Metabolically Defined Body Size Phenotypes and Risk of Endometrial Cancer in the European Prospective Investigation into Cancer and Nutrition (EPIC)

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ABSTRACT

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Background: Obesity is a risk factor for endometrial cancer but whether metabolic dysfunction is associated with endometrial cancer independent of body size is not known.

Methods: The association of metabolically defined body size phenotypes with endometrial cancer risk was investigated in a nested case–control study (817 cases/ 817 controls) within the European Prospective Investigation into Cancer and Nutrition (EPIC). Concentrations of C-peptide were used to define metabolically healthy (MH; <1st tertile) and metabolically unhealthy (MU; ≥1st tertile) status among the control participants. These metabolic health definitions were combined with normal weight (NW); body mass index $(BMI) < 25$ kg/m² or waist circumference (WC)<80 cm or waist-to-hip ratio (WHR)<0.8) and overweight (OW; BMI≥25 kg/m² or WC≥80 cm or WHR≥0.8) status, generating four phenotype groups for each anthropometric measure: (i) MH/NW, (ii) MH/OW, (iii) MU/ NW, and (iv) MU/OW.

Introduction

Endometrial cancer is the second most common gynecological cancer worldwide, with 604,127 new cases and 341,831 deaths

Results: In a multivariable-adjusted conditional logistic regression model, compared with MH/NW individuals, endometrial cancer risk was higher among those classified as MU/NW $[OR_{WC}, 1.48; 95%$ confidence interval (CI), 1.05-2.10 and OR_{WHR}, 1.68; 95% CI, 1.21-2.35] and MU/OW (OR_{BMI}, 2.38; 95% CI, 1.73-3.27; OR_{WC}, 2.69; 95% CI, 1.92–3.77 and ORWHR, 1.83; 95% CI, 1.32–2.54). MH/OW individuals were also at increased endometrial cancer risk compared with MH/NW individuals (OR_{WC} , 1.94; 95% CI, 1.24-3.04).

Conclusions: Women with metabolic dysfunction appear to have higher risk of endometrial cancer regardless of their body size. However, OW status raises endometrial cancer risk even among women with lower insulin levels, suggesting that obesityrelated pathways are relevant for the development of this cancer beyond insulin.

Impact:Classifying women by metabolic health may be of greater utility in identifying those at higher risk for endometrial cancer than anthropometry per se.

reported in 2020 (1). Higher body mass index (BMI≥25 kg/m²) is a well-established risk factor for endometrial cancer (2–5). A metaanalysis of prospective studies has shown that every 5 kg/m^2 increase in BMI is associated with a 60% increase in endometrial

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Note: Supplementary data for this article are available at Cancer Epidemiology, Biomarkers & Prevention Online (http://cebp.aacrjournals.org/).

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cancer risk (6). Recently, several studies have also shown that waist circumference (WC) and waist-to-hip ratio (WHR), both indicators of central adiposity, may be associated with endometrial cancer risk independently of BMI (7, 8). Potential biological mechanisms linking obesity with endometrial cancer development include alterations in the metabolism of endogenous hormones, such as sex steroids, insulin, and inflammation (9–11).

Hyperinsulinemia, a condition characterized by elevated levels of insulin in the fasting state, has been positively associated with endometrial cancer risk in several prospective studies (12, 13), and in a Mendelian randomization analysis (5). C-peptide, a marker for pancreatic insulin secretion, has also generally been associated with endometrial cancer risk (12, 14). Mechanistically, insulin may promote endometrial cancer development through direct mitogenic effects on the growth of endometrial cells, and indirectly via sex hormone disruption (15, 16).

Metabolic dysfunction has been associated with a number of adverse health outcomes independent of BMI (17–26). Indeed, over a third of adults in the normal weight (NW) range may have metabolic dysfunction that puts them at elevated cardiometabolic disease risk (27). Accumulating evidence suggests that individuals with metabolic dysfunction, either in the NW or overweight (OW)/obese BMI range, are at greater risk of developing colorectal, breast, pancreatic, prostate and bladder cancers, compared with subjects who are metabolically healthy (MH; refs. 17, 18, 24, 25, 28). However, whether metabolic dysregulation also raises endometrial cancer risk independent of obesity is less clear. A study conducted within the Framingham Heart Study found that metabolic dysregulation (based on elevated blood glucose) was associated with higher risk of endometrial cancer among women with OW and obesity, but not among women within the normal range of BMI and WHR (20). However, another study in the SEER-Medicare–linked database found that metabolic syndrome (comprised of having three or more parameters out of clinical range, including central obesity, fasting glucose, blood pressure, and triglycerides) remained associated with endometrial cancer even after adjusting for level of obesity (29). However, to our knowledge no studies have specifically evaluated hyperinsulinemia in relation to endometrial cancer according to body size in a large-scale prospective cohort.

To address these current gaps in the literature, we conducted an investigation of metabolically defined body size phenotypes (based on C-peptide levels combined with anthropometric measures) and their association with endometrial cancer risk in a nested case– control study within the European Prospective Investigation into Cancer and Nutrition (EPIC).

Materials and Methods

Study population

EPIC is an ongoing multicenter prospective cohort study designed to assess the relationship between diet, lifestyle, and genetic and metabolic factors with cancer and other chronic diseases. A detailed description of the cohort has been published elsewhere (30, 31). In summary, a total of 521,324 participants (\sim 70% female) were recruited between 1992 and 2000 from 23 centers across 10 European countries (Denmark, France, Germany, Greece, Italy, the Netherlands, Norway, Spain, Sweden, and the United Kingdom). Written informed consent was provided by all participants. The study was in accordance with human subjects' protection principles (Declaration of Helsinki) and was approved by the ethical review boards from the International Agency for Research on Cancer (IARC) and from all local centers.

Follow-up and ascertainment of endometrial cancer

Incident endometrial cancer cases were identified using cancer registries in Norway, United Kingdom, Spain, Italy, and the Netherlands and using a combination of sources such as active follow-up of study subjects, cancer and pathology registries, and health insurance records in France and Germany. The collection and standardization of clinical and pathological data on each cancer site were performed following a detailed protocol. The end of follow-up was established as the latest date of follow-up for cancer incidence, death or end of follow-up, whichever came first. Censoring dates for complete follow-up from cancer registries were between December 2009 and December 2013. Endometrial cancer cases (C540–549) were identified using the $10th$ Revision of the International Classification of Diseases ICD-10) and the 3rd Revision of the International Classification of Diseases for Oncology (ICD-O-3). Endometrial cancer type 1 histologies included endometrioid adenocarcinoma, adenosquamous carcinoma, adenocarcinoma with squamous metaplasia, adenocarcinoma not otherwise specified, adenocarcinoma in adenomatous polyp, mucinous adenocarcinoma, mucin-producing adenocarcinoma (codes 8380, 8560, 8570, 8140, 8210, 8480, and 8481). The inclusion of adenocarcinoma not otherwise specified in Type 1 is justified because endometrioid adenocarcinoma is the most common type of adenocarcinoma. Type 2 histologies included squamous cell carcinoma, clear cell adenocarcinoma, mixed cell adenocarcinoma, serous cystadenocarcinoma, papillary serous cystadenocarcinoma (codes 8070, 8310, 8323, 8441, and 8460). Other histologies were not classified into either type (codes 8000, 8010, 8020, 8260, 8950, and 8980).

Selection of case and control subjects

Incident endometrial cancer cases were identified after the baseline blood collection and before the end of the follow up in each study center. Women who had a previous cancer or had undergone hysterectomy at the time of blood collection were excluded. For each case, one control participant was randomly chosen from the overall EPIC cohort of women who were free of cancer at the time of diagnosis of the index case. An incidence density sampling protocol for control selection was used, such that controls could include participants who became a case later in time, whereas each control could also be sampled more than once. The matching factors for cases and controls were study center, fasting status, age at blood collection, time of day at blood collection $(\pm 4 h)$, menopausal status, exogenous hormone use and phase of menstrual cycle at blood collection.

Laboratory measurements

Blood samples were collected at baseline according to standardized procedures and stored in the central EPIC biorepository at IARC $(-196^{\circ}C,$ liquid nitrogen) for all countries included in this study. C-peptide was measured in two phases. In the first phase, 378 serum samples were measured by an immunoradiometric assay (Immunotech), with intrabatch coefficients of variation (CV) <3% and interbatch CVs <11% for a C-peptide concentration of 0.50 nmol/L (14). In the second phase, 1,256 plasma samples were measured by an ELISA assay (Mercodia) with intrabatch CV <7% and interbatch CVs <6% for a C-peptide concentration of 0.66 nmol/l (32). All measurements were performed in the immunoassay laboratory at IARC. Samples from matched case–control sets were assayed in the same analytical batch. Laboratory personnel were blinded to case–control status of the samples. Concentrations of C-peptide for cases and controls by method of analysis are presented in Supplementary Table S1.

Assessment of anthropometric, lifestyle, and dietary exposures

All participants underwent assessment of anthropometrics, lifestyle, dietary intake, medical history, and demographics at baseline. Standard protocols for the measurement of body weight and height were used in all centers, except for Oxford, and Norway where these were self-reported. However, previous studies have shown these self-reported anthropometric measures are valid for identifying associations in epidemiological studies (33, 34). Assessed weight and height were used to calculate BMI (kg/m²). WC was measured either at the narrowest torso circumference or at the midpoint between the lower ribs and iliac crest. WC was divided by hip circumference to generate the WHR. Lifestyle and medical history self-reported questionnaires collected information on education, smoking status, alcohol consumption, and physical activity level, diabetes, and reproductive history (menopausal status, oral contraceptive use, menopausal hormone use, age at menarche and menopause, and age and number of full-term pregnancies). The validated Cambridge physical activity index was used to classify past-year physical activity levels in occupational, leisure, and household domains (35). Validated country/center-specific dietary questionnaires were used to obtain information on dietary intake. Different types of dietary questionnaires were used in each study center, including semiquantitative food frequency questionnaires (FFQ) with or without an estimation of individual average portion size and diet history questionnaires combining a FFQ and 7-day dietary recalls (30, 31).

Metabolically defined body size phenotype definitions

Concentrations of C-peptide among the control population were used to define metabolic health status. Individuals were classified as MH if below the first tertile (Supplementary Table S2) or metabolically unhealthy (MU) if above the first tertile. This definition of metabolic health was derived given that the risk of endometrial cancer was elevated in women in the 2nd and 3rd tertiles of C-peptide compared with those in the 1st tertile (Supplementary Table S3). In addition, the same procedure was performed using quartiles (1st quartile as MH) and median values (<median as MH) of C-peptide standardized concentration amongst the control population (Supplementary Table S2).

These metabolic health definitions were then combined with NW (BMI<25 kg/m² or WC< 80 cm or WHR< 0.8) and OW (BMI≥25 kg/m² or WC≥ 80 cm or WHR≥ 0.8) status, generating four phenotype groups for each of the three anthropometric measures separately (in total 12 groups; 4×3): MH/NM; MH/OW; MU/NW; and MU/OW. The WC and WHR cutoff points were based on those from the International Diabetes Federation (36), which are gender and ethnic-specific cutoff points for European populations.

Statistical analysis

Descriptive analyses were performed and differences between cases and controls were assessed using paired sample t test for continuous variables and paired χ^2 test for categorical variables. Descriptive analyses were also performed between metabolically defined body size phenotype groups among the controls. As C-peptide was measured in two phases (in 2007 and then in 2019), standardized values were used in the analysis. The standardization was done by phase of the measurements, with all features following the reduced, centered normal distribution (mean $= 0$ and $SD = 1$). Partial Pearson correlations in the control group adjusted for batch and age at blood collection, between levels of C-peptide and anthropometrics variables were computed (Supplementary Table S4). Conditional logistic regression, stratified by case–control set, was used to compute odds ratios (OR) and 95% confidence intervals (CI) for the associations between metabolically defined body size phenotypes and endometrial cancer. The MH/NW was used as the reference category. The basic model was built on matching factors only, whereas the adjusted model was built on matching factors and a list of known risk factors for endometrial cancer that can potentially act as confounders, including: age at menopause (age at menopause $\langle 50; \geq 50$ years; missing), age at menarche (continuous), parity (0; 1; 2; >2; missing), hormone use (yes; no; missing), physical activity index (inactive; moderately inactive; moderately active; active; missing), smoking status (never; former smoker and current smoker; unknown), educational level (primary/no schooling; technical/professional/secondary and longer education; missing), total energy intake (continuous), alcohol intake (continuous), height (continuous), and diabetes (yes; no; missing). A separate model, including only OW participants and with the MU/OW category as reference, was also run. As sensitivity analyses, all models were rerun using the phenotypes defined on the basis of quartiles or on median level of C-peptide cutoff points. Also, analyses were repeated considering only the upper tertile as MU. Sensitivity analyses were also performed among postmenopausal women only; among non-exogenous hormone users only; among fasting participants only; among endometrial cancer type 1 only (defined by histology as explained in case ascertainment section); and among individuals from phase 2 only (as explained in laboratory measurements section). Furthermore, sensitivity analyses were conducted excluding cases diagnosed within the first 2 years of follow-up and their matched controls and excluding participants with diabetes. Statistical tests used in the analysis were all two-sided, and a P value of <0.05 was considered statistically significant. Analyses were conducted using SAS software.

Data availability

EPIC data and biospecimens are available for investigators who seek to answer important questions on health and disease in the context of research projects that are consistent with the legal and ethical standard practices of IARC/WHO and the EPIC Centers. The primary responsibility for accessing the data belongs to IARC and the EPIC centers. Access to materials from the EPIC study can be requested by contacting epic@iarc.fr.

Results

The current analysis used data from 1,634 women who were included in a nested case–control study with available C-peptide levels. A total of 817 women were classified as incident endometrial cancer cases and 817 were classified as matched controls. Among the cases, a total of 728 women were classified as type 1, 40 women were classified as type 2, and 49 women had unknown tumor type.

Table 1 shows that endometrial cancer cases had older age at menopause, but younger age at first menstrual period and lower number of full-term pregnancies than the controls. Endometrial cancer cases also had higher levels of C-peptide and greater BMI and WC than controls. In line with this, a higher proportion of control participants were classified as MH/NW and MH/OW compared with cases considering all anthropometric cutoff points. The baseline characteristics of the control group participants by metabolically defined body size phenotypes are shown in Table 2. Compared with the MH/NW group and considering the BMI classification, a greater proportion of MU/NW control participants reported having longer

Table 1. Baseline characteristics of participants in a nested case-control study within EPIC.

Abbreviations: BMI, body mass index; HRT, hormone replacement therapy; NA, Not applicable because was used as a matching factor; WC, waist circumference; WHR, waist-to-hip ratio.

^aPaired sample t test for continuous variable and paired χ^2 test for categorical variables.

^bAmong parous women.

^cMetabolically healthy/normal weight (BMI < 25 kg/m² or waist circumference <80 cm or waist-to-hip ratio <0.8) plus below tertile 1 of C-peptide. ^dMetabolically healthy/overweight (BMI ≥ 25 kg/m², or waist circumference ≥80 cm or waist-to-hip ratio ≥0.8), plus below tertile 1 of C-peptide. e Metabolically unhealthy/normal weight (BMI < 25 kg/m² or waist circumference <80 cm or waist-to-hip ratio <0.8), plus above tertile 1 of C-peptide. ^fMetabolically unhealthy/overweight (BMI ≥ 25 kg/m², or waist circumference ≥80 cm or Waist-to-hip ratio ≥0.8), plus above tertile 1 of C-peptide. ^gMedian (Interquartile range) among controls: 1.57 (1.05-2.32) and cases: 1.75 (1.16-2.64).

^hMedian (Interquartile range) among controls: 2.5 (0.3-10.8) and cases: 2.1 (0.2-9.3).

Table 2. Baseline characteristics of control group participants by metabolic health (hyperinsulinemia)-defined body size phenotypes using anthropometric cutoff points in the EPIC. Table 2. Baseline characteristics of control group participants by metabolic health (hyperinsulinemia)–defined body size phenotypes using anthropometric cutoff points in the EPIC.

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(%). cAmong parous women. NW, normal weight; OW/OB, overweight and obesity.

⁴Metabolically healthy/normal weight (BMI < 25 kg/m² or Waist circumference <80 cm or Waist-to-hip ratio <0.8) plus below tertile 1 of C-peptide.
"Metabolically healthy/overweight (BMI ≥ 25 kg/m², or Waist circumfe

education, higher alcohol intake, and greater prevalence of current smoking and was less frequently classified as physically active. In contrast with this, control participants in the MU/OW group (considering the BMI classification) were less likely to be current smokers and to have longer education, reported lower alcoholic intake and were more frequently classified as physically active than MH/OW. It is important to note that around 40% of the controls were classified in the MU/OW group whereas only around 11% were classified in the MH/ OW group. The results based onWC andWHR were broadly similar to those based on BMI.

The results for the associations between metabolically defined body size phenotypes and endometrial cancer risk when adjusted for potential cofounders are described below by the phenotype categories (Table 3).

MH/OW

When using BMI and WHR cutoff points, participants classified as MH/OW were at a higher risk of endometrial cancer compared with MH/NW participants, albeit the associations were not statistically significant (OR_{BMI} , 1.40; 95% CI, 0.91-2.15 and OR_{WHR} , 1.17; 95% CI, 0.75–1.81) and were at a statistically significant lower risk of endometrial cancer than their MU/OW counterparts (ORBMI, 0.44; 95% CI, 0.26-0.74 and OR_{WHR}, 0.43; 95% CI, 0.25–0.76). In contrast, when using WC cutoff points, MH/OW women were at statistically significant higher risk of endometrial cancer compared with MH/NW participants (OR, 1.94; 95% CI, 1.24–3.04) and they were at lower risk of endometrial cancer compared with the MU/OW (OR, 0.80; 95% CI, 0.49–1.31), although the association was not statistically significant.

MU/NW

MU/NW were at statistically significant higher risk of endometrial cancer than their MH/NW counterparts when using WC (OR, 1.48; 95% CI, 1.05–2.10) and WHR (OR, 1.68; 95% CI, 1.21–2.35) cutoff points, whereas the results for the BMI cutoff points were nonsignificant (OR, 1.16; 95% CI, 0.82–1.64).

MU/OW

MU/OW participants were at statistically significantly higher risk of endometrial cancer compared with MH/NW participants considering BMI (OR, 2.38, 95% CI, 1.73–3.27), WC (OR, 2.69; 95% CI, 1.92–3.77), and WHR (OR, 1.83; 95% CI, 1.32–2.54) cutoff points.

Sensitivity analyses

Similar results were observed when excluding cases diagnosed within the first 2 years of follow-up, excluding individuals with diabetes, as well as when the analyses were restricted to individuals with type 1 endometrial cancer or restricted to phase 2 samples (Supplementary Table S5). The results restricted to non-exogenous hormone users and to fasting subjects were also broadly similar; however, most of the results were not statistically significant due to the reduced sample size (Supplementary Table S5). Exclusion of premenopausal participants did not lead to substantial changes in the study results for BMI cutoff points, but a few changes were observed for WC and WHR cutoff points (Supplementary Table S5). Sensitivity analyses also showed similar results when using C-peptide quartiles and median cutoff points to define the metabolic health body size phenotypes (Supplementary Table S6). In addition, results defining the

Table 3. Risk of endometrial cancer incidence associated with metabolic health-defined body size phenotypes using anthropometric and C-peptide tertile cutoff points in the EPIC.

Note: In bold, we highlight the results that were statistically significant. Sub-sample analyses are also presented in this table. Values are OR (95% CI). BMI, Body Mass Index; WC, waist circumference, WHR, Waist-to-Hip ratio. Basic model was conditioned on matching factors only. Adjusted model was conditioned on matching factors, with additional adjustment for age at menopause, age at menarche, parity, hormone use, physical activity index, smoking status, educational level, alcohol intake, height, energy intake and diabetes; P_{trend}

^aMetabolically healthy/normal weight (BMI < 25 kg/m² or waist circumference <80 cm or waist-to-hip ratio <0.8) plus below tertile 1 of C-peptide.

bMetabolically healthy/overweight (BMI ≥ 25 kg/m², or waist circumference ≥80 cm or waist-to-hip ratio ≥0.8), plus below tertile 1 of C-peptide.

^cMetabolically unhealthy/normal weight (BMI < 25 kg/m² or waist circumference <80 cm or waist-to-hip ratio <0.8), plus above tertile 1 of C-peptide. ^dMetabolically unhealthy/overweight (BMI ≥ 25 kg/m², or waist circumference ≥80 cm or waist-to-hip ratio ≥0.8), plus above tertile 1 of C-peptide.

upper tertile as the MU group mirrored the main findings (Supplementary Table S7).

Discussion

In this prospective analysis of metabolic health and endometrial cancer risk, MU/NW and MU/OW participants, defined by C-peptide levels, were at higher endometrial cancer risk compared with MH/NW women. In addition, MH/OW women were at higher endometrial cancer risk compared with MH/NW women. These results indicate that women with higher levels of insulin are at elevated risk of endometrial cancer regardless of their body size; however, being OW raises endometrial cancer risk regardless of insulin profile.

Many, but not all, prior studies have shown a similar pattern of results for the relationships of metabolically defined body size phenotypes with cardiovascular disease, type 2 diabetes, all-cause mortality, open-angle glaucoma and obesity-related cancers (17–26, 28, 37, 38). Our results lend further support to the notion that, even though higher body size metrics are associated with increased endometrial cancer risk, the assessment of metabolic dysfunction regardless of body size may be an additional tool for risk stratification. Importantly, the study showed that NW women with metabolic dysfunction have elevated risk for endometrial cancer. The potential mechanisms underlying this relationship may involve the direct effect of insulin on normal endometrial and malignant cells, as the insulin receptor is commonly expressed in the tumor cells (39). However, multiple other factors may occur downstream of insulin signaling to impact endometrial tumorigenesis, such as chronic inflammation and sex hormone disruption (10, 15, 16, 40).

The factors influencing the development of metabolic dysfunction have been investigated and several hypotheses have been proposed, including differences in body fat distribution, poor diet and physical inactivity, and chronic inflammation (21, 41–43). It has been suggested that individuals with metabolic dysfunction tend to have higher intakes of sugar, sugar-sweetened beverages, and saturated fat as well as lower intakes of fruits, whole grains, and protein from vegetable sources compared with MH individuals (21). On the other hand, MH individuals tend to spend more time in moderate to vigorous physical activities and less time in sedentary activities compared with MU individuals (41, 44). Adipose tissue biology and function, including the genetic determinants of body fat distribution, depot-specific fat metabolism, adipose tissue plasticity and, particularly, adipogenesis also play a role (42). However, more research is needed to better understand the mechanisms underlying the development of metabolic dysfunction, including the potential role of the gut microbiota (42).

In the current analysis, individuals with OW or obesity, regardless of their metabolic health status, were at elevated endometrial cancer risk compared with MH/NW individuals. This is in line with previous results from the EPIC cohort showing that obesity (including higher WC and WHR) was associated with higher endometrial cancer risk compared with NW individuals (4). The results for the WC-specific cutoff point were stronger and more consistent compared with the other anthropometric cutoff points. These findings suggest that greater abdominal fat accumulation may impact endometrial cancer risk irrespective of insulin levels. A potential pathway underlying this relationship may include higher levels of estrogen that are synthesized with greater abdominal fat in both premenopausal (45) and postmenopausal women (46), given that higher exposure to unopposed estrogen is an established risk factor for endometrial cancer (47–50). Adipocyte hypertrophy– and hyperplasia-stimulated pro-inflammatory immune response, chronic fibrosis, and vascular inflammation are also potential mechanisms that create a microenvironment conducive to carcinogenesis (47, 51).

To our knowledge, this is the first investigation of metabolically defined body size phenotypes based on C-peptide levels and endometrial cancer risk in a prospective cohort setting. The long-term follow-up and high number of incident endometrial cancer cases recorded is a major strength of this study. However, some limitations of the current study should also be considered. First, although there is no universal definition of "metabolic health," the analysis used only C-peptide levels as a marker of metabolic health whereas there are more than 30 other possible definitions that have been used in different studies, including homeostatic model assessment of insulin resistance (HOMA-IR; using insulin and glucose measures; refs. 21, 43). C-peptide may be a better indicator for long-term insulin secretion than measuring insulin levels owing to its longer halflife (52). In the current study, hyperinsulinemia was defined on the basis of tertiles of C-peptide level in controls, which was supported by the results for the association between C-peptide tertiles and endometrial cancer risk showing elevated risk for the upper two tertiles. This methodology has also been used in previous EPIC studies classifying individuals according to their metabolically defined body sized phenotypes (17). Furthermore, analyses that used quartiles and median of C-peptide levels showed a similar pattern of results. However, future studies should aim to define clinically relevant cutoff points for normal C-peptide levels, which can potentially be used for stratification for endometrial cancer risk. Finally, results from the current study are largely applicable to white European women and future studies should investigate other populations, such as black women who tend to have worse prognosis from endometrial cancer (53, 54).

In conclusion, we have shown that women with metabolic dysfunction appear to have higher risk of endometrial cancer regardless of their body size. Therefore, it is possible that using only anthropometric measurements to identify women at higher risk of endometrial cancer would exclude normal-weight individuals with poor metabolic health and could underestimate the risk among OW individuals with hyperinsulinemia. MU/NW women represented 20% to 30% of the current sample; therefore, this proportion of women would be missed when using only body size for identifying women at higher risk of endometrial cancer. Thus, classifying populations by metabolically defined body size phenotypes may be of greater utility in identifying individuals at higher risk for endometrial cancer who would not have otherwise been identified solely by anthropometric measures. Our findings also showed that OW status may raise endometrial cancer risk even among women with lower insulin levels, suggesting that obesityrelated pathways are important for this cancer beyond insulin. The combination of anthropometric measures with metabolic parameters, such as C-peptide, may allow more precise identification of the strata of the population at greater endometrial cancer risk, which could be targeted for prevention strategies.

Authors' Disclosures

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Authors' Contributions

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References

- 1. Sung H, Ferlay J, Siegel RL, Laversanne M, Soerjomataram I, Jemal A, et al. Global cancer statistics 2020: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. CA Cancer J Clin 2021;71:209–49.
- 2. World Cancer Research Fund/American Institute for Cancer Research. Diet, nutrition, physical activity and cancer: a global perspective. Continuous Update Project Expert Report 2018. Available from: [dietandcancerreport.org.](dietandcancerreport.org)
- 3. Zhang Y, Liu H, Yang S, Zhang J, Qian L, Chen X. Overweight, obesity and endometrial cancer risk: results from a systematic review and meta-analysis. Int J Biol Markers 2014;29:e21–9.
- 4. Friedenreich C, Cust A, Lahmann PH, Steindorf K, Boutron-Ruault MC, Clavel-Chapelon F, et al. Anthropometric factors and risk of endometrial cancer: the European prospective investigation into cancer and nutrition. Cancer Causes Control 2007;18:399–413.
- 5. Nead KT, Sharp SJ, Thompson DJ, Painter JN, Savage DB, Semple RK, et al. Evidence of a causal association between insulinemia and endometrial cancer: a mendelian randomization analysis. J Natl Cancer Inst 2015;107:djv178.
- 6. Renehan AG, Tyson M, Egger M, Heller RF, Zwahlen M. Body-mass index and incidence of cancer: a systematic review and meta-analysis of prospective observational studies. Lancet 2008;371:569–78.
- 7. Aune D, Navarro Rosenblatt DA, Chan DSM, Vingeliene S, Abar L, Vieira AR, et al. Anthropometric factors and endometrial cancer risk: a systematic review and dose–response meta-analysis of prospective studies. Ann Oncol 2015;26: 1635–48.
- 8. Omiyale W, Allen NE, Sweetland S. Body size, body composition and endometrial cancer risk among postmenopausal women in Biobank. Int J Cancer 2020;147:2405–15.
- 9. Bianchini F, Kaaks R, Vainio H. Overweight, obesity, and cancer risk. Lancet Oncol 2002;3:565–74.
- 10. Dossus L, Lukanova A, Rinaldi S, Allen N, Cust AE, Becker S, et al. Hormonal, metabolic, and inflammatory profiles and endometrial cancer risk within the EPIC Cohort—a factor analysis. Am J Epidemiol 2013;177:787–99.
- 11. Trabert B, Eldridge RC, Pfeiffer RM, Shiels MS, Kemp TJ, Guillemette C, et al. Prediagnostic circulating inflammation markers and endometrial cancer risk in the prostate, lung, colorectal and ovarian cancer (PLCO) screening trial: pre-

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diagnostic inflammation markers and endometrial cancer. Int J Cancer 2017;140: 600–10.

- 12. Hernandez AV, Pasupuleti V, Benites-Zapata VA, Thota P, Deshpande A, Perez-Lopez FR. Insulin resistance and endometrial cancer risk: a systematic review and meta-analysis. Eur J Cancer 2015;51:2747–58.
- 13. Gunter MJ, Hoover DR, Yu H, Wassertheil-Smoller S, Manson JE, Li J, et al. A prospective evaluation of insulin and insulin-like growth factor-i as risk factors for endometrial cancer. Cancer Epidemiol Biomarkers Prev 2008;17:921–9.
- 14. Cust AE, Allen NE, Rinaldi S, Dossus L, Friedenreich C, Olsen A, et al. Serum levels of C-peptide, IGFBP-1 and IGFBP-2 and endometrial cancer ris: results from the European prospective investigation into cancer and nutrition. Int J Cancer 2007;120:2656–64.
- 15. Mu N, Zhu Y, Wang Y, Zhang H, Xue F. Insulin resistance: a significant risk factor of endometrial cancer. Gynecol Oncol 2012;125:751–7.
- 16. Nagamani M, Stuart CA. Specific binding and growth-promoting activity of insulin in endometrial cancer cells in culture. Am J Obstet Gynecol 1998;179: 6–12.
- 17. Murphy N, Cross AJ, Abubakar M, Jenab M, Aleksandrova K, Boutron-Ruault MC, et al. A nested case–control study of metabolically defined body size phenotypes and risk of colorectal cancer in the European Prospective Investigation into Cancer and Nutrition (EPIC). PLoS Med 2016;13:e1001988.
- 18. Gunter MJ, Xie X, Xue X, Kabat GC, Rohan TE, Wassertheil-Smoller S, et al. Breast cancer risk in metabolically healthy but overweight postmenopausal women. Cancer Res 2015;75:270–4.
- 19. Ogorodnikova AD, Kim M, McGinn AP, Muntner P, Khan U, Wildman RP. Incident cardiovascular disease events in metabolically benign obese individuals. Obesity 2012;20:651–9.
- 20. Moore LL, Chadid S, Singer MR, Kreger BE, Denis GV. Metabolic health reduces risk of obesity-related cancer in framingham study adults. Cancer Epidemiol Biomarkers Prev 2014;23:2057–65.
- 21. Smith GI, Mittendorfer B, Klein S. Metabolically healthy obesity: facts and fantasies. J Clin Invest 2019;129:3978–89.
- 22. Park YM, White AJ, Nichols HB, O'Brien KM, Weinberg CR, Sandler DP. The association between metabolic health, obesity phenotype and the risk of breast cancer. Int J Cancer 2017;140:2657–66.

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- 23. Dobson R, Burgess MI, Sprung VS, Irwin A, Hamer M, Jones J, et al. Metabolically healthy and unhealthy obesity: differential effects on myocardial function according to metabolic syndrome, rather than obesity. Int J Obes 2016;40:153–61.
- 24. Kim JW, Ahn ST, Oh MM, Moon DG, Cheon J, Han K, et al. Increased incidence of bladder cancer with metabolically unhealthy status: analysis from the National Health Checkup database in Korea. Sci Rep 2020;10:6476.
- 25. Kim JW, Ahn ST, Oh MM, Moon DG, Han K, Park HS. Incidence of prostate cancer according to metabolic health status: a Nationwide Cohort Study. J Korean Med Sci 2019;34:e49.
- 26. Jung Y, Han K, Park HYL, Lee SH, Park CK. Metabolic health, obesity, and the risk of developing open-angle glaucoma: metabolically healthy obese patients versus metabolically unhealthy but normal weight patients. Diabetes Metab J 2020;44:414.
- 27. Tomiyama AJ, Hunger JM, Nguyen-Cuu J, Wells C. Misclassification of cardiometabolic health when using body mass index categories in NHANES 2005– 2012. Int J Obes 2016;40:883–6.
- 28. Chung HS, Lee JS, Song E, Kim JA, Roh E, Yu JH, et al. Effect of metabolic health and obesity phenotype on the risk of pancreatic cancer: a nationwide populationbased cohort study. Cancer Epidemiol Biomarkers Prev 2021;30:521–8.
- 29. Trabert B, Wentzensen N, Felix AS, Yang HP, Sherman ME, Brinton LA. Metabolic syndrome and risk of endometrial cancer in the United States: a study in the SEER–Medicare Linked Database. Cancer Epidemiol Biomarkers Prev 2015;24:261–7.
- 30. Slimani N, Kaaks R, Ferrari P, Casagrande C, Clavel-Chapelon F, Lotze G, et al. European Prospective Investigation into Cancer and Nutrition (EPIC) calibration study: rationale, design and population characteristics. Public Health Nutr 2002;5:1125–45.
- 31. Riboli E, Hunt KJ, Slimani N, Ferrari P, Norat T, Fahey M, et al. European Prospective Investigation into Cancer and Nutrition (EPIC): study populations and data collection. Public Health Nutr 2002;5:1113–24.
- 32. Dossus L, Kouloura E, Biessy C, Viallon V, Siskos AP, Dimou N, et al. Prospective analysis of circulating metabolites and endometrial cancer risk. Gynecol Oncol 2021;162:475–81.
- 33. Spencer EA, Appleby PN, Davey GK, Key TJ. Validity of self-reported height and weight in 4808 EPIC–Oxford participants. Public Health Nutr 2002;5:561–5.
- 34. Skeie G, Borch K, Mode N, Henningsen M. Validity of self-reported body mass index among middle-aged participants in the Norwegian Women and Cancer study. Clin Epidemiol 2015;7:313–23.
- 35. Wareham NJ, Jakes RW, Rennie KL, Schuit J, Mitchell J, Hennings S, et al. Validity and repeatability of a simple index derived from the short physical activity questionnaire used in the European Prospective Investigation into Cancer and Nutrition (EPIC) study. Public Health Nutr 2003;6:407–13.
- 36. Alberti KGM, Zimmet P, Shaw J. The metabolic syndrome—a new worldwide definition. Lancet 2005;366:1059–62.
- 37. Kabat GC, Kim MY, Stefanick M, Ho GYF, Lane DS, Odegaard AO, et al. Metabolic obesity phenotypes and risk of colorectal cancer in postmenopausal women: obesity phenotypes and risk of colorectal cancer. Int J Cancer 2018;143: 543–51.
- 38. Akinyemiju T, Moore JX, Pisu M, Judd SE, Goodman M, Shikany JM, et al. A prospective study of obesity, metabolic health, and cancer mortality: metabolically healthy obesity and cancer. Obesity 2018;26:193–201.
- 39. Perry RJ, Shulman GI. Mechanistic links between obesity, insulin, and cancer. Trends Cancer 2020;6:75–8.
- 40. Zhang AMY, Wellberg EA, Kopp JL, Johnson JD. Hyperinsulinemia in obesity, inflammation, and cancer. Diabetes Metab J 2021;45:285–311.
- 41. Conus F, Allison DB, Rabasa-Lhoret R, St-Onge M, St-Pierre DH, Tremblay-Lebeau A, et al. Metabolic and behavioral characteristics of metabolically obese but normal-weight women. J Clin Endocrinol Metab 2004;89:5013–20.
- 42. Iacobini C, Pugliese G, Blasetti Fantauzzi C, Federici M, Menini S. Metabolically healthy versus metabolically unhealthy obesity. Metabolism 2019;92:51–60.
- 43. Brandão I, Martins MJ, Monteiro R. Metabolically healthy obesityheterogeneity in definitions and unconventional factors. Metabolites 2020;10:48.
- 44. Klitgaard HB, Kilbak JH, Nozawa EA, Seidel AV, Magkos F. Physiological and lifestyle traits of metabolic dysfunction in the absence of obesity. Curr Diab Rep 2020;20:17.
- 45. Hetemäki N, Mikkola TS, Tikkanen MJ, Wang F, Hämäläinen E, Turpeinen U, et al. Adipose tissue estrogen production and metabolism in premenopausal women. J Steroid Biochem Mol Biol 2021;209:105849.
- 46. Hetemäki N, Savolainen-Peltonen H, Tikkanen MJ, Wang F, Paatela H, Hämäläinen E, et al. Estrogen metabolism in abdominal subcutaneous and visceral adipose tissue in postmenopausal women. J Clin Endocrinol Metab 2017; 102:4588–95.
- 47. Donohoe CL, Doyle SL, Reynolds JV. Visceral adiposity, insulin resistance and cancer risk. Diabetol Metab Syndr 2011;3:12.
- 48. Hemsell DL, Grodin JM, Brenner PF, Siiteri PK, Macdonald PC. Plasma precursors of estrogen. II. Correlation of the extent of conversion of plasma androstenedione to estrone with age. J Clin Endocrinol Metab 1974;38:476–9.
- 49. Kleinman D, Karas M, Danilenko M, Arbell A, Roberts CT, LeRoith D, et al. Stimulation of endometrial cancer cell growth by tamoxifen is associated with increased insulin-like growth factor (IGF)-I induced tyrosine phosphorylation and reduction in IGF binding proteins. Endocrinology 1996;137:1089–95.
- 50. Rodriguez AC, Blanchard Z, Maurer KA, Gertz J. Estrogen signaling in endometrial cancer: a key oncogenic pathway with several open questions. Horm Cancer 2019;10:51–63.
- 51. Calle EE, Kaaks R. Overweight, obesity and cancer: epidemiological evidence and proposed mechanisms. Nat Rev Cancer 2004;4:579–91.
- 52. Bonser AM, Garcia-Webb P, Harrison LC. C-peptide measurement: methods and clinical utility. Crit Rev Clin Lab Sci 1984;19:297–352.
- 53. Cote ML, Alhajj T, Ruterbusch JJ, Bernstein L, Brinton LA, Blot WJ, et al. Risk factors for endometrial cancer in black and white women: a pooled analysis from the epidemiology of endometrial cancer consortium (E2C2). Cancer Causes Control 2015;26:287–96.
- 54. Jamison PM, Noone A-M, Ries LAG, Lee NC, Edwards BK. Trends in endometrial cancer incidence by race and histology with a correction for the prevalence of hysterectomy, SEER 1992 to 2008. Cancer Epidemiol Biomarkers Prev 2013;22:233–41.