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Evaluating metformin strategies for cancer prevention: a target trial emulation using electronic health records

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ABSTRACT

Background: Metformin users appear to have a substantially lower risk of cancer than nonusers in many observational studies. These inverse associations may be explained by common flaws in observational analyses that can be avoided by explicitly emulating a target trial.

Methods: We emulated target trials of metformin therapy and cancer risk using populationbased linked electronic health records from the UK (2009-2016). We included individuals with diabetes, no history of cancer, no recent prescription for metformin or other glucose-lowering medication, and hemoglobin A1C (HbA1c) <64 mmol/mol (<8.0%). Outcomes included total cancer and 4 site-specific cancers (breast, colorectal, lung, prostate). We estimated risks using pooled logistic regression with adjustment for risk factors via inverse-probability weighting. We emulated a second target trial among individuals regardless of diabetes status. We compared our estimates with those obtained using previously applied analytic approaches. **Methods:** We emulated target trials of metformin therapy and cancer risk using population-
oased linked electronic health records from the UK (2009-2016). We included individuals with
tiabless, no history of cancer, no re

Results: Among individuals with diabetes, the estimated 6-year risk differences (metformin – no metformin) were -0.2% (95% CI: -1.6%, 1.3%) in the intention-to-treat analysis and 0.0% (95% CI: -2.1%, 2.3%) in the per-protocol analysis. The corresponding estimates for all site-specific cancers were close to zero. Among individuals regardless of diabetes status, these estimates were also close to zero and more precise. By contrast, previous analytic approaches yielded estimates that appeared strongly protective.

Conclusions: Our findings are consistent with the hypothesis that metformin therapy does not meaningfully influence cancer incidence. The findings highlight the importance of explicitly emulating a target trial to reduce bias in the effect estimates derived from observational analyses.

INTRODUCTION

Observational studies suggest that users of metformin, a first-line treatment for diabetes, have a substantially lower risk of cancer compared with nonusers.¹⁻¹⁰ The prospect of reducing cancer risk with a safe and affordable medication such as metformin is very appealing. However, secondary analyses of randomized trials in diabetes prevention suggest that metformin does not have a cancer-protective effect. The effect estimates from randomized trials are imprecise, and thus difficult to interpret conclusively, because they are based on a relatively small number of cases of total cancer. 11

Evaluating metformin for the prevention of site-specific cancers using randomized trials may not be feasible given the large sample size and long follow-up that would be required. Observational datasets, such as the ones available in electronic health records, can be used to explicitly emulate (hypothetical) target trials that address these limitations. However, the use of observational databases requires adequate emulation procedures, including the comparison of clinically realistic treatment strategies, the avoidance of biases related to mishandling of time zero of follow-up (selection bias and immortal time bias), and sufficient adjustment for confounding for treatment initiation.^{12,13} accordary analyses of randomized trials in diabetes prevention suggest that metformin does not
awe a cancer-protective effect. The effect estimates from randomized trials are imprecise, and
thus difficult to interpret conc

Selection bias, due to the inclusion of prevalent users, and immortal time bias, due to the use of postbaseline treatment information to assign treatment groups, can be eliminated by a sound emulation of the target trial, as described in detail previously.^{12,14} Confounding can be reduced, for example, by restricting the analysis to individuals with indications for treatment initiation if these indications are strong risk factors for the outcome of interest. Specifically, if diabetes (an indication for metformin initiation) were a risk factor for cancer, the observational analysis would restrict eligibility to individuals with diabetes to adjust for confounding by

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diabetes; otherwise, no restriction to individuals with diabetes would be necessary even though the prevalence of diabetes is expected to be much higher among individuals who receive metformin than among those who do not receive it.

In this study, we used a large database of linked electronic health records from primary care, hospitalizations, and mortality registrations to emulate target trials of clinically relevant strategies of metformin therapy for the prevention of total and site-specific cancer. We conducted separate analyses among individuals with type 2 diabetes and among individuals regardless of diabetes status.

METHODS

Specification of the target trials

We designed this observational analysis to emulate target trials (i.e., hypothetical pragmatic trials that would have answered the causal questions of interest) of metformin as compared with no metformin for the prevention of cancer. The key protocol components of these target trials are summarized in **Table 1**.

Eligibility criteria for the target trial among individuals with diabetes include age ≥ 30 years between April 1, 2009 and February 29, 2016, diagnosis of type 2 diabetes, no history of cancer (except nonmelanoma skin cancer), no metformin contraindication (hepatic or renal impairment or lactic acidosis), HbA1c <64 mmol/mol (<8.0%), no prescription for metformin or other glucose-lowering medication within the past year, at least 1 year of up-to-standard data in a Clinical Practice Research Database (CPRD) general practice (defined as high-quality data deemed suitable for use in research¹⁵), and at least 1 year of potential follow-up, as well as known HbA1c measured within the past year and known smoking and body-mass index measured within the past 4 years. Baseline is defined as the first month in which all eligibility care, hospitalizations, and mortality registrations to emulate target trials of clinically relevant
trategies of metformin therapy for the prevention of total and site-specific cancer. We conduct
exparate analyses among i

criteria are met. The target trial among individuals regardless of diabetes status has the same eligibility criteria except for type 2 diabetes and otherwise shares the same protocol.

The dynamic strategies to be compared are (1) initiation of metformin therapy at baseline and continuation over follow-up until the development of a contraindication (hepatic or renal impairment or lactic acidosis) or cancer diagnosis and (2) no initiation of metformin therapy over follow-up until the development of an indication (HbA1c \geq 64 mmol/mol [\geq 8.0%]). When clinically warranted during the follow-up (i.e., upon the development of these conditions), individuals and their clinicians would decide whether to start, stop, or switch therapy. These are clinically relevant strategies, in contrast to the static strategies evaluated in previous observational studies under which individuals were not allowed to deviate from their assigned treatment strategy when clinically appropriate. $16,17$ mpairment or lactic acidosis) or cancer diagnosis and (2) no initiation of metformin therapy overlollow-up until the development of an indication (HbA1c \geq 64 mmol/mol $[\geq$ 8.0%]). When
chinically warranted during the f

The outcomes of interest are incident total cancer and the 4 most common site-specific invasive cancers in this population: female breast, colorectal, lung (non-small cell), and prostate. Previous validation studies have confirmed 95% of cancers recorded in this database.¹⁸

For each eligible individual, follow-up starts at treatment assignment (baseline) and ends upon the outcome of interest, death, loss to follow-up (transfer out of the practice, or incomplete follow-up [2 years after the last recorded lab prognostic factors or 4 years after the last recorded lifestyle prognostic factors]), 6 years after baseline, or the administrative end of follow-up (end of practice data collection or February 29, 2016), whichever happens first.

The causal estimands of interest are the intention-to-treat effect of being assigned to the treatment strategies and the per-protocol effect of adhering to them.

In the intention-to-treat analysis, risks (cumulative incidences) can be estimated nonparametrically via the Kaplan–Meier estimator or parametrically via a pooled logistic regression model for the monthly probability of the outcome that includes an indicator of assigned strategy, a flexible function of months since randomization (linear and quadratic terms), and a product term between the treatment indicator and time. The predicted values from this model are used to estimate 6-year cancer risks under each strategy. If the model also needs to include baseline covariates, the risks will be standardized to the distribution of the baseline covariates (see **eMethods 1 http://links.lww.com/EDE/C36** for details). The same model without the product term can be used to approximate the hazard ratio (for comparison with estimates from previous studies) because the monthly risk of the outcome is low.¹⁹

In the per-protocol analysis, this pooled logistic regression model is fit to the data after censoring individuals if and when they deviate from their assigned treatment strategy. Specifically, individuals in the initiator group are censored when they stop metformin (unless they develop a contraindication or cancer) and individuals in the non-initiator group are censored when they start metformin (unless they develop an indication). To adjust for factors associated with adherence, time-varying inverse-probability weights are estimated via a pooled logistic regression model for the monthly probability of treatment that includes baseline and timevarying factors. After the development of one of the above conditions, the weights for adherence remain constant until the end of follow-up. Estimated weights are truncated at their 99th percentile to prevent outliers from having an undue influence on the analyses. nclude baseline covariates, the risks will be standardized to the distribution of the baseline
covariates (see **eMethods 1 <u>http://links.lww.com/EDE/C36</u>** for details). The same model
without the product term can be used t

Nonparametric bootstrapping with 500 samples can be used to calculate percentile-based 95% confidence intervals for risk estimates, and robust variances can be used to calculate conservative 95% confidence intervals for hazard ratio estimates.

To identify potential subgroups of individuals for whom the treatment strategies may be most beneficial, analyses are conducted separately in subsets of the eligible population defined at baseline according to age (≤ 70 vs. ≥ 70 years), sex (male vs. female), and, in the target trial among individuals with diabetes, time since diabetes diagnosis ≤ 1 vs. ≥ 1 year).

Emulation of the target trials

We explicitly emulated the target trials described above using observational data from the CPRD, Hospital Episode Statistics, and Office of National Statistics. These population-based datasets are comprised of longitudinal UK electronic health records from primary care consultations, admitted hospitalization episodes, and death registrations for approximately 15 million individuals, accessed through the CALIBER resource.^{15,20} Longitudinal primary care data on demographics, lifestyle factors, symptoms, diagnoses, clinical examination findings, laboratory test results, referrals, and prescriptions were recorded by general practitioners in the CPRD. Hospitalization data were obtained through linkage with Hospital Episode Statistics. Mortality data were obtained through linkage with the Office of National Statistics. Disease phenotypes were derived using algorithms that combine information on diagnoses, symptoms, laboratory values, physiologic measures, prescriptions, and procedures, which were created and validated using an established methodology.^{21,22} CPRD. Hospital Episode Statistics, and Office of National Statistics. These population-based
datasets are comprised of longitudinal UK electronic health records from primary care
consultations, admitted hospitalization epi

We used the observational data to emulate each protocol component of the target trials as closely as possible (**Table 1**). We classified individuals into 1 of 2 treatment groups according to their prescription records at baseline and assumed these groups were exchangeable at baseline conditional on the covariates in **Table 2**. The analysis to estimate the observational analogues of the intention-to-treat and per-protocol effects proceeded as for the target trials, with adjustment for these baseline covariates to emulate randomization (and incorporation of their time-varying values into the inverse-probability weights for the per-protocol analysis, as in the target trial) and with sequential emulation for statistical efficiency (see **eMethods 1**

<http://links.lww.com/EDE/C36>).

Specifically, we emulated the target trial as a sequence of trials²³⁻²⁵ starting at each of the 71 months between April 2009 and February 2015. This accommodates the fact that individuals may meet the eligibility criteria at several times over follow-up and is more statistically efficient than choosing just one of those times as time zero.²⁶ Separately for each of the 71 months, eligible individuals were classified into a treatment group and followed until the outcome of interest, death, loss to follow-up, 6 years after baseline, or the administrative end of follow-up, whichever happened first. We then conducted a pooled analysis over all 71 emulated trials and estimated the observational analogues of intention-to-treat and per-protocol effects.

Sensitivity analyses

We performed several sensitivity analyses to address potential misclassification, residual confounding, and selection bias. Specifically, we (1) increased the maximum gap between successive prescriptions from 30 to 60 days, (2) additionally adjusted for practice region (at the Strategic Health Authority level), family history of cancer, cancer screening in the past year, and influenza vaccination in the past year (a marker of health care seeking behavior) as potential confounders, (3) truncated weights at their 99.5th percentile, and (4) additionally applied weights for censoring due to loss to follow-up. We also allowed individuals with diabetes to discontinue metformin upon the initiation of insulin therapy. To explore the potential influence of reverse causation, we lagged treatment values by 6 months. may meet the eligibility criteria at several times over follow-up and is more statistically efficien
han choosing just one of those times as time zero.²⁶ Separately for each of the 71 months,
eligible individuals were cl

It might be argued that the protective effect of metformin reported by previous observational analyses is not metformin-specific but the result of glycemic control more generally. We therefore emulated a third target trial of intensification to metformin–sulfonylurea dual therapy (a second-line diabetes treatment that adds an oral hypoglycemic medication to metformin) vs. continuation of metformin monotherapy and cancer incidence, among individuals receiving metformin monotherapy for diabetes (see **eMethods 2<http://links.lww.com/EDE/C36>** for details). The eligibility criteria are the same as those described above, except they require current use of metformin therapy and additionally require HbA1c \geq 48 mmol/mol (\geq 6.5%), an indication that treatment intensification from metformin monotherapy to dual therapy may be needed. Given these eligibility criteria, all individuals in this comparison of first- and second-line treatments are expected to have a similar stage and severity of diabetes.

Conventional analyses

Some of the estimates from previous observational studies may be partly explained by 3 types of deviations from target trial emulation: (1) mishandling of time zero by comparing everusers vs. never-users of metformin therapy over the follow-up, $27,28$ (2) comparison of unrealistic (static) strategies of metformin vs. no metformin therapy, regardless of clinical indications for stopping or starting treatment, 17 and (3) failure to apply the same eligibility criteria to all treatment groups under study.17,29 The latter occurred when comparing initiators of metformin monotherapy who had received no prescription for any glucose-lowering medication in the past 6 months (predominantly individuals diagnosed with diabetes recently) vs. initiators of metformin– sulfonylurea dual therapy who had received metformin monotherapy but no prescription for other glucose-lowering medications (predominantly individuals who were diagnosed with diabetes some time ago and for whom metformin was insufficient).¹⁷ Previous studies subsequently censored individuals in each treatment group when they switched from their baseline treatment.¹⁷ We replicated each of these analytic decisions in our own data among individuals with diabetes. current use of metformin therapy and additionally require HbA1c \geq 48 mmol/mol (\leq 6.5%), an andication that treatment intensification from metformin monotherapy to dual therapy may be needed. Given these eligibility

Ethical approval

The CPRD has been granted generic ethical approval for observational studies that make use of only anonymized data and linked anonymized National Health Service healthcare data (Multiple Research Ethics Committee ref. 05/MRE04/87). This study was approved by the Medicines and Healthcare Products Regulatory Agency Independent Scientific Advisory Committee (protocol 16_221) and the Harvard T.H. Chan School of Public Health Institutional Review Board.

RESULTS

Figure 1 shows a flowchart of patient selection, and **Table 2** shows the baseline characteristics of the 44,237 individuals eligible for the emulated trial among individuals with type 2 diabetes, and the 216,785 individuals eligible for emulated trial among individuals regardless of diabetes status. Compared with metformin non-initiators at baseline, metformin initiators were, on average, younger, had higher HbA1c and body-mass index, and included a higher proportion of individuals with any recent specialist referral. Among individuals with diabetes, metformin initiators also had a shorter time since diabetes diagnosis. Among individuals regardless of diabetes status (who satisfied the requirement for a recent HbA1c measurement, among the other eligibility criteria), 89% of metformin initiators and 34% of metformin non-initiators had type 2 diabetes (data not tabulated). Medicines and Healthcare Products Regulatory Agency Independent Scientific Advisory
Committee (protocol 16_221) and the Harvard T.H. Chan School of Public Health Institutional
Review Board.
AESULTS
Figure 1 shows a flo

In the emulated trial among individuals with diabetes, 2,777 individuals developed cancer, including 272 female breast, 365 colorectal, 368 lung, and 416 prostate cancers, over the 6-year follow-up (median 3.3 years, interquartile range 2.0-4.9 years). In the emulated trial among individuals regardless of diabetes status, 7,507 individuals developed cancer, including

821 female breast, 934 colorectal, 1,001 lung, and 1,214 prostate cancers, over the 6-year followup (median 2.1 years, interquartile range 1.3-2.6 years).

Table 3 shows the estimated 6-year risks for cancer comparing metformin with no metformin. In the emulated trial among individuals with diabetes, the estimated observational analogue of the intention-to-treat 6-year risk difference was -0.2% (95% CI: -1.6%, 1.3%) for total cancer, and ranged from -0.4% to 0.7% across cancer sites. The estimated observational analogue of the per-protocol 6-year risk difference was 0.0% (95% CI: -2.1%, 2.3%) for total cancer, and ranged from -0.2% to 1.6% across cancer sites. Estimates were similar in the emulated trial among individuals regardless of diabetes status, but confidence intervals were narrower. Risk curves under each strategy were almost overlapping (**Figure 2**). Estimates for total cancer were similar (1) in subgroups defined at baseline according to age, sex, and time since diabetes diagnosis (**eTable 2 http://links.lww.com/EDE/C36**), (2) under several sensitivity analyses for potential misclassification, residual confounding, and selection bias due to loss to follow-up (**eTables 3-6 http://links.lww.com/EDE/C36**), (3) when allowing individuals with diabetes to discontinue metformin upon the initiation of insulin therapy (**eTable 7 http://links.lww.com/EDE/C36**), (4) when lagging treatment values by 6 months (intentionto-treat hazard ratio 1.08 among individuals with diabetes and among individuals regardless of diabetes status), and (5) when only adjusting for age (among individuals with diabetes: intentionto-treat hazard ratio 1.01, per-protocol hazard ratio 1.01; among individuals regardless of diabetes status: intention-to-treat hazard ratio 1.05, per-protocol hazard ratio 1.04); and identical when additionally adjusting for use of other glucose-lowering medications. analogue of the intention-to-treat 6-year risk difference was -0.2% (95% CE-1.6%, 1.3%) for
cotal cancer, and ranged from -0.4% to 0.7% across cancer sites. The estimated observational
analogue of the per-protocol 6-year r

In the emulated trial among individuals receiving metformin monotherapy for diabetes, the estimated effect of treatment intensification to metformin–sulfonylurea dual therapy vs.

continuation of metformin monotherapy was also near null (intention-to-treat hazard ratio for total cancer 1.03, 95% CI: 0.71, 1.50; per-protocol hazard ratio for total cancer 0.89, 95% CI: 0.45, 1.75; see **eMethods 2<http://links.lww.com/EDE/C36>**).

Conventional analyses

In analyses that replicated the analytic approaches of some previous observational studies in our data among individuals with diabetes, estimates were near null when we compared initiators of metformin monotherapy vs. initiators of metformin–sulfonylurea dual therapy, identified by applying different eligibility criteria to each treatment group (hazard ratio for total cancer 1.17, 95% CI: 0.92, 1.48). When we compared static treatment strategies of metformin vs. no metformin therapy, estimates for total cancer were near null (hazard ratio 0.97, 95% CI: 0.84, 1.13), but estimates for lung cancer were further from the null (hazard ratio 0.64, 95% CI: 0.39, 1.05) than in the primary analysis (**eTable 8 http://links.lww.com/EDE/C36**). Analyses that compared ever-users vs. never-users of metformin therapy over the follow-up also resulted in strong inverse associations (hazard ratio for total cancer 0.54, 95% CI: 0.49, 0.60) (data not tabulated). We found a similar pattern when comparing ever-users vs. never-users of sulfonylureas (hazard ratio for total cancer 0.67, 95% CI: 0.57, 0.79). In analyses that replicated the analytic approaches of some previous observational studient our data among individuals with diabetes, estimates were near null when we compared mitiators of metformin monotherapy vs. initiat

DISCUSSION

After emulating target trials using the electronic health records of 216,785 individuals, we found little indication that metformin therapy influences cancer incidence over the study period, regardless of whether eligibility was restricted to having diabetes. These findings are consistent with secondary analyses of randomized trials in diabetes prevention,¹¹ but inconsistent with previous observational studies that have reported a substantially lower risk of cancer among users of metformin compared with nonusers. $5,6,8,9$

The approach of explicitly specifying the protocol of the target trial and its observational emulation prevents common biases in observational analyses. Specifically, the extreme, apparently beneficial, effect estimates from previous observational studies may be partly explained by mishandling of time zero and by the comparison of unrealistic static strategies.

Unhitching eligibility assessment and treatment assignment from time zero can lead to substantial selection bias and immortal time bias, as previously discussed.^{12,30-33} When we replicated this flaw by classifying individuals according to their observed treatment use over follow-up via a comparison of ever-users vs. never-users of metformin, we obtained an apparently protective estimate for cancer of an implausible magnitude (hazard ratio 0.54).

In the real world, treatment strategies are dynamic. Static strategies are unrealistic because they require that individuals continue taking treatment even after the onset of contraindications or toxicity. As a result, using real world data to compare static strategies can lead to positivity violations and bias. When we replicated this flaw by comparing static strategies of metformin vs. no metformin over follow-up, we found a more "protective" estimate for lung cancer than in our analyses comparing more realistic dynamic strategies (hazard ratio 0.64 vs. 0.75). Further exploration of this issue was limited by the imprecision of our estimates for lung cancer. Unhitching eligibility assessment and treatment assignment from time zero can lead to
substantial selection bias and immortal time bias, as previously discussed,^{12,30,25} When we
eplicated this flaw by classifying individ

Following the basic principles of study design can avoid time-related biases in observational studies. Indeed, a systematic review showed that previous observational studies identified as least likely to be affected by these biases also suggested no effect of metformin on cancer risk, as in the present study.⁷ The proposed target trial approach can be viewed as a guide to implement sound principles of causal inference and study design, 34 as well as a way to estimate appropriately adjusted measures of absolute risk and to evaluate clinically realistic

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dynamic treatment strategies. Our study had additional strengths. The electronic health records capture rich longitudinal data on demographic and clinical features that allowed us to characterize individuals with high resolution and adjust for many potential confounders. Unlike some previous studies on this topic, we were able to distinguish type 1 from type 2 diabetes, adjust for time since diabetes diagnosis and HbA1c to minimize potential confounding by disease duration and severity, and incorporate HbA1c into our eligibility criteria and treatment strategies to reduce potential confounding as well as positivity violations. Also, our approach to emulate a sequence of target trials is more statistically efficient than emulating a single target trial.²⁶

Our study also had some potential limitations. First, as in any observational analysis, assignment to a treatment strategy was not randomized. If the 2 treatment groups had different distributions of risk factors, then the effect estimates would be confounded. However, much less confounding by indication is expected when evaluating unintended effects (e.g., cancer outcomes) vs. intended effects (e.g., coronary heart disease, death)³⁵; the treatment groups were similar at baseline with respect to their demographic characteristics and medical history; and we adjusted for many potential baseline and time-varying confounders. Some unmeasured variables that may be imbalanced between the treatment groups are diet and physical activity (e.g., metformin non-initiators may have achieved diabetes control via improved diet and increased exercise) in the diabetes-only analysis and non-diabetes indications for metformin (e.g., polycystic ovary syndrome) in the general analysis. Smoking is a strong risk factor for lung cancer that was coarsely measured (as never, former, or current smoker); however, estimates for lung cancer were similar when only adjusting for age, suggesting a potentially limited role for confounding in this setting (intention-to-treat hazard ratio among individuals with diabetes 1.00 adjust for time since diabetes diagnosis and HbA1c to minimize potential confounding by
disease duration and severity, and incorporate HbA1c into our eligibility criteria and treatment
strategies to reduce potential confou

vs. 1.00 in the primary analysis, among individuals regardless of diabetes status 0.99 vs. 0.93 in the primary analysis). Second, we were limited by our reliance on prescription records and diagnosis codes, which may contribute to measurement error and residual confounding. However, previous validation studies have confirmed a high proportion of recorded cancers $(95%)$ and other diagnoses in our study data.^{18,36} Third, the length of follow-up may have been insufficient to capture slowly progressing cancers. However, previous observational studies with comparable or shorter follow-up have reported a substantially lower risk of total and site-specific cancers, 37 including prostate cancer, 38 among metformin users.

Most previous studies of metformin and cancer were restricted to individuals with diabetes, an indication for metformin initiation.⁸ This restriction protects against bias that would arise if diabetes were a risk factor for cancer, conditional on the other measured clinical features. In the present study, estimates were similar regardless of whether eligibility was restricted to this indication though, as expected, estimates were more precise when not making this restriction. The choice of eligibility criteria when emulating any target trial using observational data will be guided by these considerations about the comparability of the treatment groups. When evaluating intended effects of treatment (e.g., statins and risk of death), confounding by indication may be a larger concern, and it may therefore be important to restrict eligibility to individuals with an indication for treatment (e.g., coronary heart disease). When evaluating unintended effects of treatment, as in the present study, confounding by indication may be a smaller concern³⁵ and lower variance may be achieved by omitting the indication from the eligibility criteria. Note that, to adjust for potential confounding by glycemic status, we required a recent measure of HbA1c as an eligibility criterion, even for individuals without diabetes, which increased the proportion 95%) and other diagnoses in our study data, ^{18,36} Third, the length of follow-up may have been
asufficient to capture slowly progressing cancers. However, previous observational studies with
comparable or shorter followof individuals in our study who had a reason to have their HbA1c assessed (e.g., a cardiometabolic disorder).

In summary, our findings suggest that metformin therapy does not meaningfully influence cancer incidence over 6 years. Our explicit emulation of a target trial helped to reduce bias that may contribute to discrepancies between the effect estimates derived from observational analyses and randomized trials. Our analysis also highlights how more precise effect estimates may be obtained by omitting the indication for treatment from the eligibility criteria in cases where this restriction is not necessary to achieve comparability between the 2 treatment groups, as will often be the case in observational analyses that evaluate unintended effects of medications. bias that may contribute to discrepancies between the effect estimates derived from observation
analyses and randomized trials. Our analysis also highlights how more precise effect estimates
any be obtained by omitting the

AUTHOR CONTRIBUTIONS

B.A.D., X.G.-A., S.D. and M.A.H. conceived the overall study. All authors contributed to the design and analysis. B.A.D. analyzed the data. R.W.L. provided key input in processing data from the database. All authors contributed to the interpretation of the results. B.A.D. wrote the first draft of the manuscript, and all authors reviewed, revised, and approved the final version of the manuscript.

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FIGURE LEGENDS

Figure 1. Selection and flow of eligible individuals when emulating a target trial of metformin therapy and cancer risk (a) among individuals regardless of diabetes status and (b) among individuals with diabetes, 2009-2016. Panel B shows the flow of individuals after applying the additional eligibility criterion of type 2 diabetes. Numbers in parentheses represent unique individuals in each group. Counts of initiator and non-initiator individuals do not sum to the total number of eligible individuals because some eligible individuals contributed to both groups in different nested trials.

Figure 2. Estimated risk of cancer by metformin therapy among individuals with diabetes (observational analogue to an intention-to-treat [a] and per-protocol [b] analysis), and among individuals regardless of diabetes status (observational analogue to an intention-to-treat [c] and per-protocol [d] analysis), using linked electronic health records from Clinical Practice Research Datalink, Hospital Episode Statistics, and Office of National Statistics, 2009-2016. Shaded areas represent pointwise 95% confidence intervals. Motitional eligibility criterion of type 2 diabetes. Numbers in parentheses represent unique
andividuals in each group. Counts of initiator and non-initiator individuals do not sum to the tota
number of eligible individual

Table 1. Specification and emulation of pragmatic target trials of metformin therapy and cancer risk using linked electronic health records from Clinical Practice Research Datalink, Hospital Episode Statistics, and Office of National **Statistics**

Table 1. Specification and emulation of pragmatic target trials of metformin therapy and cancer risk using linked electronic health records from Clinical Practice Research Datalink, Hospital Episode Statistics, and Office of National Statistics.

Abbreviations: eGFR, estimated glomerular filtration rate; HbA1c, hemoglobin A1c; MDRD; Modification of Diet in Renal Disease.

a eGFR (mL/min/1.73m²) = 175 * (serum creatinine [μmol/L]/88.4)^{-1.154} * age^{-0.203} * 0.742 (if female) * 1.210 (if Black). We note that alternative prediction algorithms (e.g., the CKD-EPI 2021 eGFR creatinine equation) have been more recently defined.

Table 2. Baseline characteristics of eligible individuals with type 2 diabetes mellitus and regardless of diabetes status when emulating target trials of metformin therapy and cancer risk using linked electronic health records from Clinical Practice Research Datalink, Hospital Episode Statistics, and Office of National Statistics, 2009-2015^a .

Table 2. Baseline characteristics of eligible individuals with type 2 diabetes mellitus and regardless of diabetes status when emulating target trials of metformin therapy and cancer risk using linked electronic health records from Clinical Practice Research Datalink, Hospital Episode Statistics, and Office of National Statistics, 2009-2015^a .

a Baseline ranges from April 2009 to February 2015. Phenotype definitions are available at https://www.caliberresearch.org/

b Each individual may contribute to more than 1 emulated trial.

^c Other cardiovascular disease includes acute rheumatic fever, chronic rheumatic heart disease, pulmonary heart disease, and other circulatory disease.

^d Antihypertensive use includes all primary care prescriptions from British National Formulary chapters 2.2.1 thiazides and related diuretics, 2.2.3 potassium-sparing diuretics and aldosterone antagonists, 2.2.4 potassiumsparing diuretics with other diuretics, 2.4 beta-adrenoceptor blocking drugs, 2.5 hypertension and heart failure, 2.6.2 calcium-channel blockers. er cardiovascular disease includes acute rheumaic fever, chronic rheumaic heart disease, pulmonary happenesses.

And chere friculationy disease.

Hypertensive use includes all primary care prescriptions from British Nation

Table 3. Estimated 6-year standardized risks and hazard ratios^a for cancer comparing metformin therapy with no metformin therapy, using linked electronic health records from Clinical Practice Research Datalink, Hospital Episode Statistics, and Office of National Statistics, 2009-2016.

Abbreviation: CI, confidence interval.

a Adjusted for age, sex, body-mass index, smoking status, hemoglobin A1c, months since last measure of hemoglobin A1c, coronary heart disease, hypertension, cerebrovascular disease, other cardiovascular disease, antihypertensive use, aspirin use, nonsteroidal anti-inflammatory drug use, any specialist referral in the past 3 months. Estimates for breast and colorectal cancer additionally adjusted for hormone replacement therapy and oral contraceptive use. The emulated trial among individuals with diabetes additionally adjusted for time since diabetes diagnosis. Estimated risk differences were standardized to the joint distribution of the baseline covariates.

^b The number of events in the initiator and non-initiator groups do not sum to the total number of events because some individuals contributed as events to both groups in different nested emulated trials. The number of events is lower in the per-protocol analysis because of the censoring under this approach (see Methods). humber of events in the initiator and non-initiator groups do not sum to the total number of events becausing metromin netistion at baseline with no metromin initiation at baseline.
Actions metromin initiation at baseline

^c Comparing metformin initiation at baseline with no metformin initiation at baseline.

^d Comparing metformin initiation at baseline and continuation over follow-up until the development of a contraindication or cancer with no metformin initiation over follow-up until the development of an indication.

Figure 1a

a Among individuals regardless of diabetes status

Figure 1b

b Among individuals with diabetes

Among individuals with diabetes