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Personalizing prognosis in colorectal cancer: A systematic review of the quality and nature of clinical prognostic tools for survival outcomes

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Abstract

Integrating diverse types of prognostic information into accurate, individualized estimates of outcome in colorectal cancer is challenging. Significant heterogeneity in colorectal cancer prognostication tool quality exists. Methodology is incompletely or inadequately reported. Evaluations of the internal or external validity of the prognostic model are rarely performed. Prognostication tools are important devices for patient management, but tool reliability is compromised by poor quality. Guidance for future development of prognostication tools in colorectal cancer is needed.

Keywords

nomogram; colorectal neoplasms; prognosis; clinical prediction tool

Introduction

The Tumour Node Metastasis (TNM) staging classification system is the foundation of prognostication in colorectal cancer; however, variation in survival and optimal clinical management strategies exist within stage groupings.[1–3] The 7th edition of the UICC/AJCC anatomic stage introduced anatomically-based subgroupings within stage II and III

disease to account for significant prognostic heterogeneity within these groups.[4] While these additional stratifications were successful, the prognostic power of stage for predicting overall survival in an individual patient could be further enhanced by a number of clinical, disease and patient characteristics.[5] Established prognostic factors include depth of tumor invasion into the intestinal wall and presence of nodal metastases,[5] performance status, co-morbid conditions such as diabetes, the presence of venous or lymphatic invasion, and tumour grade.[6,7] Additional complexity in personalized prognostication lies in newly identified biologic, genetic and other molecular information, which have yet a validated role for colon or rectal cancer.[8–11]

Clinicians and patients are continually challenged as to how to best incorporate established and novel prognostic information alongside anatomic stage into a single, individualized estimate of outcome. Clinical prognostication tools, traditionally based on statistical regression models, are one method of combining prognostic information that avoids further stratification of the TNM staging system, which is based on an inelastic mathematical bin model.[12,13] If appropriately developed and validated, these tools have the potential to integrate and personalize the prognostic information available for individual patients and provide refined risk estimates for application to uncertain clinical management scenarios. [14]

The landscape of prognostication tool quality and clinical relevance is currently unknown in colorectal cancer. The Molecular Modellers Working Group (MMWG) of the American Joint Committee on Cancer (AJCC) was formed to understand how information beyond stage could be used to individualize survival prognostication and personalize patient management. The MMWG chose to review the quality and usability of currently available clinical prognostic tools that predict survival in colorectal and four other cancers as their first task. [15–17] The work of the MMWG established the platform for AJCC for the Precision Medicine Core (PMC) of the 8th Edition of the AJCC Cancer Staging Manual, which is envisioned to continue and expand as a service to the oncology community [18,19]. In this article we provide a detailed catalogue and evaluation of publicly available colorectal cancer prognostication tools.

Materials and Methods

Search Strategy and Selection Criteria

Prognostication tools were identified and documentation on their development and validation gathered using three strategies: 1) A search of the peer-reviewed published literature (including a systematic literature review and cited reference search); 2) A search of the web-based scientific community; and 3) Correspondence with individual tool developers when a web-based tool had no corresponding scientific journal article or technical report.

The search strategy was executed in OVID Medline, OVID Embase and HealthStar from Jan 1, 1996–October 6th, 2015. Medical subject headings (MeSH) did not exist for prognostication tools and so a combination of alternate headings and key words were used following consultation with a scientific librarian. Each set of search terms was modified for the specific search engine. For example, the following search terms were used in Medline:

“models, statistical/”, “prognosis”, “predict* model*”, “nomogram/”, “prognos* model*”, and “colorectal neoplasm/”. The searches were limited to English language. Clinically relevant tools originally published prior to 1996 were also included, but these were identified through validation articles found in the systematic literature review. Seemingly eligible studies were excluded if they met any of the following *a priori* exclusion criteria: 1) assessment of the prognostic impact of a single factor (unless it was updating the accuracy of an existing prognostic tool); 2) inappropriate analytic purpose (e.g. multivariate modeling not aimed at prognostication, application of novel statistical methods); 3) not specific to colorectal cancer patients; 4) not original data/research (e.g. editorial, review) or 5) the outcome was not survival. Studies reporting on genomic classifiers built entirely using gene expression data were not the focus of the review and were excluded.

Prognostication tools in this paper include those developed to estimate the probability of survival at a particular point along the disease trajectory (e.g. at diagnosis, following treatment) or for the purpose of using a survival probability to inform treatment decision-making. Eligible survival end-points included all time-to-death analyses (e.g. overall survival, cause-specific survival, relative survival), as well as vital status analyses (e.g. probability of death at 5-years post-diagnosis). Generally speaking, some form of statistical model underlies most prognostication tools, and we use the terms prognostication tool and prognostication model interchangeably. A single reviewer (AM) assessed the titles and abstracts of citations for inclusion. At the beginning, a second reviewer evaluated a random sample of 20 citations to evaluate reliability. Percent agreement was 85%. The first reviewer was conservative and included more articles at the abstract phase than the second reviewer. These differences were easily resolved, and a discussion of discordant decisions determined that the rules for inclusion and exclusion were being applied consistently. A cited reference search using Web of Science was performed. This was implemented to decrease the probability of missing a relevant article. These peer-reviewed literature search strategies to identify prognostic tools in colorectal cancer were supplemented by a Google web-based search. Search terms included: “clinical prediction tool cancer”, “online calculator cancer”, and “nomogram cancer”. The AJCC contacted tool developers for details on tool development if a tool identified in the Google search did not have a supporting article in the peer-reviewed literature and a technical document was not publically available.

Data Abstraction

A detailed report on data abstraction form development and key definitions was published previously. [16,20,21] The data abstracted allowed an evaluation of tool development and validation methodology and clinical relevance. The final criteria include all key elements described by the Transparent Reporting of a multivariable prediction model for Individual Prognosis Or Diagnosis (TRIPOD) guidelines[20,21] and the Critical Appraisal and Data Extraction for Systematic Reviews of Prediction Modelling Studies (CHARMS) checklist[22]. Clinical relevance was informally assessed by considering the prognostic factors’ relevance to the clinical population and to the question addressed by the tool, and by considering the format of the tool (whether or not the equation was provided, usability in a clinical setting). General descriptive information such as study design, study population characteristics and outcome measurement were abstracted, as well as specific details on tool

development (statistical modeling decisions, candidate variable selection) and validation (internal validation methods, measures of model predictive accuracy).

Summary

Key tool development and validation terminology are reported elsewhere.[16,20,21] Descriptive statistics related to tool development and validations are reported in summary tables. The assessment of a tool's calibration and/or discriminative power was defined as a formal statistical evaluation of the internal or external validity. Model calibration assesses how closely the predicted values of the outcome match the observed outcomes in the study sample. Model discrimination assesses the ability of the model to distinguish between individuals who do and do not have the event, at a particular point in time. These established methods are considered the best means of evaluating a clinical prediction model.[12,20,21] We also described when tools were assessed informally through a comparison of survival time distributions across prognostic groups (Kaplan-Meier survival curves). Note however that although these are the same statistical methods that are often used to evaluate the prognostic ability of TNM stage[2,3] they are not considered statistically robust nor are they considered best practice for clinical prediction tool predictive performance assessment. [20,21]

Results

Literature Search Results

Figure 1 describes the search results. The scientific literature review and web-based search identified 53 tools predicting survival in colon or rectal cancer,[23–73] reported across 63 articles. [23–91] Two articles reported on the development of two tools each.[40,58] Eighteen articles contained external validations only.[74–91] One article updated two tools with additional prognostic information.[76] We did not identify any articles evaluating the effectiveness or implementation of tools in clinical practice. Documentation in the peer-review literature was not available for six prediction tools. Correspondence with tool developers added technical documentation for two of those tools, and we were told that the remaining four were pending publication in the peer-review literature.

Tool Development Methods

Table 1 describes key information abstracted on the development of each tool. Further supporting aggregate data are reported. Thirty-nine tools were developed for prognosis in colorectal cancer patients; eight tools were targeted to rectal cancer and six targeted to colon cancer patients respectively. Twenty-nine tools (55%) were designed to predict overall survival (defined as the time between an index date and death from any cause), eleven tools predicted disease-specific survival, six tools did not specify the type of survival outcome, two tools predicted both overall and disease-specific survival, and two predicted cumulative survival at a stated time point (e.g. probability of surviving 5 years). Conditional overall survival, conditional diseases-specific survival and cumulative survival (not otherwise specified) were predicted by one tool each. Thirty (57%) of the tools were presented for clinical use as risk scores or risk groupings. Risk groups are not recommended by the TRIPOD guidelines and their validation is only possible if the risk groups are assigned the

average outcome value, which is rarely done. Twelve were presented as nomograms, and ten as web-based calculators; one tool was presented as a prognostic tree. Only 6 of 53 tools reported the underlying statistical equation with variable coefficients and the intercept, where appropriate. This information is required for external validation using established and appropriate statistical methods.[20,21]

All included prognostication tools were created using data collected for a purpose other than the development of a clinical prediction tool (Table 1). Three tools were developed using data from prospective cohort studies designed with the purpose of investigating prognostic factors, and six others from data collected for one or more randomized controlled trials that were not initially designed to create and/or evaluate a clinical outcome prediction tool. Seventeen tools (32%) were developed using data on cancer populations in the United States, eleven from Japan (21%), five from the United Kingdom (9%) and four each from France and Germany. In the 47 studies that reported colorectal case selection methods, 18 accrued data on patients from multiple institutions and 26 studies used data from a single institution. Data were collected from patients diagnosed or treated for colorectal cancer between 1960 and 2011 and 70% of tools (37/53) were developed on data from patients diagnosed in 2006 or earlier. Sample size for tool development was not reported for six tools, while it ranged from 71 to 128,853 (median = 426) patients when reported. Twenty-eight studies (53%) did not report the number of deaths occurring over the study period. When reported, the number of reported deaths ranged from 52 to 1077 (median = 263).

Populations and Prognostic Factors

The populations addressed by each prognostication tool are described in Table 1. Thirty-five tools (66%) were developed to aid clinical management decisions for patients diagnosed with metastatic colorectal cancer and the majority of these were developed specifically for patients with liver metastasis (26/35 tools). Seven were for all patients with metastatic disease and one each were for patients with lung metastases and patients with malignant spinal cord compression from colorectal cancer. Five were for use with patients diagnosed with all TNM stages of colorectal cancer. Ten tools targeted stage I–III colorectal cancer populations. Two tools were designed for patients with locally advanced and metastatic disease. One tool targeted prognostication in patients with Duke's B colorectal cancer.

Table 2 outlines information on tool development methodology, including prognostic factor selection methods, underlying statistical model, analytic methods for missing data, and the format of continuous variables. Thirty-two tools (60%) did not provide details on the eligibility criteria used to select prognostic factors for the prediction tool and 11 tools applied p-value cut-points or other statistical rules for variable selection. There was significant heterogeneity in the prognostic factors included in tools addressing the same clinical population. For example, none of the 40 prognostic factors included in the 26 tools for patients with liver metastases were common to all tools (Figure 2). The number of liver metastases was the most commonly included variable (22/26 tools) in tools designed for patients with liver metastases. Eighteen variables were found in one liver metastases prediction tool (Table 3). 44/53 tools reported using Cox Proportional Hazards models for time to event data as the basis for their prediction tool. 23 (43%) studies did not define the

index date (e.g. date of diagnosis, date of liver resection), which is critical in order to validate a tool in external populations, as well as for clinical application.

Internal Validity

Forty-five tool development studies (85%) included an evaluation of internal validity: an assessment of the predictive accuracy of the model using the same data used for model development. The majority of these evaluations incorrectly used the entire dataset (100% of patients) that the model was initially developed in to perform their assessment, rather than a form of re-sampling (apparent validation). The recommended approach to evaluating model performance is bootstrapping or cross-validation.[20,21] Bootstrapping methods create new training sets to evaluate model performance by drawing the individuals with replacement from the full data. Cross-validation evaluates model performance by repeatedly randomly splitting the original sample into training (model development) and testing (model validation) sets. Twelve studies used bootstrapping or cross-validation methods.

Twenty-three of the 45 internal validity evaluations (50%) were comparisons of the survival distributions using the log-rank statistic among risk scores or groupings, or among particular risk sets determined by values of the prognostic factors. Model calibration was assessed for 16 tools, generally by providing or referencing graphs (11/16); however, calibration slopes or intercepts and the relationship of the lines to the overall line of identity were rarely discussed. Twenty-one tools evaluated the discriminative ability of the prediction model and reported a concordance index. Concordance indices may take on values from 0.5 (model predictions are similar to chance) to 1.0 (perfect prediction). The values reported in the included studies ranged from 0.59 to 0.81.

External Validity

Half of the tools (27/53) did not have an evaluation of external validity (predictive accuracy in an independent sample separate from the one used for tool development). Seventy-nine assessments of external validity were performed on 26 tools by 33 studies, including those studies that both developed and validated a tool in the same publication. Many of the tools were evaluated multiple times, by different authors. For example, the scoring system developed by Fong and colleagues was validated in 19 separate populations,[35] and the risk classification system by Nordlinger and colleagues was validated in 10 separate populations. [54] The predictive accuracy of seven other tools was evaluated in at least three validation populations.[37,38,45,52,56,58,73]

Of the 26 tools with some evaluation of external validity, 22 had at least one assessment of model calibration, discrimination or another measure of overall model fit. Forty assessments of external validity (51%) examined only the statistical significance of separation of survival curves by risk strata. This method is not endorsed by TRIPOD nor does it appropriately assess predictive performance of the model[20,21]. Nine assessments of model calibration in the additional sample population were performed, four of which were accompanied by a calibration plot; five reported sub-group calibration. The discriminative ability of the evaluated prediction tool was reported as a concordance statistic in 33/79 of external validity evaluations; the range of values across all tools was 0.52 to 0.83.

Discussion

This study summarized available information on 53 colorectal cancer prognostication tools identified from the peer-reviewed literature and web-based resources. These tools were most commonly intended to help inform clinical management decisions in stage IV patients with liver metastases. There were considerable differences in the prognostic factors included in tools designed to prognosticate in similar clinical sub-populations (e.g. within tools for patients with liver metastases). In many cases, tool development methodology was incompletely reported or inadequate. A large number of internal and external validity assessments were performed; however, the majority did not adhere to recommended guidelines for appropriate statistical methodology[20,21]. It is apparent that a framework for moving the science of prognostic tool development and validation forward, as well as its clinical application in oncology, is still needed in order to address the deficiencies highlighted in this systematic review.

The systematic problems identified in the methods used to develop and validate colorectal cancer prognostication tools support the findings of other authors,[92–95] and call for action in the improvement of prognostic tools in oncology. Over 50% of the tools in this review categorize patients into risk groups rather than providing individual probability estimates of survival, decreasing the accuracy for the individual patient.[21] Only 10% of studies with internal validity assessments used bootstrapping, the recommended method for evaluating internal validity. Although 79 external validation exercises were performed, they evaluated a subset of the prognostic tools developed, and half of tools remained with no assessment of generalizability. In addition, 50% of the internal and external validations performed did not adhere to best practices for evaluating predictive performance and did not include an evaluation of calibration or discrimination. The TRIPOD guidelines, published early in 2015, were designed to assist clinicians and scientists in reporting clinical prediction tool studies. However, it is still too early to measure the impact this reporting guideline will have on the quality of future prognostic tool work in oncology.[20,21]

This study also provided an in depth look at the clinical populations and situations addressed by existing tools and the complement of prognostic factors used to make the survival predictions. Gaps remain in the coverage of clinical populations currently addressed by reliable prognostic tools. The majority of tools (67%) attempted to refine prognosis for patients with metastatic disease, reflecting increased uncertainty in clinical management and the need for better risk assessment to understand the benefit of treatment. A need to refine prognosis to inform decisions around the use of adjuvant chemotherapy and radiation in stage II and III patients was also identified. ACCENT, Numeracy, and Adjuvant Online! have been developed for understanding prognosis and the benefit of adjuvant chemotherapy in stage III disease.[25,59,73,77]

The lack of consistency in which prognostic factors were included in the prognostication tools identified in this review highlights the need for improved understanding of prognosis in colorectal cancer, and cautions authors when developing prognostic tools to think carefully about the inclusion of established prognostic factors and the transferability of their findings. Even when prognosis was being refined in the same clinical population, the prognostic

factors included in these tools varied. None of the 40 prognostic factors used in one or more of the 26 tools predicting survival in patients with colorectal liver metastases were common to all of those tools. In the stage IV population with liver metastases, variation in the prognostic factors used across tools may reflect a gap in understanding prognosis for that population or a lack of confidence in the validity of some of those factors. A recent review of prognosis in patients with colorectal liver metastases highlighted 20 different potential prognostic factors, many of which were not included within any of the included tools in our review.[96] In tools designed for non-metastatic patients, the inclusion of clinically significant prognostic factors summarized in the AJCC 7th Edition of the Staging Manual, such as tumour regression grade, serum CEA or tumour deposits were not universal across all prognostic tools reviewed. [4] We have reported similar heterogeneity in prognostic information across prognostic tools in lung cancer[16] and melanoma[17].

The pervasive reliance on retrospective data from single institutions significantly limits what prognostic information may be included in the development of new tools and their widespread clinical usefulness. Only nine prognostication tools for colorectal cancer were developed using prospectively collected data. Designing studies to collect all relevant prognostic information will provide the best individualized estimates of prognosis. New biomarkers and prognostic factors may not be collected in many databases. Half (26/53) relied on data from single institution studies. Prognostication tools developed using data from multi-institutional studies are more likely than single institutional studies to result in relevant, generalizable models. Advances in our ability to understand colorectal cancer will necessitate weighing their added outcome prediction value to existing, affordable, baseline prognostic tools in the future.

This systematic literature review has a number of limitations. Our review may underestimate the number of existing prediction tools designed for survival in colorectal cancer, given the lack of literature search terms at the time to identify relevant studies. However, to account for this both a cited reference search and web-based resources search were performed to widen the net and capture all relevant tools and documentation. The review was also restricted to English language only studies, which may create a language reporting bias. However, the tools included in the study appeared to be developed and validated across a variety of countries and populations. Finally, we did not include tools developed solely using genomic data, as we considered these outside the scope of the review. Therefore, the results of this review may not be representative of the methods and relevance of studies carried out in that area.

The need to refine prognosis for individual patients within TNM stages remains. TNM stage defined the clinical population addressed by the majority of prognostication tools. Within each stage grouping, numerous tools were identified that relied on a multitude of additional prognostic information to individualize predictions. This suggests that clinical prognostic tools may be a viable and clinically relevant option to individualize prognosis without redesigning the TNM stage classification system itself. The AJCC PMC has taken the first step to assuring that existing meritorious tools are made known to the community by establishing the criteria for AJCC endorsement and evaluating tools in major disease areas according to these guidelines.[18,19] The AJCC intends to continue to play a leadership role

in the evaluation, development, and promotion of high-quality prognostication tools in coordination with other authoritative groups such as the PROGNosis RESearch Strategy (PROGRESS) Partnership.[13] The evaluation of prognostic models has been incorporated into the 8th Edition of the staging manual.[18]

Overall, prognostication tools are pervasive in colorectal cancer and may be particularly useful in clinical management when the outcome is uncertain. Guidance in the future direction of prognostication tool development and validation in colorectal cancer is needed. Moving forward, many key clinical and methodological issues in the development, validation and clinical usability need to be addressed. However, addressing statistical and methodological concerns alone will not improve this research area, until consideration is given to the practical strengths and limitations of the literature. We need to build capacity and infrastructure to perform optimal prognostic tool research to realize the potential benefit of these tools for the future.[13,20,21] Collaborative, primary research grants with the objective of developing useful prognostic tools using prospectively collected data, and that include an appropriate assessment of internal and external validity, as well as the evaluation of impact on decision-making will be critical to improving the quality of prognostic tools available for use in colorectal cancer.

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Synopsis

Many prognostication tools have been developed as aids to colorectal cancer patient management, but little is known about their quality. We performed a systematic literature review of colorectal cancer prognostication tools in the peer-reviewed literature and web-based resources. Guidance for future development of prognostication tools in colorectal cancer is needed to assure the quality and clinical utility of these important instruments.

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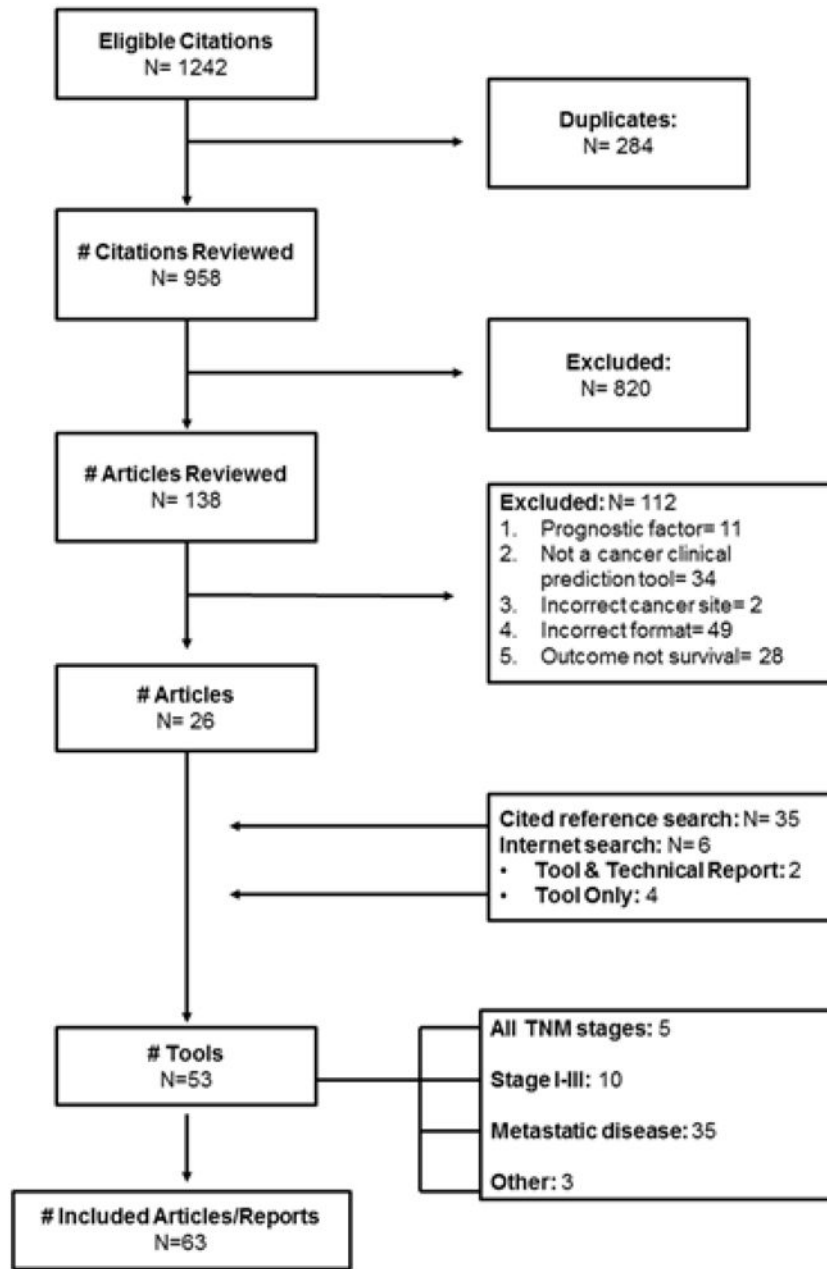


Figure 1. Search results for clinical prognostic tools and their validation

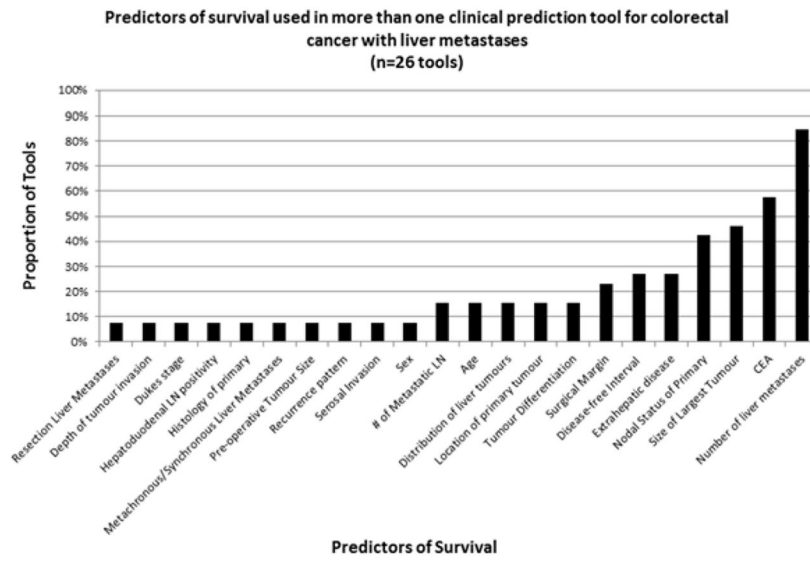


Figure 2. Prognostic factors used in clinical prediction tools targeted at decision-making and prognosis in patients with colorectal cancer and liver metastases

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Table 1

Details on included prognostic tools for colorectal cancer

Tool Citation	Tumour Location	Population	Dates of Data Collection	Study Design	Sample Size	# Events	Duration of Follow-Up	Outcome	Final Variables in Model	Internal Validation	External Validation
Prognostic tools not specific to a particular TNM stage grouping											
[25]	Colon	Stage I–IV	1988–1999	Other	NR	NR	NR	DSS	Age, sex, comorbidity, depth of invasion, # of positive lymph nodes, # of examined lymph nodes, histologic grade	None	Calibration: Y
[50]	Colon	Stage I–IV	NR	Other	NR	NR	NR	NOS	Age, sex, tumour diameter, # positive lymph nodes, CEA, histological type, grade, site, farthest tumour extension (including metastasis)	None	None
[31]	Colon	Stage I–IV	1988–2000	Retrospective Cohort	83,419	NR	Median 87 months	CDSS	Age, sex, ethnicity, tumour grade, AJCC stage	Approach: Bootstrap Calibration: Y Discrimination: 0.816	None
[39]	CRC	Stage I–IV	1990–1999	Retrospective Cohort	**	NR	Mean 43.9 months	OTHER	Tumour depth of invasion, # of metastatic lymph nodes, metastasis, CEA, differentiation, resectability, tumor location, blood transfusion	Approach: Apparent Overall: R ² : -0.64, -0.65	None
[49]	CRC	Locally advanced/Metastatic	1990–1998	Retrospective Cohort	1057	161	2 years	OTHER	Performance status, differentiation, primary site, Duke's stage, number of sites of metastatic disease, CEA, response rate, treatment, number of chemotherapy lines	None	None
[56]	CRC	Dukes B	1988–1996	Prospective Cohort	268	63	Median 65 months	DSS	Peritoneal involvement, venous invasion, margin involvement, tumour perforation	Approach: Apparent*	None
[65]	Rectal	Locally advanced	1992–2003	RCT-PC	2242	850	Median 55.2 months	OS	pT stage, pN stage, cT stage, age, adjuvant chemotherapy, surgery procedure, radiotherapy dose, sex	Approach: Apparent Discrimination: 0.68	Calibration: Y Discrimination: 0.70
[66]	Rectal	Stage I–IV	1994–2003	Retrospective Cohort	42,830	NR	NR	COS	Age, sex, race, stage	Approach: Bootstrap Calibration: Y Discrimination: 0.75	None
Stage I–III											

Tool Citation	Tumour Location	Population	Dates of Data Collection	Study Design	Sample Size	# Events	Duration of Follow-Up	Outcome	Final Variables in Model	Internal Validation	External Validation
[33]	CRC	Stage II	1988–1997	Retrospective Cohort	238	53	Median 110 months	DSS	Tumour growth pattern, extent of tumour spread beyond muscularis propria	Type: Apparent* None	None
[24]	CRC	Stage I–III, PALN	2001–2011	Retrospective	409	NR	NR	OS	LVI, pN+ status, serum CEA 10ng/mL, short axis diameter PALNs 10mm	Type: Apparent* None	None
[73]	Colon	Stage II or III	NR	RCT-PC	3,302	NR	Maximum 8 years	OS	# of positive lymph nodes, depth of tumour, tumour grade, age	Type: Bootstrap Calibration: Y Discrimination: 0.655 Calibration: Y	Calibration: Y
[55]	Rectal	Stage II–III	1986–2005	Retrospective	833	263	Median (in survivors) 51 months	OS	Gender, age, CEA, tumor location, T stage, N stage, Ratio of metastatic lymph nodes, adjuvant chemotherapy, adjuvant chemoradiotherapy	Type: Cross-Validation Discrimination: 0.67	Discrimination: 0.76
[29]	Rectal	Stage I–III	NR	NR	NR	NR	NR	NOS	Age, sex, race, grade, stage, extent of surgery	None	None
[28]	Rectal	Stage I–III	NR	NR	NR	NR	NR	NOS	Age, sex, race, grade, stage, extent of surgery	None	None
[27]	Rectal	Stage I–III	NR	NR	NR	NR	NR	NOS	Age, sex, race, grade, stage, extent of surgery	None	None
[59]	Colon	Stage III	RCT-PC	1989–2002	15995	NR	NR	OS	Age, sex, race, BMI, performance status, tumor grade, tumor stage, ratio of the number of positive lymph nodes to nodes examines, number and location of primary tumors, systemic treatment class	Type: Bootstrap Calibration: Y Discrimination: 0.66	Calibration: Y
[72]	Rectal	Stage I–III	RCT-PC	1987–2002	2618	1077	Median 39.5–75.3 months	OS	Age, gender, tumour distance, surgery type, residual disease, p-T, p-N, presence of post-operative complications	Type: Cross-Validation Calibration: Y Discrimination 0.752	None
[68]	Colon	Stage I–III	Retrospective Cohort	1994–2005	128,853	NR	NR	OS	T-stage, n-stage, number of positive LN, tumour differentiation, patient age, sex	Type: Split Sample Calibration: Y Discrimination: 0.68	None
Stage IV											
[32]	CRC	Stage IV	NR	RCT-PC	803	NR	median 35.3–46.3 months	OS	performance status, LDH, ALP, number of metastatic sites, time to metastasis	Approach: Split Sample Calibration: Y Discrimination: 0.60	Discrimination: 0.63

Tool Citation	Tumour Location	Population	Dates of Data Collection	Study Design	Sample Size	# Events	Duration of Follow-Up	Outcome	Final Variables in Model	Internal Validation	External Validation
[34]	CRC	Stage IV	1995–2010	Retrospective	443	385	Median 62.4 months (range 55.6–77.6)	OS	Number of liver metastases, PCI, type of surgery	Approach: Apparent Calibration: Y Discrimination: 0.61	None
[71]	CRC	Stage IV	1997–2007	Retrospective	1133	278	NR	OS	Post-operative CEA, depth of tumor invasion, lymph node metastasis, peritoneal dissemination	Approach: Split Sample Calibration: Y Discrimination: 0.64	None
[44]	CRC	Stage IV	1984–1999	Prospective	9624	NR	NR	OS	Depth of tumor invasion, regional lymph node metastasis, histologic grade, liver metastasis, lung metastasis, distant lymph node metastasis, peritoneal metastasis, noncurative resection for metastatic lesions	Approach: Apparent*	None
[38]	CRC	Stage IV	1982–1996	RCT-PC	3817	NR	NR	OS	ECOG, WBC count, number of tumor sites, alkaline phosphatase, platelets, location of primary, hemoglobin, peritoneal metastases	Approach: Apparent*	Calibration: Y Discrimination: 0.52 & 0.54 Overall: Schemper: 1.6%
[62]	CRC	Stage IV	2005–2008	Retrospective	124	74	NR	OS	Performance status, pathology, peritoneal metastasis, LDH, PFS interval	Approach: Apparent*	Survival Curves
[30]	Rectal	Stage IV	NR	Other	NR	NR	NR	NOS	Age, sex, race, grade, stage, extent of surgery	None	None
[26]	CRC	Liver mets	1988–1999	Retrospective Cohort	138	99	mean 48.7 months	OS	location of primary, number of liver metastases, preoperative CA19-9, preoperative tumour size	Approach: Apparent*	Discrimination: 0.64
[23]	CRC	Liver mets	2003–2010	Retrospective	100	66	Mean 60 weeks (range 10–238 weeks)	OS	Prior liver surgery, CEA, transaminase toxicity, CT size of two largest lesions	Approach: Apparent Discrimination: 0.81	Discrimination: 0.83
[35]	CRC	Liver mets	1985–1998	Retrospective Cohort	1001	393	median 32 months	OS	nodal status of primary, disease-free interval before presentation of liver metastases, number of tumours, preoperative CEA level, size of the largest tumour	Approach: Apparent Overall: R ² : 0.92	Discrimination: 0.533–0.68
[36]	CRC	Liver mets	1996–2007	Retrospective Cohort	280	NR	median 50.1 months	OS	CEA, number of liver metastases, recurrence pattern, recurrence pattern	Approach: Bootstrap Discrimination: 0.67	None

Tool Citation	Tumour Location	Population	Dates of Data Collection	Study Design	Sample Size	# Events	Duration of Follow-Up	Outcome	Final Variables in Model	Internal Validation	External Validation
[37]	CRC	Liver mets	1981–1996	Retrospective Cohort	243	225	median 32 months	OS	tumour number, tumour size, interval between resection of primary and liver metastases, distribution of liver tumours	None	Discrimination: 0.53–0.64
[40](pre-op)	CRC	Liver mets	1990–1998	Retrospective Cohort	578	337	median 55.2 months	OS, DSS	extrahepatic disease, number of metastatic LN, histology of primary, prehepatectomy CEA level, number of hepatic tumours	Approach: Bootstrap Calibration: Y Discrimination: 0.66	Discrimination: 0.69
[40](post-op)	CRC	Liver mets	1990–1998	Retrospective Cohort	578	337	median 55.2 months	OS, DSS	extrahepatic disease, hilar metastatic LN, number of metastatic LN (primary), histology of primary, surgical margin, prehepatectomy CEA	Approach: Bootstrap Calibration: Y Discrimination: 0.68	Discrimination: 0.7
[42]	CRC	Liver mets	1985–1999	Retrospective Cohort	148	NR	NR	OS	extent of liver metastasis, depth of tumour invasion, peritoneal metastasis	Approach: Apparent*	None
[43]	CRC	Liver mets	1986–2004	Retrospective Cohort	1477	NR	NR	DSS	sex, age, primary site, disease-free interval, pre-operative CEA, number of tumours, largest site of metastasis, bilateral resection, number of involved lobes, primary N stage	Approach: Bootstrap Calibration: Y Discrimination: 0.688	Discrimination: 0.602 & 0.62
[45]	CRC	Liver mets	1993–2006	Retrospective Cohort	201	98	median 31 months	OS	number of metastases, time of diagnosis of the liver metastases, CEA level	Approach: Apparent*	Discrimination: 0.54–0.58
[46]	CRC	Liver mets	1994–2005	Retrospective Cohort	138	NR	median 47.2 months	OS	liver resection margin, CEA, number of liver metastasis, lymph node status	Approach: Apparent*	Survival Curves
[47]	CRC	Liver mets	1977–1997	Retrospective Cohort	135	65	median 73 months	OS	% liver involvement, colic LN, Duke's stage, number of metastases, size of metastases, preoperative: -GT, GPT, type of resection, resection margins	Approach: Apparent*	None
[48]	CRC	Liver mets	1993–2006	Retrospective Cohort	700	NR	median 34 months	OS	Number of metastases, inflammatory response to tumour	Approach: Apparent*	None
[51]	CRC	Liver mets	1980–2002	Retrospective Cohort	369	NR	mean 4.11 years	DSS	Hepatic LN metastases, number of LN around primary tumour, CEA at hepatectomy, number of liver metastases	Approach: Split Sample*	None

Tool Citation	Tumour Location	Population	Dates of Data Collection	Study Design	Sample Size	# Events	Duration of Follow-Up	Outcome	Final Variables in Model	Internal Validation	External Validation
[52]	CRC	Liver mets	1981–1997	Retrospective Cohort	81	NR	median 36.3 months	DSS	Serosal invasion, LN positivity, number of liver metastases, diameter of largest hepatic metastasis, extrahepatic metastasis	Approach: Apparent*	Discrimination: 0.60–0.65
[53]	CRC	Liver mets	1990–2005	Retrospective Cohort	121	52	median 68 months	OS	Location of hepatic metastases, number of metastatic tumours, LN status of primary	Approach: Apparent*	None
[54]	CRC	Liver mets	1968–1990	Retrospective Cohort	1532	689	median 19 months	OS	Age, extension of primary into serosa, lymphatic spread, time interval from primary tumour to metastases, size of largest liver lesion, number of liver lesions, resection margin	Approach: Apparent*	Discrimination: 0.55–0.64
[58](pre-op)	CRC	Liver mets	1987–2005	Retrospective Cohort	929	459	median 26.4 months	DSS	Primary tumour LN status, primary tumour differentiation, CEA, largest tumour diameter, extrahepatic metastatic disease, resection margin	Approach: Split Sample Calibration: Y Discrimination: 0.805	Discrimination: 0.59–0.66
[58](post-op)	CRC	Liver mets	1987–2005	Retrospective Cohort	929	459	median 26.4 months	DSS	Primary tumour LN status, primary tumour differentiation, CEA, number of hepatic metastases, largest tumour diameter, extrahepatic metastatic disease	Approach: Split Sample Calibration: Y Discrimination: 0.781	Discrimination: 0.63–0.74
[60]	CRC	Liver mets	1988–2002	Retrospective Cohort	337	NR	median 16.4 months	CS	Duke's stage, CEA, alkaline phosphatase, number of liver lesions, albumin	Approach: Apparent*	Survival Curves
[61]	CRC	Liver mets	1995–2009	Retrospective	382	327	Median (in survivors) 47 months	DSS	Extrahepatic disease, pN category, number of liver lesions	Approach: Apparent*	None
[63]	CRC	Liver mets	2002–2007	Retrospective	88	76	Median 99 months	OS	Response to systemic therapy, number of CLM, maximum size of CLM, CEA	Approach: Apparent*	None
[64]	CRC	Liver mets	1995–2005	Retrospective Cohort	285	NR	median 4.4–4.6 years	OS	Tumour grade, nodal status	Approach: Apparent Discrimination: 0.59	None
[67]	CRC	Liver mets	1993–2006	Retrospective Cohort	252	NR	NR	DSS	Number of liver metastases, preoperative CEA levels, resection of liver metastases	Approach: Apparent*	None
[69]	CRC	Liver mets	1992–1996	Prospective Cohort	478	NR	NR	OS	Maximum liver metastasis diameter, # of liver metastases	Approach: Apparent*	Discrimination: 0.54–0.58

Tool Citation	Tumour Location	Population	Dates of Data Collection	Study Design	Sample Size	# Events	Duration of Follow-Up	Outcome	Final Variables in Model	Internal Validation	External Validation
[70]	CRC	Liver mets	1960–1995	Retrospective Cohort	662	428	median 3 years	DSS	number of positive lymph nodes, presence of extrahepatic metastases	Approach: Bootstrap Discrimination: 0.61	Discrimination: 0.54–0.58
[41]	CRC	Lung mets	1980–1998	Retrospective Cohort	313	179	median 29 months	OS	Diameter of largest liver tumour, interval from primary tumour to metastasis, hepatoduodenal LN positivity, blood transfusions, primary cancer regional LN, number of metastases ¹	Approach: Bootstrap Calibration: Y Discrimination: 0.72	Discrimination: 0.66 & 0.81
[57]	CRC	MSSC	NR	Retrospective	121	NR	NR	NOS	Primary histology, number of pulmonary tumours, hilar or mediastinal LN, extrathoracic disease, prethoracotomy CEA level	Approach: Apparent*	None

1 = 3-year survival (dichotomous); 5-year survival (dichotomous); 2 =: more than 2 year survival;

** = 3-year survival N=93, 5-year survival N=71;

MSSC=malignant spinal cord compression; OS= overall survival; NOS= not otherwise specified; CRC=colorectal cancer; NR= not reported; DSS= disease specific survival; Schemper=percent of variability in survival explained by the model; LN=lymph nodes

Table 2

Methodological criteria evaluated for clinical prognostication tools for patients with colorectal cancer (n=53)

Methodological Criterion	N (%)
Prognostic Factor Selection Method	
Literature-based/clinical reasoning	7 (13)
Screened using univariable analysis	12 (23)
Available in existing dataset	2 (4)
Method not specified	32 (60)
Methods for Handling Missing Data	
Complete case analysis	8 (15)
Imputation	9 (17)
Unknown variable category used	1 (2)
Method not specified	35 (66)
Description of Handling Continuous Predictors	
Linear	2 (4)
Cubic spline	8 (15)
Transformation	2 (4)
Dichotomized/categorized	41 (77)
Analytic Model Used	
Cox proportional hazards regression	44 (83)
Logistic regression	2 (4)
Recursive partition and amalgamation (RECPAM)	1 (2)
Other	2 (4)
Method not specified	4 (7)
Statistical Model Assumptions Checked	7 (13)

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Table 3

Prognostic factors included in only one tool predicting survival for patients with colorectal cancer and liver metastases (n=23 tools)

lymph nodes around primary
% liver involvement
Albumin
Alkaline Phosphatase
Bilateral resection
Blood transfusion
Colic lymph nodes
Hilar metastatic lymph nodes
Inflammatory Response to Tumour
Peritoneal metastasis
Pre-Operative CA 19-9
Preoperative GPT
Preoperative GT
Recurrence pattern
Resection of Liver Mets (yes/no)
Response to systemic treatment
Transaminase toxicity
Type of resection

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Table 4

Details of evaluations of tool internal and external validity (n=53 tools)

Performance Measure	Internal Validation (n= 45 tools)	External Validation (n= 79 validations)
Internal Validation Method *		
Apparent	27 (60)	--
Cross-Validation	2 (4)	--
Split Sample	6 (14)	--
Bootstrapping	10 (22)	--
External Validation Method		
Independent	--	61 (77)
Geographic	--	13 (16)
Temporal	--	3 (4)
Other **		2 (3)
Overall Model Performance		
R-squared	2 (4)	2 (3)
Calibration		
Graph (Plot/intercept/slope)	11 (25)	4 (5)
Hosmer/Lemeshow statistic	3 (7)	0 (0)
Sub-group calibration ***	2 (5)	5 (6)
Discrimination		
C-statistic ****	21 (48)	33 (42)
Survival Analysis Only with Significance Test		
	22 (50)	40 (51)

* One tool applied both split sample and bootstrap methods;

** An RCT(s) was used to develop the prognostic tool, and an additional RCT was used for validation;

*** Tables comparing predicted and observed values for groups of patients were provided;

**** Concordance index based on the ROC for binary data, Harrell's C statistic for models using time to event data