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Prognostic models for predicting recurrence and survival in women with endometrial cancer (Protocol)

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[Prognosis Protocol]

Prognostic models for predicting recurrence and survival in women with endometrial cancer

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ABSTRACT

Objectives

This is a protocol for a Cochrane Review (prognosis). The objectives are as follows:

To review all prognostic models that combine two or more clinical, histological or molecular variables, or a combination of these variables, to provide an individualised assessment of risk of recurrence or death from disease and evaluate their performance to predict these outcomes in people undergoing curative treatment for endometrial cancer.



BACKGROUND

Description of the health condition and context

Endometrial cancer, which is also referred to as womb or uterine cancer, is the most common gynaecological cancer in high-income countries. Endometrial cancer arises from the lining of the womb (endometrium). Over 90% of people with endometrial cancer are diagnosed following investigations for postmenopausal bleeding or heavy or irregular vaginal bleeding (Clarke 2018). It is estimated that 9% of those presenting with postmenopausal bleeding have an underlying endometrial cancer (Clarke 2018).

In the USA, more than 19 in every 100,000 women will be diagnosed with endometrial cancer each year. In comparison, only one to four per 100,000 women each year are diagnosed with this cancer in countries in Sub-Saharan Africa, the Middle East and South-Central Asia (Lortet-Tieulent 2018). Despite this variation, endometrial cancer is becoming more common throughout the world and has nearly overtaken cervical cancer to become the most common gynaecological cancer in middle-income countries (Bray 2018). This is thought to be due to the association of this cancer with excess body weight, later childbearing and smaller families. A small number of endometrial cancers (less than 5%) are caused by an inheritable genetic condition, known as Lynch syndrome. People with Lynch syndrome have gene alterations that increase their risk of bowel, ovarian and endometrial cancer. People with this gene change will have a 30% to 60% lifetime risk of endometrial cancer.

The outlook for most people diagnosed with endometrial cancer is good. The majority (70%) will present when the cancer is still confined to the womb (i.e. early-stage disease) and will be completely cured of their disease. Even when all stages of endometrial cancer are considered, nearly 80% of people affected can expect to survive for at least five years after their diagnosis, if they live in a high-income countries (ONS 2016).

Initial treatment for endometrial cancer usually involves surgery to remove the womb and other potentially affected organs. After surgery, these organs are examined by a pathologist to gain information about how far the cancer has spread (i.e. stage) and its type. This provides prognostic information, allowing an estimation to be made about how likely the cancer is to recur. Many of those affected will be cured completely by surgery alone and never require further treatment. However, approximately 10% to 15% of people with endometrial cancer will develop a recurrence (Bendifallah 2017; Salani 2017). The prognosis for those that develop recurrence is poor and most of these people will die from their disease (Bendifallah 2017). Being able to predict who these people might be offers the opportunity to prevent recurrence and potentially improve the rate of cure in this group. It also might allow us to identify those who are very unlikely to have recurrence and may be able to avoid additional treatment and its associated risks.

Additional treatment in the form of chemotherapy or radiotherapy (or both), sometimes referred to as adjuvant therapy, is offered to those deemed to be at higher risk of recurrence. This may include localised internal radiation applied to the top of the vagina (i.e. vaginal brachytherapy), radiation delivered from outside the body (i.e. external beam radiotherapy), drugs given to kill cancer cells (i.e. chemotherapy), or combinations of these. The most effective way of delivering this additional therapy, and the sequence in which it should be delivered, remains unclear (de Boer 2018; Galaal 2014; Matei 2019) All treatments are associated with potential side-effects, which may limit the tolerability of the treatment and cause long-term treatment-related side effects such as bowel and bladder problems, loss of sensation and vaginal discomfort (Ao 2020; de Boer 2016). Determining the optimal selection strategy will hopefully reduce under- and over-treatment in those diagnosed with endometrial cancer.

Additionally, being able to predict the risk of recurrence in people treated for endometrial cancer may allow for the tailoring of follow-up. In the UK, the British Gynaecological Cancer Society recommends that people who have completely recovered physically after their treatments and who are deemed at the lowest risk of recurrence (less than 10% risk of recurrence) can be empowered to contact their cancer care provider if they notice vaginal bleeding, new lumps or bumps or swelling of their abdomen, rather than attending routine follow-up appointments. This is known as patient-initiated follow-up (Newton 2020). This frees up time for healthcare providers to focus on providing individualised and holistic care for those who continue to have physical or psychological problems after treatment. Those at higher risk of recurrence can be offered either face-to-face or telephone follow-up before moving to patient-initiated follow-up two or more years post-treatment, as after this time the risk of recurrence declines (Newton 2020).

Description of the prognostic models

Rather than being a single disease, endometrial cancer should be thought of as a group of cancers that share a common origin, i.e. the womb lining. The different types of endometrial cancer are thought to develop through slightly different biological mechanisms (Hecht 2006; Piulats 2017). This may explain the differences seen in the disease courses and ultimately, why some people are more likely to develop a recurrence or die from their disease.

There are various models that try to group features of tumours and the people affected by them by their risk of developing a recurrence or dying from the disease. The most predictive individual feature is the extent of spread of the disease at the time of diagnosis (i.e. the stage). The International Federation of Gynecology and Obstetrics (FIGO) staging system is the most commonly used system to describe the extent of spread in endometrial cancer. It provides a common framework to describe spread of disease and is regularly updated to reflect the evolving understanding of the relative importance of different findings (Creasman 2009; Pecorelli 2009). The extent of spread (stage) is classified from 1 to 4, with 1 being the earliest stage and having the lowest likelihood of recurrence and death; and 4 being the most advanced stage, with the highest likelihood of recurrence and death. In the last update in 2009, some of the substages in early-stage endometrial cancer were removed, as no significant difference was seen in survival of people in these groups. Additionally, spread of the cancer to the glandular tissue of the cervix was downgraded to Stage 1 and the presence of free-floating cancer cells in the fluid surrounding the abdominal organs was no longer considered to provide additional information about the risk of recurrence.

Other recognised risk factors for recurrence include age, tumourrelated features, such as grade (i.e. how changed the cells appear), depth of invasion of the cancer into the muscle wall of the womb, lymphovascular space invasion (i.e. whether cancer cells



are seen within the blood and lymph vessels in the womb), site of the tumour within the womb, tumour subtype and tumour size (Singh 2019). These factors have been combined into a number of different risk classification systems to help determine who might benefit from adjuvant therapy (Table 1). Although there is significant overlap between those classified at highest risk within all of the classifications systems, the incorporation of different risk factors into each model can lead to different conclusions in some cases. Additionally, because the classification and grading of womb cancer are very subjective, this can leads to disagreement in approximately one-third of endometrial cancers designated as high risk (Gilks 2013; Guan 2011; Han 2013 Hoang 2013). The poor performance of prognostic models based predominantly on features noted at pathological examination may, therefore, result in over- or under-treatment in a significant proportion of people with endometrial cancer.

Data from the Cancer Genome Atlas project demonstrated that endometrial cancer could be sub-categorised into four groups based upon distinct gene-level (i.e. molecular) alterations (Levine 2013). The presence of these features allows pathologists to objectively identify people with tumours that are likely to recur and then to die of their disease. This has driven the development of molecular classification systems in endometrial cancer (Talhouk 2015) that have been shown to be reproducible in various settings (Bosse 2018; Britton 2019; Kommoss 2018; Talhouk 2017; Travaglino 2020).

This review aims to evaluate the comparative performance of all current prognostic models to predict recurrence and survival in those receiving curative treatment for endometrial cancer. It is hoped that the results of this review will be used to continue refining tools to guide whether people with this cancer are offered adjuvant therapy in addition to surgery, and influence decisions regarding follow-up following initial treatment.

Health outcomes

- Recurrence within five years of diagnosis
- Recurrence at any time after diagnosis
- Death due to endometrial cancer
- · Death from any cause after endometrial cancer

Why it is important to do this review?

With increasing numbers of people affected by endometrial cancer, and efforts to diagnose it at earlier stages, it will be increasingly important to provide accurate estimates of prognosis to guide treatment and follow-up decisions. The James Lind Alliance Womb Cancer Alliance project in 2016 identified that research to evaluate predictors of recurrence was amongst the top 30 research priorities for healthcare professionals and people affected by endometrial cancer (Wan 2016). It is important to evaluate whether the predictive performance of current risk models is adequate because these are widely used to guide treatment, and more recently to streamline follow-up. This topic is also particularly timely given the move towards clinical trials offering specific treatments to people with tumours that express particular molecular markers.

OBJECTIVES

To review all prognostic models that combine two or more clinical, histological or molecular variables, or a combination of these

variables, to provide an individualised assessment of risk of recurrence or death from disease and evaluate their performance to predict these outcomes in people undergoing curative treatment for endometrial cancer.

METHODS

Criteria for considering studies for this review

Please refer to Table 2 for the type of patients, interventions, comparators, outcomes, timings and settings (PICO) considered relevant to this review.

Types of studies

We will include all retrospective and prospective studies that evaluate the performance of prognostic models to predict recurrence or survival in people who have undergone curative treatment for endometrial cancer. We will only include studies where both surgical and histopathological features are used to describe the extent of spread of the disease (i.e. those published after the 1988 FIGO classification was introduced) to allow comparisons of this to other staging systems found within the searches. We will consider studies describing the development of a model with and without external validation. These studies will only be included if the model was applied to data that would be available following surgical staging but prior to adjuvant therapy and where there is follow-up of at least five years. Where external validation using an independent data set has not been applied, we will only include studies where there is evidence of appropriate adjustments made for over-optimism.

Targeted population

We will include studies in adults (18 years and over) with primary endometrial carcinomas of any histotype who have undergone surgery with curative intent. We will exclude studies or data pertaining to people with non-endometrial cancers that have metastasised to the body of the uterus, people with advanced endometrial cancers in whom surgical staging was not performed, people treated with chemotherapy, radiotherapy or hormonal therapy alone and people considered ineligible for surgery or who declined surgery.

Types of prognostic predictive model

We will assess all prognostic models that combine two or more clinical, histological or molecular variables, or a combination of these variables, and use a mathematical function to predict recurrence or survival in people with endometrial cancer. In this context, we define molecular variables as any recognised marker of gene function. These may include, but are not limited to: somatic or germline DNA variation, DNA methylation status, and protein expression.

Types of outcomes to be predicted

Primary outcomes

 Recurrence within five years of diagnosis: this outcome was chosen as the primary outcome as many of these models are used to determine the need for additional adjuvant therapy. In clinical practice, five years of follow-up is the current standard. Whilst survival may not be extended by adjuvant treatment (Johnson 2011; Kong 2012), the time in which the patient is free of cancer-related symptoms or requiring treatment (or both) is

important in facilitating shared decision making, determining the efficacy of new treatment strategies, and planning provision

- of healthcare services for those treated for endometrial cancer.
- Recurrence at any time following initial treatment: we acknowledge that although 80% of recurrences occur within the first three years after treatment, a small subgroup of people with endometrial cancer will develop later recurrences. These people may have disease characteristics that put them at higher risk of late recurrences. Predicting recurrence at any time point is therefore also important, but will be considered separately.
- All-cause mortality: we will consider all-cause mortality as this has implications on shared decision-making and planning of broader healthcare services.
- Cancer-specific mortality: we will consider cancer-specific mortality separately as this has implications for assessing the efficacy of cancer treatments and for planning cancer services.

Search methods for identification of studies

Electronic searches

We will perform electronic searches of the following databases and sources.

- Ovid MEDLINE
- Ovid Embase
- Web of Science

As no specific search filters exist to identify prognostic model studies, we will combine and modify search filters suggested by Geersing 2012 and Pace 2014 to formulate a search strategy to identify models that seek to predict recurrence or survival (or both) in people with endometrial cancer. We will ensure that this is returning appropriate results by screening the MEDLINE searches and adapting this as appropriate, before translating these searches for use in the other sources (see Appendix 1).

In 1988, FIGO staging switched from a clinical staging system to a system that incorporated both surgical and histopathological findings; therefore, we will only include studies published after 1988 and up to the present day to allow comparisons of this to other staging systems found within the searches. We will include studies published in all languages and where these are published in languages other than English, we will have these translated as appropriate to reduce language bias. For all included studies, we will check for retractions and corrections. If there are more than twelve months between our original search and publication of the review, we will perform a further search to check for new studies prior to submission.

Searching other resources

In addition to the electronic searches outlined above, we will cross-check the reference lists of the retrieved articles, relevant reviews and guidelines pertaining to the management of endometrial cancer published by national and international specialist societies such as, but not limited to, the Society of Gynecologic Oncology, the European Society of Gynaecological Oncology, the European Society of Medical Oncology. We will search grey literature including Open Grey, ProQuest Dissertations & Theses Global, Databases of ongoing trials that involve the metaRegister of Controlled Trials (mRCT) (www.isrctn.com/page/mrct) and ClinicalTrials.gov. We intend to contact authors of

relevant studies, experts, and investigators who are known to be active in the field for further information on ongoing studies.

Data collection

Selection of studies

The titles and abstracts of all potentially eligible studies identified through the searches will be screened in duplicate by teams of two independent review authors (AM, EB, BR, IES, YLW, TDJW). We will retrieve the full-text study reports of all records considered potentially eligible by any review author. The full reports of those studies identified at the initial screening will be independently reviewed by two review authors (YLW, IES). These review authors will make a recommendation regarding which of these studies should be included and record the reason for exclusion for any ineligible studies. A third review author (EC) will resolve any disagreements through discussion amongst all three review author group will be consulted. The decisions made regarding the studies for inclusion will be collated using the author-support tool, Covidence (Covidence 2017).

We will remove duplicates and collate multiple reports from the same study into a single record. We will document the total number of retrieved records and included and excluded studies using a flow diagram, as recommended by the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) statement (Page 2021) . We will also complete 'Characteristics of excluded studies' tables.

Data extraction and management

Data extraction will be based on the 'CHecklist for critical Appraisal and data extraction for systematic Reviews of prediction Modelling Studies' (CHARMS) checklist (Moons 2014). We will use a standardised data collection form to record the data extracted agreed upon by the study team; we will pilot this form on three included studies (see Appendix 2).

Assessment of risk of bias in included studies

We intend to classify each study into one of the three categories (model development with or without external validation in the same publication and external validation study of a previously developed model only); we will assess risk of bias for each developed or validated model and for each outcome predicted by the model. We will adapt the 'Prediction model Risk Of Bias ASsessment Tool' (PROBAST) to assess risk of bias for each developed or validated model, for each outcome (Wolff 2019). PROBAST is structured around the assessment of four key domains, as follows.

- Patient selection.
- Predictors.
- Outcome.
- Analysis.

Each domain is judged for risk of bias (low, high or unclear); signalling questions for each domain facilitate this (see Appendix 3). Signalling questions are rated as yes (Y), probably yes (PY), probably no (PN), no (N) or no information (NI). All signalling questions are phrased so that "yes" indicates the absence of bias. Any signalling question rated as "no" or "probably no" indicates the

potential for bias. Two review authors (IES, EB) will independently make a judgement to determine whether each domain should be rated as "high", "low" or "unclear" risk of bias.

Prognostic model evaluation is judged to be at low risk of bias if all domains were rated as being at low risk of bias or if the prediction model was based on a very large data set and included some form of internal validation. However, if a prognostic model was developed without any external validation, and it was rated as being at low risk of bias for all domains, we will consider categorising it as high risk of bias. If at least one domain is judged to be at high or unclear risk of bias, then prognostic model evaluation is judged to be at high or unclear risk of bias.

The first three domains are also rated for concerns regarding the applicability of the model to the research question (low/ high/unclear). If there are low concerns about applicability for all domains, the prediction model evaluation is judged to have low concerns otherwise; if at least one domain is judged to have high or unclear concerns, then prediction model evaluation is judged to have high or unclear concerns about applicability, respectively. We intend to resolve any discrepancy by discussion and involvement of a third independent review author to solve any disagreement. Should there be insufficient information to make a judgement, we intend to contact the corresponding authors via email to request additional information to be able to make sound judgements on the 'Risk of bias' assessment.

Measures of association or predictive performance measures to be extracted

We will follow the guidance for meta-analysing the predictive performance of prognostic models and will extract the performance measures with associated precision measures (standard error or confidence interval) for each studied model from the development and their corresponding validation studies (Debray 2017). Two main statistical measures will be extracted: discrimination and calibration. Discrimination refers to model's ability to distinguish between individuals who develop and do not develop the outcome. The concordance (C) statistic is the most commonly used discrimination measure. It ranges from 0.5 (no discriminative ability) to 1 (perfect discriminative ability) (Snell 2018). For survival models that assess time to recurrence or death, the 'D' statistic will be extracted. The D statistic is a discrimination measure that gives the log hazard ratio of a model's predicted risks, dichotomised at the median value. Calibration refers to a model's accuracy of predicted risk probabilities and indicates the extent to which expected outcomes (predicted from the model) and observed outcomes agree. We will extract the following calibration performance measures: calibration plots (expected probabilities plotted against observed outcome frequencies), calibration slopes and observed-to-expected ratios (O:E ratios, ratios between observed and expected numbers of events or outcome frequencies) (Moons 2015).

Unit of analysis issues

We will compare the institutions and year of inclusion of all validation cohorts belonging to a model to determine whether there is a potential overlap in participants. When the same cohort was used to validate two or more different prognostic models, we will consider them as separate studies.

Dealing with missing data

We intend to contact the corresponding authors of primary studies to provide information on missing data for our analyses. In case of any non-response, we can estimate performance measures and standard errors such as the O:E ratio and C statistic, using formulas described by Debray and colleagues (Debray 2017; Debray 2018). If data are still insufficient, we will conduct a sensitivity analysis to explore the impact of the missing data.

Assessment of heterogeneity

We will investigate potential sources of clinical and statistical heterogeneity based on extracted items mentioned in the section "Data extraction and data management", according to the CHARM checklist. We plan to inspect forest plots to assess between-study heterogeneity and calculate tau², l² and prediction intervals which provide a range for the potential performance in a new validation study (Riley 2011), although this may be very wide if less than 10 studies are included. We will also conduct subgroup analyses to explore potential sources of heterogeneity. Specific attention will be paid to the year of diagnosis, as FIGO staging for this condition changed in 2009 as discussed in the Background. Heterogeneity is expected to arise due to differences in treatment regimens undergone by the participants in the different studies.

Assessment of reporting deficiencies

The majority of studies reporting on prognostic models are not prospectively registered and do not have published protocols (Peat 2014); subsequently, assessment of potential publication bias is difficult. If a study protocol is available, we will check articles for protocol violations.

Calibration measures, especially calibration slope and plot, are also not commonly reported (Bouwmeester 2012) despite the existence of methodological and reporting recommendations, i.e. the TRIPOD statement (Transparent Reporting of a multivariable prediction model for Individual Prognosis or Diagnosis). This may limit the reliability and applicability of many studies (Bouwmeester 2012). Where available, we will report the calibration measures used, and will make it clear in the review when these were not available.

Data synthesis

Data synthesis and meta-analysis approaches

We will consider conducting a meta-analysis of a model's discrimination and calibration performance if it has been developed or validated at least three times for the outcome in question in independent data sets. We will pool only validation studies of the outcome that a model was developed for. We will perform Logit transformation for the C statistic and the log transformation for O:E ratio to improve normal between-study distribution (Snell 2018). If meta-analysis is performed, a random-effects approach is essential to allow for unexplained heterogeneity across studies. We will perform meta-analyses through methods proposed by Debray 2017, using the meta-analysis packages in the R statistics language, which include metafor and metamisc (metamisc 2020).

Where the C statistic is not reported, we will not estimate it. Should the C statistic be provided without measures of uncertainties, we will calculate the standard error and confidence intervals based



on the P value or the combination of sample size and the number of events, if available, according to the recommendations of Newcombe and colleagues (Debray 2017; Debray 2018; Newcombe 2006).

Subgroup analysis and investigation of heterogeneity

A subgroup analysis will be performed to investigate whether the change in FIGO staging over time has affected the performance of the models. We will conduct meta-regression, if needed, which extends a standard random-effects meta-analysis model by including study-level covariates. The dependent variable is the (transformed) estimate of the model performance measure (C statistic or O:E ratio) (Deeks 2021). Study-level or summarised patient-level data then will be used as explanatory or independent variables. Meta-regression explores the cause and extent of the between-study heterogeneity (Riley 2011). However, if the number of studies less than 10, the power to detect a genuine association with meta-regression will usually be low.

Sensitivity analysis

We will perform sensitivity analyses by excluding studies with high risk of bias for at least one domain, studies in people receiving adjuvant therapy, and studies with missing data.

Conclusions and summary of findings

We will use the definitions laid out by the GRADE Working Group and the GRADE checklist to evaluate levels of certainty (Meader 2014). This will be presented in the form of a tabulated summary of our findings, generated in GRADEpro GDT (GRADEpro GDT; Meader 2014; Mustafa 2013); see Appendix 4. Currently there is no GRADE assessment specifically for prognostic model studies. However, if GRADE guidelines for prognostic model studies are published before we start analysing our results, we will use that guidance instead.

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ADDITIONAL TABLES

Table 1. Commonly used models to classify risk of recurrence in endometrial cancer

Risk classification models in endometrial cancer	Reference	Prognostic factors included	Risk categories from lowest risk to highest

The International Federa- tion of Gynecology and Ob- stetrics (FIGO) staging	Creasman 2009	Extent of tumour spread, spread to regional lymph nodes, metastasis to distant sites	Stages 1-4	
Post Operative Radiation Therapy in Endometrial Car- cinoma (PORTEC)-1 classifi-	Creutzberg 2000	FIGO stage, tumour grade, histological type,	Low	
		extent of myometrial invasion, age > 60	Intermediate	
cation			High-intermediate	
			High	
Gynecologic Oncology	Keys 2004	FIGO stage, tumour grade, histological type, extent of myometrial invasion, LVSI, stratified age category (i.e. < 50, 50 to 69, > 70)	Low	
Group (GOG) 99 classifica- tion			Low-intermediate	
			High-intermediate	
			High	
Survival effect of para-	Todo 2010	FIGO stage, tumour grade, histological type	Low	
aortic lymphadenectomy (SEPAL) in Endometrial can-		(i.e. endometrioid or non-endometrioid), ex- tent of myometrial invasion, LVSI	Intermediate	
cer classification			High	
European Society of Med-	Colombo 2013	FIGO stage, tumour grade, histological type	Low	
ical Oncology (ESMO) classi- fication		(i.e. endometrioid or non-endometrioid), ex- tent of myometrial invasion	Intermediate	
			High	
ESMO-modified classifica- tion	Bendifallah 2014	FIGO stage, tumour grade, histological type	ESMO low risk/LVSI absent	
	Bendifallah 2014	(i.e. endometrioid or non-endometrioid), ex- tent of myometrial invasion, LVSI	ESMO low risk/LVSI present	
	Bendifallah 2014		ESMO intermediate risk/LVSI absent	
	Bendifallah 2014 Bendifallah 2014		ESMO intermediate risk/LVSI present	
	Bendifallah 2014		ESMO high risk/LVSI absent	
			ESMO high risk/LVSI present	
The Cancer Genome Atlas	Levine 2013	Ultramutation of POLE proofreading gene,	POLE hypermutated	
classification	Levine 2013	microstatellite instability, DNA copy number	Microsatellite instability hy-	
	Levine 2013		permutated	
	Levine 2013		Copy number low	
			Copy number high	
TransPORTEC classification	Stelloo 2015	Ultramutation of POLE proofreading gene, microstatellite instability, p53 gene mutation	POLE proofreading-mutant tumours	
	Stelloo 2015		Microsatellite instable tu-	
	Stelloo 2015		mours	
			No specific molecular profile (NSMP) tumours	

Table 1. Commonly used models to classify risk of recurrence in endometrial cancer (Continued)

Table 1. Commonly used models to classify risk of recurrence in endometrial cancer (Continued)

			p53-mutant tumours
Proactive Molecular Risk Classifier for Endometrial Cancer (ProMisE) classifica- tion	Talhouk 2017	Ultramutation of POLE proofreading gene, MMR IHC, p53 IHC	POLE mutant
	Talhouk 2017		MMR IHC abnormal
			p53 wild type
			p53 abnormal

LVSI - lymphovascular space invasion, POLE - DNA polymerase epsilon, MMR IHC - mismatch repair immunohistochemistry p53 IHC - p53 immunohistochemistry

Characteristics	Details of what will be considered
Population	Women with a new diagnosis of primary endometrial cancer
Intervention	All prognostic models that combine at least two clinical, histological and/or molecular variables to predict recurrence or survival in the population
Comparator	_
Outcomes	Recurrence at any site primarily and secondarily, cancer-specific survival
Timing	Baseline: moment when need for adjuvant therapy decided, i.e. directly following staging surgery but prior to adjuvant therapy
	Outcome: moment when recurrence diagnosed or death from endometrial cancer occurs
Setting	Risk stratification of women for adjuvant therapy in any setting

Table 2. PICO question to be considered

APPENDICES

Appendix 1. MEDLINE Ovid search strategy

1. exp Endometrial Neoplasms/

2. ((endomet* or uter* or womb*) adj4 (cancer* or neoplas* or carcinom* or adenocarcinom* or malignan* or tumor* or tumour*)).ti,ab. 3. 1 or 2

4. validat\$.mp. or predict\$.ti. or rule\$.mp.

5. (predict\$ and (outcome\$ or risk\$ or models\$)).mp.

6. ((history or variable\$ or criteria or scor\$ or characteristic\$ or finding\$ or factor\$) and (predict\$ or model\$ or decision\$ or identif\$ or prognos\$)).mp.

7. decision\$.mp. and ((model\$ or clinical\$).mp. or LOGISTIC MODELS/)

8. (prognostic and (history or variable\$ or criteria\$ or scor\$ or characteristic\$ or finding\$ or factor\$ or model\$)).mp.

9. exp ROC Curve/

10. ("stratification" or "ROC Curve" or "discrimination" or "discriminate" or "c-statistic" or "c statistic" or "Area under the curve" or AUC or "calibration" or "indices" or "algorithm" or "multivariable").mp.

11. ((risk* or molecular or prognostic) adj3 (stratification or classification)).ti,ab.

12. ((clinical* or histopathological* or histological* or clinicopathological) adj5 (feature* or classification* or grading* or model*)).ti,ab.

13. (Post Operative Radiation Therapy in Endometrial Carcinoma or PORTEC or Gynecologic Oncology Group 99 classification or GOG 99 classification or "survival effect of para-aortic lymphadenectomy" or SEPAL or ESMO* or "ESMO-modified" or Cancer Genome Atlas Classification or TGCA* or TransPORTEC* or ProMisE*).mp.

14. 4 or 5 or 6 or 7 or 8 or 9 or 10 or 11 or 12 or 13

15. exp Recurrence/



16. exp Neoplasm Recurrence, Local/

- 17. (recurren* or progress* or relapse* or repeat* or return*).ti,ab.18. (prevent* or limit* or control*).ti,ab.
- 19. 17 and 18

20. surveillance.ti,ab.

21. 15 or 16 or 19 or 20

22. 3 and 14 and 21

Appendix 2. CHARM data extraction checklist

Domain	Key items				
Source of data	(e.g. cohort, case-control, randomised trial participants, or registry data)				
Participants	Participant eligibility and recruitment method (e.g., consecutive participants, location, number of centers, setting, inclusion and exclusion criteria)				
	Participant description				
	Details of treatments received, if relevant				
	Study dates (e.g. recruitment period, follow up period, length of follow-up)				
Outcome(s) to be predicted	Definition and method for measurement of outcome				
	Was the same outcome definition (and method for measurement) used in all patients?				
	Type of outcome (e.g. single or combined endpoints)				
	Was the outcome assessed without knowledge of the candidate predictors (i.e. blinded)?				
	Were candidate predictors part of the outcome (e.g. in panel or consensus diagnosis)?				
	Time of outcome occurrence or summary of duration of follow-up				
Model Development	Modelling method (e.g. logistic, survival, neural network, or machine learning techniques)				
	Modelling assumptions satisfied				
	Method for selection of predictors for inclusion in multivariable modelling (e.g. all candidate pre- dictors, pre-selection based on unadjusted association with the outcome)				
	Method for selection of predictors during multivariable modelling (e.g. full model approach, back- ward or forward selection) and criteria used (e.g. P value, Akaike Information Criterion)				
Model Performance	Calibration (calibration plot, calibration slope, Hosmer-Lemeshow test) and Discrimination (C-statistic, D-statistic, log-rank) measures with confidence intervals				
	Classification measures (e.g. sensitivity, specificity, predictive values, net reclassification improvement) and whether a-priori cut points were used				
Model Evaluation	Method used for testing model performance: development dataset only (random split of data, re- sampling methods e.g. bootstrap or cross-validation, none) or separate external validation (e.g. temporal, geographical, different setting, different investigators)				
	In case of poor validation, whether model was adjusted or updated (e.g. intercept recalibrated, predictor effects adjusted, or new predictors added)				

(Continued)				
Results	Final and other multivariable models (e.g. basic, extended, simplified) presented, including predic- tor weights or regression coefficients, intercept, baseline survival, model performance measures (with standard errors or confidence intervals)			
	Any alternative presentation of the final prediction models, e.g. sum score, nomogram, score chart, predictions for specific risk subgroups with performance			
	Comparison of the distribution of predictors (including missing data) for development and valida- tion datasets			
Sample size	Number of participants and number of outcomes/events			
	Number of outcomes/events in relation to the number of candidate predictors (Events Per Vari- able)			
Missing data	Number of participants with any missing value (include predictors and outcomes)			
	Number of participants with missing data for each model			
	Handling of missing data			
Interpretation and discussion	Interpretation of presented models			
	Comparison with other studies, discussion of generalisability, strengths and limitations.			

Appendix 3. 'Risk of bias' signaling questions for each domain

- Patient selection
 - * What study design was used?
 - * Was the inclusion and exclusion of participants appropriate?
 - * Was participant selection similar to the development study?
- Predictors
 - * Were predictors defined and assessed in a similar way for all participants?
 - * Were predictor assessments made without knowledge of outcome data?
 - * Are all predictors available at the time the model is intended to be used?
- Outcome
 - * Was the outcome determined appropriately?
 - * Was a pre-specified or standard outcome definition used?
 - * Were predictors excluded from the outcome definition?
 - * Was the outcome defined and determined in a similar way for all participants?
 - * Was the outcome determined without knowledge of predictor information?
 - * Was the time interval between predictor assessment and outcome determination appropriate?



- Analysis
 - * Were there a reasonable number of participants with the outcome?
 - * Were there a reasonable number of participants with the outcome?
 - * Were continuous and categorical predictors handled appropriately?
 - * Were all enrolled participants included in the analysis?
 - * Were participants with missing data handled appropriately?
 - * Was selection of predictors based on univariable analysis avoided?
 - * Were complexities in the data (e.g. censoring, competing risks, sampling of controls) accounted for appropriately?
 - * Were relevant model performance measures evaluated appropriately?
 - * Was model overfitting, underfitting and optimism in model performance accounted for?
 - * Do predictors and their assigned weights in the final model correspond to the results from multivariable analysis?

Appendix 4. Draft 'Summary of findings' table

Prognostic models for predicting recurrence in women with endometrial cancer

Population: people with a new diagnosis of primary endometrial cancer **Index model:** all clinicopathological and molecular prognostic models used to predict recurrence and survival in the population

Comparator: all models will be compared to FIGO staging (2009)

Timing: recurrence within five years and at any time following diagnosis and death at any time following diagnosis **Setting:** risk stratification for adjuvant therapy

Outcomes	Measure	No of partici- pants	Summary measure	Pooled re- sult(95% CI)	Comment
		(studies)			
Recurrence within five years of diagnosis					
Recurrence at any time following diagnosis					
All-cause mortality					
Cancer-specific mortality					
GRADE Working Group grades of evidence					

High quality: Further research is very unlikely to change our confidence in the estimate of effect.

Moderate quality: Further research is likely to have an important impact on our confidence in the estimate of effect and may change the estimate.

Low quality: Further research is very likely to have an important impact on our confidence in the estimate of effect and is likely to change the estimate.

Very low quality: We are very uncertain about the estimate.

CONTRIBUTIONS OF AUTHORS

YLW & IES drafted the protocol, which was reviewed and agreed by all authors. The search strategy was developed by YLW, TDJW and AM with the assistance of Information Specialist, Jo Platt from the Cochrane Gynaecological, Neuro-oncology and Orphan Cancers group. All authors will contribute to the selection of the studies to include and TDJW, BR, IES, EB will extract data from studies and carry out the analysis. The analysis will be interpreted and the final review will be drafted by YLW, IES, RJE & EJC.



DECLARATIONS OF INTEREST

YLW: no conflicts of interests to declare IES: no conflicts of interests to declare TDJW: no conflicts of interests to declare BR: no conflicts of interests to declare EB: no conflicts of interests to declare RJE: no conflicts of interests to declare EJC: no conflicts of interests to declare

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