample (P=0.014; Figure 1D),

with a trend towards higher levels of 2-AAA in Asian (2-AAA ion count 359 $\times 10^4 \pm 45 \times 10^4$)

compared to Black (2-AAA ion count 249 $\times 10^4 \pm 65 \times 10^4$) and white (2-AAA ion count 279 $\times 10^4$

 $\pm 75 \times 10^4$) individuals, although there were only three individuals self-identifying as Asian in this

- sample, so the differences did not reach statistical significance in post hoc tests. There was no
- association between 2-AAA and age in either dataset.
- 232

233 Plasma 2-AAA levels associate with dyslipidemia in healthy individuals and PWH

- Higher plasma 2-AAA was associated with lower HDL cholesterol (2-AAA Study $r^2=0.267$,
- P<0.001; HATIM r^2 =0.579, P<0.001; Figure 2 A, B), and higher triglycerides (2-AAA Study

236 $r^2=0.246$, P=0.027; HATIM $r^2=0.526$, P=0.007; **Figure 2 C, D**). There was no significant

- association with LDL cholesterol.
- 238

239 Higher plasma 2-AAA levels associate with diabetes status in PWH

240 There were significant differences in plasma 2-AAA by diabetes status within PWH in the

241 HATIM sample (P<0.001, Figure 3). Individuals with diabetes had significantly higher levels of

242 2-AAA (ion count $312 \times 10^4 \pm 75 \times 10^4$) than both the insulin sensitive (ion count $233 \times 10^4 \pm$

243 $60x10^4$, P<0.001) and the pre-diabetic (ion count $262x10^4 \pm 58x10^4$, P=0.005) groups in models

adjusted for sex, race, BMI and smoking status.

245

246 Plasma 2-AAA associates with elevated fasting glucose, insulin, and HbA1c in PWH

247 Across all PWH individuals in HATIM, plasma 2-AAA was associated with increased fasting

- 248 glucose (r2=0.576, P<0.001), fasting insulin (r2=0.623, P<0.001), HOMA-IR (r2=0.538,
- 249 P<0.001) and hemoglobin A1c (r2=0.580, P<0.001). In secondary analyses split by diabetes
- status, 2-AAA associated with glucose and HbA1c only in the individuals with diabetes
- 251 (P<0.0001 for diabetes, vs P>0.5 for insulin sensitive and pre-diabetes), but 2-AAA was
- associated with insulin in both people with and without diabetes (P<0.02 insulin sensitive,

253	P < 0.002 diabetes). In the 2-AAA Study, a small number of people (n=25) had evidence of
254	potential impaired fasting glucose (IFG, defined as glucose >100mg/dL but <125 mg/dL). While
255	plasma 2-AAA levels were slightly higher within the individuals with IFG (82.5 vs. 75.4 ng/ml),
256	this difference did not reach statistical significance.
257	
258	Elevated plasma 2-AAA levels associate with differences in anthropometrics, adipose tissue,
259	and liver density
260	We found a significant association between plasma 2-AAA and higher BMI in the 2-AAA Study
261	$(r^2=0.275, P<0.001, model adjusted for sex and race), but this was not significant in HATIM.$
262	However, in HATIM, higher plasma 2-AAA was significantly associated with increased waist
263	circumference ($r^2=0.219$, P<0.001), as well as greater visceral adipose tissue volume ($r^2=0.225$,
264	P<0.001), but not with measures of subcutaneous or pericardial adipose tissue. In HATIM, 2-
265	AAA was negatively associated with liver density ($r^2=0.192$, P=0.003; Figure 4). Lower liver
266	density is a marker of higher proportion of ectopic fat in the liver.
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269	DISCUSSION
270	We measured plasma 2-AAA in two independent samples of individuals across the spectrum of
271	healthy (no diagnosed diseases) to high cardiometabolic risk (diabetes and treated HIV
272	infection). 2-AAA was elevated in diabetes but did not appear to be significantly elevated based
273	on HIV status. We found that plasma 2-AAA is elevated in men compared with women, and in
274	Asian compared with other self-identified ancestries. These associations are constant in both
275	healthy individuals and PWH. We confirmed associations between 2-AAA and both low HDL

and high TG, and between 2-AAA and diabetes. We report novel relationships between 2-AAA
and visceral adipose tissue measured by CT, and between 2-AAA and higher liver fat. Our data
further confirm 2-AAA as an important candidate for further prognostic and therapeutic
consideration.

280 Plasma 2-AAA levels differed by sex, an association that has been reported previously in 281 Mexican young adults (Guevara-Cruz et al., 2018). Men have relatively higher risk of CVD than 282 pre-menopausal women, yet the mechanisms underlying this difference are not fully understood 283 (Tsao et al., 2022). We further report differences by self-reported race, with Asian individuals 284 having higher 2-AAA than other groups. Individuals of Asian ancestry have relatively higher risk 285 of T2D and some CVD given the same risk factor profile as individuals of European ancestry 286 (Ma and Chan, 2013; Buljubasic et al., 2020). The mechanisms underlying this are incompletely 287 understood, and the risk factor profile for CVD in Asians may differ when compared with 288 European ancestry (Paul et al., 2017). While the original discovery of 2-AAA as a diabetes 289 metabolite was in European ancestry (Wang et al., 2013), 2-AAA has also been reported to 290 associate with T2D in Chinese individuals (Wang et al., 2022). Whether differences in 2-AAA 291 may play a role in mediating the relative increased risk in men compared with women, and Asian 292 compared with other ancestries, remains to be determined.

We previously reported that plasma 2-AAA associates with both lower HDL cholesterol and higher triglycerides (Shi et al., 2022). We replicated those associations in the current study, establishing that this relationship is consistent across multiple different samples, including in a cohort of persons with HIV. Based on genetic evidence, 2-AAA drives the decrease in HDL (Shi et al., 2022). While low HDL cholesterol is consistently associated with increased cardiometabolic risk (Castelli et al., 1986; Emerging Risk Factors Collaboration et al., 2009),

interventions to alter HDL have shown no benefit (Kingwell et al., 2014). This could be due to
differences in HDL composition or function, or due to a causal biomarker that is upstream of
HDL. This raises the intriguing hypothesis that elevated 2-AAA, rather than low HDL per se,
may be driving increased cardiometabolic risk. However, careful mechanistic studies are
required to interrogate this further.

304 2-AAA was originally discovered as a predictor of diabetes, and is associated with 305 increased insulin secretion in animal models and cells (Wang et al., 2013). In the setting of 306 experimental hyperglycemia in overweight and obese, but otherwise healthy individuals, 2-AAA 307 was significantly decreased following 24 hours of hyperglycemia (Perkins et al., 2019). 2-AAA 308 has been shown to be reduced in the acute setting in response to insulin infusion (Irving et al., 309 2015). We found that 2-AAA was significantly higher in PWH who have diabetes, than in PWH 310 who were insulin sensitive or pre-diabetic. This is similar to what has been reported in HIV-311 negative individuals (Wang et al., 2013; Razquin et al., 2019), and suggests that the relationship 312 between 2-AAA and diabetes is consistent across different settings, including against the 313 background of well-controlled HIV infection, a population at increased risk of cardiometabolic 314 disease (Spieler et al., 2022). We found no significant difference in plasma 2-AAA levels based 315 on HIV status in the HATIM cohort within the subset of individuals with diabetes, further 316 suggesting that 2-AAA is a useful biomarker of cardiometabolic risk in multiple at-risk 317 populations. 2-AAA was associated with increased fasting glucose, fasting insulin, and 318 hemoglobin A1c in the HATIM study. However, the association between 2-AAA and glucose 319 was only significant in individuals with diabetes; 2-AAA was not associated with fasting glucose 320 in insulin sensitive individuals in the 2-AAA Study or HATIM, or in individuals with pre-321 diabetes in HATIM. In contrast, 2-AAA was associated with higher insulin in individuals with

322 and without diabetes. This distinction between the glycemic and insulin axis is consistent with 323 the hypothesis that 2-AAA is an early marker or driver of hyperinsulinemia and is associated 324 with elevated insulin before the development of overt hyperglycemia or diabetes. These data 325 further support a mechanism where elevated 2-AAA precedes the onset of hyperglycemia, and 326 associates with hyperinsulinemia even in individuals who appear insulin sensitive. Associations 327 between 2-AAA and hyperglycemia are likely secondary to insulin resistance. However, further 328 in-depth studies are required to assess potential reciprocal regulation of 2-AAA and insulin. 329 2-AAA was positively associated with BMI in the 2-AAA study, but not in the HATIM 330 study. However, there was a significant association between 2-AAA and waist circumference in 331 HATIM. This may suggest that the relationship between 2-AAA and adiposity is modulated by 332 HIV-associated effects on adipose distribution (Koethe et al., 2020). Previous studies have also 333 highlighted an association between 2-AAA and obesity, including both BMI and waist 334 circumference (Dugas et al., 2016; Ho et al., 2016; Libert et al., 2018; Lee et al., 2019). While 335 one study has found that 2-AAA is protective against obesity and diabetes in mice (Xu et al., 336 2019), these findings are in contrast to all other studies, and may be related to specific metabolic 337 anomalies in the mouse model used (Xu et al., 2018; Wang et al., 2021, 1). In our study, 2-AAA 338 associated with increased visceral fat in HATIM, but not subcutaneous or pericardial fat. These 339 data are consistent with a previous study, where 2-AAA was associated with metabolically 340 unhealthy central obesity, compared with metabolically healthy peripheral obesity (Gao et al., 341 2016). Thus, 2-AAA may relate specifically to pathogenic adipose tissue dysfunction, rather than 342 to obesity itself.

Plasma 2-AAA associated with lower liver density, which corresponds to higher liver fat,
and is considered a measure of hepatic steatosis. Previous data in mice found an association

345 between 2-AAA and liver mass (Wu et al., 2014), however, to our knowledge our study 346 describes this for the first time in humans. Elevated 2-AAA may thus be a risk factor for hepatic 347 steatosis and development of fatty liver disease, however, whether this is independent of 348 associations with BMI, visceral fat and circulating lipids remains to be determined. 349 Our study had several strengths. We analyzed plasma 2-AAA in two separate samples of 350 well-phenotyped individuals, including both healthy individuals and PWH across the diabetes 351 spectrum, allowing us to assess whether the relationship between 2-AAA and cardiometabolic 352 risk markers is consistent in the settings of chronic viral-induced inflammation and in 353 individuals without diagnosed disease. 2-AAA was not measured in many previous 354 metabolomic studies, and is not consistently detected or reported on popular metabolomics 355 panels (e.g. Metabolon). Thus, the importance of this metabolite in cardiometabolic health may 356 be under-appreciated. We used a targeted assay in the 2-AAA study to quantify 2-AAA. 357 providing important data on circulating levels in healthy individuals. To our knowledge, this is 358 the first study to measure associations between 2-AAA and metabolic disease in PWH. PWH 359 suffer a disproportionate burden of diabetes, hypertension, fatty liver, and dyslipidemia 360 compared to HIV negative persons (Currier et al., 2008; Vodkin et al., 2015; Maurice et al., 361 2017; Nansseu et al., 2018), and allows for validation of the relevance of 2-AAA to disease 362 within the setting of a highly-inflammatory exaggerated phenotype. Our study also had some 363 limitations. Plasma 2-AAA was measured using a different method in HATIM compared with 364 the 2-AAA study, limiting our ability to directly compare levels of 2-AAA in PWH compared 365 with healthy individuals. However, we were able to compare levels between PWH and HIV-366 negative within a subset of individuals with diabetes. We also had limited sample size to fully

367	characterize the differences by race across both samples, with small numbers of Black
368	individuals in the 2-AAA study and small numbers of Asian individuals in the HATIM study.
369	In conclusion, our study establishes differences in plasma 2-AAA by sex and race,
370	confirms associations between 2-AAA and dyslipidemia in both healthy individuals and PWH
371	with or without diabetes, and highlights novel relationships between 2-AAA and liver fat and
372	visceral adipose tissue. Further mechanistic and longitudinal studies are required to establish
373	whether 2-AAA is causally linked to cardiometabolic disease.
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388	The authors have no relevant disclosures.
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569 TABLES & FIGURES

570

- 571 Figure 1. Plasma 2-AAA is significantly higher in men than women in the 2-AAA (A) and
- 572 HATIM Study (C). 2-AAA is higher in Asian compared to Black or white individuals in
- 573 the 2-AAA Study (B) with a similar trend in the HATIM Study (D). Data are expressed as
- 574 ng/ml for data from the 2-AAA Study and ion counts for the HATIM study.



- 592 Figure 2. Plasma 2-AAA associates with lower HDL cholesterol and higher Triglycerides in
- 593 the 2-AAA (A, C) and HATIM (B, D) studies. Data are expressed as ng/ml for data from the 2-
- 594 AAA Study and ion counts for the HATIM study.
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601 Figure 3. Plasma 2-AAA was significantly higher in PWH and diabetes, compared with



602 **PWH who were insulin sensitive or with pre-diabetes.**

Figure 4. Plasma 2-AAA was negatively associated with liver attenuation in the HATIM

Study of PWH.



630 Table 1. Characteristics of the participants of the 2-AAA Screening Study

	Male (N=72)	Female (N=189)
	Mean (SD)	Mean (SD)
Age (years)	28.96 (6.92)	27.76 (7.2)
Race (N Black, white, Asian, other)	3, 55, 8, 6	14, 139, 25, 11
BMI (kg/m^2)	24.77 (2.9)	22.94 (2.9)
Systolic Blood Pressure (mmHg)	120.57 (13.5)	111.26 (10.7)
Diastolic Blood Pressure (mmHg)	73.80 (9.5)	69.67 (8.1)
Glucose (mg/dL)	90.0 (7.6)	91.18 (8.2)
Total cholesterol (mg/dL)	166.56 (30.7)	167.94 (32.4)
HDL (mg/dL)	53.47 (10.7)	64.67 (13.1)
LDL (mg/dL)	95.56 (24.9)	87.41 (24.5)
TG (mg/dL)	87.44 (37.5)	79.29 (37.1)
2-AAA (ng/ml)	95.99 (33.8)	68.44 (27.7)

Table 2. Characteristics of participants of the HATIM Study

HIV-negative

	Insulin sensitive		Pre-Diabetes		Diabetes		Diabetes	
	Male (N=33)	Female (N=8)	Male (N=27)	Female (N=7)	Male (N=24)	Female (N=11)	Male (N=6)	Female (N=18)
	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)
Age (years)	42.15 (11.7)	46.37 (7.7)	44.40 (12.2)	49.14 (10.2)	52.62 (9.4)	48 (12.0)	51.33 (12.1)	57.61 (9.6)
Race (N Black, white, Asian, other)	10, 22, 1, 0	5, 2, 0, 1	11, 14, 1, 1	4, 3, 0, 0	10, 14, 0, 0	7, 1, 0, 3	1, 5, 0, 0	6, 11, 1, 0
BMI (kg/m ²)	30.78 (3.8)	34.21 (5.8)	33.28 (6.3)	34.03 (6.1)	34.76 (7.25)	40.01 (9.8)	38.16 (8.8)	37.31 (5.1)
Waist circumference (cm)	100.39 (12.6)	104 (12.5)	105.40 (13.9)	102.86 (14.8)	115.21 (12.6)	114.7 (17.5)	126.31 (20.4)	114.7 (13.85)
Total cholesterol (mg/dL)	174.63 (38.3)	186.6 (23.0)	175.7 (35.1)	223.71 (38.5)	173.75 (33.1)	179.45 (41.3)	183.66 (85.8)	171.44 (27.9)
LDL (mg/dL)	102.12 (34.1)	110.3 (18.8)	110.77 (44.9)	129 (35)	91.39 (28.9)	95.2 (33.9)	86.83 (27.0)	102.0 (27.1)
HDL (mg/dL)	47.06 (18.2)	54.1 (15.9)	41.22 (13.3)	65.43 (23.6)	37.96 (10.5)	49.72 (12.9)	36.33 (10.1)	47.55 (8.8)

TG	127.27 (82.9)	111.8 (64.2)	154.89 (83.4)	146.86 (56.8)	250.29	189.27 (145.4)	327.16	109.33 (45.4)
					(224.9)		(470.9)	
Glucose	87.0 (9.5)	89.38 (6.0)	111.55 (14.0)	112.42 (8.1)	203.6 (88.4)	156 (59.7)	164.33 (78.5)	128.61 (34.9)
(mg/dL)								
Insulin	19.19 (23.0)	18.39 (19.4)	55.06 (59.2)	30.9 (24.9)	38.58 (21.2)	36.62 (19.3)	46.99 (29.4)	28.92 (16.3)
(uU/mL)								
HOMA-IR	4.23 (4.9)	4.24 (4.8)	14.91 (15.6)	9.10 (7.7)	22.63 (19.1)	16.49 (14.8)	17.78 (13.0)	8.73 (6.5)
Hemoglobin	5.1 (0.46)	5.16 (0.23)	5.52 (0.5)	5.52 (0.3)	8.18 (2.3)	7.10 (1.8)	7.81 (2.0)	6.85 (0.93)
A1c (%)								
Liver	61.44 (7.5)	63.87 (3.4)	62.17 (9.1)	61.3 (12.9)	53.03 (14.6)	57.54 (11.5)	48.25 (11.6)	45.62 (19.0)
attenuation								
(HU)								
Plasma 2-AAA	2441358	1898487	2769029	2059280	3270163	2794366	3141535	2576285
(ion count)	(623282)	(225878)	(548804)	(355932)	(715556)	(739755)	(969840)	(622577)