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Multi-Institutional Validation of a New Renal Cancer–Specific Survival Nomogram

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A B S T R A C T

Purpose

We tested the hypothesis that the prediction of renal cancer-specific survival can be improved if traditional predictor variables are used within a prognostic nomogram.

Patients and Methods

Two cohorts of patients treated with either radical or partial nephrectomy for renal cortical tumors were used: one (n = 2,530) for nomogram development and for internal validation (200 bootstrap resamples), and a second (n = 1,422) for external validation. Cox proportional hazards regression analyses modeled the 2002 TNM stages, tumor size, Fuhrman grade, histologic subtype, local symptoms, age, and sex. The accuracy of the nomogram was compared with an established staging scheme.

Results

Cancer-specific mortality was observed in 598 (23.6%) patients, whereas 200 (7.9%) died as a result of other causes. Follow-up ranged from 0.1 to 286 months (median, 38.8 months). External validation of the nomogram at 1, 2, 5, and 10 years after nephrectomy revealed predictive accuracy of 87.8%, 89.2%, 86.7%, and 88.8%, respectively. Conversely, the alternative staging scheme predicting at 2 and 5 years was less accurate, as evidenced by 86.1% (P = .006) and 83.9% (P = .02) estimates.

Conclusion

The new nomogram is more contemporary, provides predictions that reach further in time and, compared with its alternative, which predicts at 2 and 5 years, generates 3.1% and 2.8% more accurate predictions, respectively.

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INTRODUCTION

Accurate prediction of cancer-specific survival in patients with renal cortical tumors (RCs) is important for counseling, planning of follow-up, and selection for appropriate adjuvant trial designs. The TNM-derived American Joint Committee on Cancer (AJCC) classification represents the gold standard staging scheme after nephrectomy for RC.¹⁻³ Moreover, integrated staging systems demonstrate that the contribution of symptoms at presentation, tumor histology, and tumor size are important in prediction of prognosis.⁴⁻⁶ We hypothesized that the application of routinely available RC-specific survival predictors in a nomogram setting could yield predictive accuracy that exceeds currently available models.

PATIENTS AND METHODS

Patient Population

Five participating institutions contributed 2,576 patients treated with either radical or partial nephrectomy for RC. This cohort constituted the nomogram development cohort, whereas 1,430 additional patients from six institutions were included in the external validation cohort (Table 1). Clinical and pathologic data for these multi-institutional cohorts were gathered prospectively at each center and included patient age at nephrectomy and sex. The symptom classification was defined as described previously.⁷ The Eastern Cooperative Oncology Group (ECOG) performance status was only available in the development cohort. The pathologic tumor specimen was used to define the 2002 T stage, tumor size, and Fuhrman grade. Histologic subtypes were defined according to the 2002 Union International Contre le Cancer classifications and only tumors of clear cell, chromophobe, and papillary histology were included.8 Presence of nodal and metastatic

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	Nomogram Development Cohort		External Validation Cohort		
Variable	No. of Patients	%	No. of Patients	%	
Center					
Rennes University Hospital, Rennes, France	818	32.3	203	14.3	
Henri Mondor University Hospital, Creteil, France	305	12.1			
University Hospital of St Etienne, Etienne, France	528	20.9			
Medical School of University "Federico II," Naples, Italy	204	8.1			
Department of Urology, University of Verona, Verona, Italy	675	26.7			
Medical University of Gratz, Gratz, Austria			553	38.9	
Radboud University Nijmegen, the Netherlands			49	3.4	
Brest University Medical School, Brest, France			49	3.4	
Medical School of Grenoble, Grenoble, France			172	12.1	
Necker Medical School, Paris, France			351	24.7	
Age, years					
Mean	60	0.7	61.5		
Median	62.0		62.6		
Range	10	-91	21-	88	
Sex					
Female	844	33.4	470	34.1	
Male	1,686	66.6	907	65.9	
stage					
1	1,187	46.9	848	61.6	
2	387	15.3	161	11	
3	900	35.6	356	25.9	
4	56	2.2	12	0.9	
umor size, mm	0	7		2	
Madian	6	./	5.0	0	
Niedian	6	.U 2E 0	5.1	U NE O	
	0.0-	-20.0	1.0-2	.5.0	
Conventional clear cell	2 245	88.7	1 208	87 -	
Panillany	2,243	8.4	1,200	07 Q I	
	73	2.9	38	2.9	
ECOG performance status 1	7/9	29.6	Missing	2.0	
	7.10	20.0	Wildonig		
	665	26.3	257	18.	
	835	33.0	746	54 3	
	832	32.9	320	23.2	
IV	198	7.8	54	3.9	
Vodal metastases (N+)	231	9.1	62	4.6	
Distant metastases (M+)	327	12.9	85	6.2	
Symptom classification					
Asymptomatic	1,148	45.4	1,132	82.2	
Local	900	35.5	146	10.6	
Systemic	482	19.1	99	7.2	
RC-specific mortality	598	23.6	168	12.2	
Dverall mortality	798	31.5	253	18.4	
ollow-up, months					
Mean	56	6.0	44.	.4	
Median	38	3.8	38	.0	
Range	0-2	286	0.2-2	249	
otal	2.530	100.0	1.377	100.0	

disease was defined according to intraoperative, pathologic, and radiographic findings. In all patients, the presence of nodal or distant metastases was confirmed by biopsy or pathologic analysis. Patients were staged preoperatively with computed tomography of the abdomen and pelvis, chest computed tomography or chest x-ray, serum electrolytes, and liver function tests. All data collection and analyses were undertaken with the approval and institutional oversight of the Institutional Review Board for the Protection of Human Subjects.

Within the nomogram development cohort of 2,576, 46 patients were excluded because of missing data on tumor size (n = 8), Fuhrman grade (n = 5), histologic subtype (n = 30), cause of death (n = 2), or symptom classification (n = 1). Within the external validation cohort of 1,430 patients, 53 patients were excluded due to missing sex (n = 1) and histologic subtype (n = 52).

The analyses of cancer-specific survival after nephrectomy were performed on 2,530 assessable patients with complete records, whereas external validation was accomplished on 1,377 patients. In both cohorts, assessment of mortality and determination of the cause of death were performed by the treating physician, who relied on chart review and/or death certificate. In cancer-specific survival analyses, perioperative deaths (within 30 days of surgery) were censored.

Statistical Analyses

Univariable and multivariable Cox regression models addressed time to cancer-specific mortality. Main predictors consisted of the 2002 TNM stages, which form the basis for the AJCC stages. Additional variables included age, sex, tumor size, symptom classification, Fuhrman grade, and the histologic subtypes. Reduced model selection was performed using a backward stepdown selection process, which used as stopping rule the Akaike's information criterion.⁹⁻¹¹ Proportional hazards assumptions were verified systematically for all proposed models, using the Grambsch-Therneau residual-based test.¹²

Given that a proportion of patients with RC die as a result of other causes, competing risk regression was used to test the significance of the described variables in predicting RC-specific mortality, after accounting for other-cause mortality. Competing risks regression models account for the effect of other-cause mortality. Strong effect of competing mortality may result in extensive censoring due to cancer-unrelated deaths. Censoring due to other-cause mortality may artificially reduce the pool of individuals at risk of RC-specific events. This may in turn overestimate the effect of RC-specific mortality. The

effect of other-cause mortality cannot be accounted for in Cox regression models. However, the use of competing risks regression, as described by Fine and Gray,¹³ can remove this limitation by distinguishing between RC-specific and other-cause mortality. Unfortunately, there are no commercially available statistical packages that would allow applying competing risks regression within a nomogram setting.

Actuarial survival probabilities were estimated using the Kaplan-Meier method. Multivariate Cox regression coefficients were then used to generate the nomogram. The predictive accuracy of nomograms is usually quantified with receiver operating characteristics–derived area under the curve estimates.^{9,10} In Cox regression models, the area under the curve is substituted with Harrell's concordance index, which was used in this analysis.^{11,12} A value of 100% indicates perfect predictions, whereas 50% is equivalent to a toss of a coin. Internal validation relied on 200 bootstrap resamples.¹⁴ Predictive accuracy estimates were compared using the Mantel-Haenszel test. Calibration plots were generated to explore the performance characteristics of the nomogram at 1, 2, 5, and 10 years after nephrectomy.

Finally, we used the external validation cohort to compare the final, reduced, nomogram-predicted RC-specific mortality versus the observed RC-specific mortality at 1, 2, 5, and 10 years. Nomogram-derived mortality predictions were computed with the nomogram formula, which was derived from the development cohort. Subsequently, the nomogram-predicted probabilities of RC-specific mortality were compared with the observed rates of RC-specific mortality at 1, 2, 5, and 10 years after nephrectomy, and the accuracy of time-specific predictions was quantified using the predicted probability validation method (val.prob) from the S-Plus design library (Statistical Sciences, Seattle, WA). Moreover, the relationship between predicted and observed rates was explored graphically using the val.surv function from the R statistical package. The University of California, Los Angeles Integrated Staging System (UISS), which predicts RC-specific survival at 2 and 5 years, was used as a

Table 2. Univariate and Multivariate Cox Regression Model for Prediction of RC-Specific Survival							
Variable	Univariate Model			Full Multivariate Model		Reduced Multivariate Model	
	Rate Ratio*	P†	Predictive Accuracy (%)	Rate Ratio*	P†	Rate Ratio*	P†
T stage	-	< .001	76.8	—	< .001	—	< .001
1b v 1a	3.918	< .001		2.497	< .001	2.646	< .001
2 v 1a	8.243	< .001		3.514	< .001	3.680	< .001
3 <i>v</i> 1a	20.049	< .001		4.660	< .001	5.274	< .001
4 v 1a	57.727	< .001		6.715	< .001	6.702	< .001
N status (positive v negative)	6.480	< .001	62.3	2.057	< .001	1.977	< .001
M status (positive v negative)	11.103	< .001	69.3	4.656	< .001	4.546	< .001
Tumor size	1.196	< .001	73.3	1.055	< .001	1.043	< .001
Fuhrman grade	—	< .001	74.2	—	< .001	—	< .001
ll v l	2.276	< .001		1.015	.933	1.018	.922
III v I	6.952	< .001		1.515	.023	1.519	.021
IV v I	15.838	< .001		2.253	< .001	2.326	< .001
Histologic type	—	.005	51.7	—	.075		
Papillary <i>v</i> clear cell conventional	0.774	.113		1.296	.118		
Chromophobe v clear cell conventional	0.195	.004		0.392	.107		
Age	1.014	< .001	53.8	1.015	< .001		
Sex (male v female)	1.216	.030	51.2	1.104	.276		
Symptom classification	—	< .001	73.1	—	< .001	—	< .001
Local v asymptomatic	2.783	< .001		1.627	< .001	1.566	< .001
Systemic v asymptomatic	9.319	< .001		2.558	< .001	2.484	< .001
Predictive accuracy, %				86.5		86.3	

Abbreviation: RC, renal cortical tumor.

*Increase in the rate of cancer-specific mortality, relative to reference categories for which the rate ratio is 1.0.

†An overall statistic (P) is provided for categorical variables. For these variables, the rate ratios are provided for each variable level.

comparison benchmark for the newly developed nomogram.⁴ Given that the UISS requires the inclusion of the ECOG performance status, its accuracy was tested within the cohort of 2,530 patients with available ECOG performance status. This validation of the UISS represents an external validation of this prognostic scheme. The Mantel-Haenszel test was used to compare the difference between the accuracy of our nomogram and that of the UISS prognostic scheme. All statistical tests were performed with S-Plus Professional and R statistical packages, and statistical significance was set at .05.

RESULTS

The descriptive statistics of the 2,530 assessable patients are listed in Table 1. T3 stages predominated and accounted for 900 patients (35.6%). Tumor size ranged from 0.5 to 25 cm (mean, 6 cm). Fuhrman grade was I in 665 (26.3%), II in 835 (33%), III in 832 (32.9%), and IV in 198 (7.8%) patients. Conventional clear cell histology was reported in 2,245 (88.7%) patients. Local symptoms were present in 900 (35.5%) patients and systemic symptoms were present in 482 (19.1%) patients.

Of 2,530 patients, 798 (31.5%) died, and 598 of 798 deaths (74.9%) were attributable to RC. As shown in Appendix Figure A1 (online only), the median time to RC-specific mortality was not reached in either the internal (mean, 15.8 years) or the external (mean, 16.2 years) cohorts.

Within the internal validation cohort of 2,530 patients, 2,043 (9,570 person-years; 80.9%), 1,648 (7,765 person-years; 66%), 937 (4,066 person-years; 34.4%), and 344 (1,242 person-years; 10.5%) patients remained at risk of dying at 1, 2, 5, and 10 years, respectively. Actuarial cancer-specific survival probabilities were 89.7% (95% CI, 88.4% to 90.9%), 83.2% (95% CI, 81.5% to 84.7%), 74.2% (95% CI, 72.2% to 76.2%), and 67.2% (95% CI, 64.6% to 69.7%) at 1, 2, 5, and 10 years after surgery, respectively (Appendix Fig A1, online only). For the internal validation cohort, Kaplan and Meier plots of cancerspecific survival according to all clinically and/or pathologically available predictor variables are shown in Appendix Figure A2 (online only).

Table 2 lists the univariate and multivariate Cox RC-specific survival models that were developed on the cohort of 2,530 patients. In univariate analyses, the TNM stages, age, symptoms, tumor size, Fuhrman grade, and histologic subtypes represented highly statistically significant predictors of cancer-specific survival (all $P \le .03$). In the full multivariate model, all included variables achieved overall statistical significance (all P < .001), except for sex (P = .3) and histologic type (P = .07).

Table 3 lists univariate and multivariate competing-risks regression models that were developed on the cohort of 2,530 patients. In multivariate competing-risks regression models, all variables (all P < .04), except for Fuhrman grade II versus I (P = .6) and for histologic subtypes (P = .1 and P = .06), were statistically significant predictors of RC-specific mortality, after accounting for other-cause mortality (Table 3).

Cox model-based analyses of univariate predictive accuracy (Table 2) revealed that T-stage represents the key univariate contributor (76.8%). The combined predictive accuracy of all variables, within the full nomogram (Fig 1) model was 86.5% and exceeded the accuracy of any individual predictor. After backward step-down variable selection, TNM stages, tumor size, Fuhrman grade, and symptom classification remained in the model. These variables yielded the most

Table 3. Univariate and Multivariate Competing Risks Regression Models				
for Prediction of RC-Specific Survival, After Accounting for				
Other-Cause Mortality				

Variable	Univariate Model P	Multivariate Model P	
T stage			
1b <i>v</i> 1a	< .001	< .001	
2 v 1a	< .001	< .001	
3 <i>v</i> 1a	< .001	< .001	
4 v 1a	< .001	< .001	
N status (positive v negative)	< .001	< .001	
M status (positive v negative)	< .001	< .001	
Tumor size	< .001	.01	
Fuhrman grade			
	< .001	.6	
III v I	< .001	.04	
IV v I	< .001	.003	
Histologic type			
Papillary v clear cell conventional	.135	.1	
Chromophobe v clear cell conventional	.004	.06	
Age	.002	.03	
Sex (male v female)	.041	.1	
Symptom classification			
Local v asymptomatic	< .001	< .001	
Systemic v asymptomatic	< .001	< .001	

predictive and the most parsimonious, reduced model nomogram with 86.3% accuracy.

The calibration plots of the internally validated (200 bootstraps) reduced model nomogram (Fig 2) are shown for 1-, 2-, 5-, and 10-year predictions. The internal validation demonstrates virtually no departures from ideal predictions. We compare the predictive accuracy of the reduced nomogram with that of the UISS. In the external validation cohort, the accuracy of our model at 1, 2, 5, and 10 years was 87.8%, 89.2%, 86.7%, and 88.8%, respectively. Conversely, at 2 and 5 years, the UISS was 86.1% and 83.9% accurate, which corresponds respectively to 3.1% (P = .007) and 2.9% (P = .02) gains, relative to the UISS. Finally, Figure 3 shows the graphical comparison between the nomogram-predicted probabilities and the actual fraction surviving within the external validation cohort. The curve virtually follows the 45-degree slope, which indicates ideal performance.

DISCUSSION

Accurate prediction of cancer control after definitive treatment for RC is important for patient counseling, follow-up, and treatment planning. Recently, multivariate models based on various clinical, pathologic, and molecular parameters have been created.⁴⁻⁶ We hypothesized that the inclusion of traditional predictors of RC-specific survival could result in more accurate predictions than those of currently available staging systems. We used Harrell's methodology to quantify these potential gains.^{9,10} Our model development cohort consisted of 2,530 assessable patients who were treated with nephrectomy for RC of various stages at five European institutions. This model was subjected to 200 bootstrap samples to validate our findings



Fig 1. Nomogram predicting renal cell carcinoma (RCC)–specific survival at 1, 2, 5, and 10 years: T, T stage; N, nodal metastases (0, no; 1, yes); M, distant metastases (0, no; 1, yes); S classification, symptom classification; yr survival, disease-specific survival at specific time points.

internally. Given that the gold standard for assessing the performance of a model rests on external validation, we used an additional cohort of 1,377 patients to validate our findings externally.

The TNM stages were combined with tumor size, Fuhrman grade, histologic subtype, age, sex, and symptoms at presentation. Cancer-specific mortality represented our end point of interest. Assessment of mortality and determination of the cause of death were performed by the treating physician, who relied on chart review and/or death certificate. This method is consistent with other large series, in which survival was assessed by the investigators.^{6,8}

We used the nomogram approach described by Harrell et al^{9,10} and popularized by Kattan et al^{15,16} and Sorbellini et al.¹⁷ The 200 bootstrap-adjusted predictive accuracy was used to validate internally the predictive accuracy of the nomogram. The same index was used by three other groups, which devised alternative multivariate prognostic RC survival models.^{6,18,19} Our results yielded a multivariate model, which constitutes the basis for our nomogram. The predictors, which were retained within the most accurate and the most parsimonious nomogram, termed the reduced nomogram, consisted of TNM stages, Fuhrman grade, tumor size, and of symptom classification.

The internally validated predictive accuracy of this nomogram (86.3%) exceeded that of the most informative individual predictor (T stage; 76.8%). External validation of the nomogram at 1, 2, 5, and 10 years after nephrectomy revealed predictive accuracy of 87.8%, 89.2%, 86.7%, and 88.8%, respectively. Conversely, the UISS prognostic scheme predicting at 2 and 5 years was less accurate, as evidenced by 86.1% (P = .006) and 83.9% (P = .02) accuracy.



Fig 2. Calibration plot of the nomogram (86.3% accurate) predicting of renal cancer-specific survival at 1, 2, 5, and 10 years, within the development cohort.



Fig 3. Comparison between nomogram-predicted probability of cancer-specific survival (x-axis) and the actual fraction surviving (y-axis) within the external validation cohort.

Kattan et al⁵ as well as Sorbellini et al¹⁷ developed multivariate nomograms predicting the probability of recurrence after nephrectomy for renal cell carcinoma. The accuracy of these two models was 74% (n = 601) and 82% (n = 701), respectively. Although, the accuracy of our model cannot be compared directly with the accuracies of these nomograms, our results indicate that the discriminant properties of our model (86.3%) are comparable to other models that addressed similar end points.

Zisman et al⁴ (n = 661) and Frank et al⁶ (n = 1,060) developed disease-specific survival models. Frank's model relied on TNM stages, tumor size cutoff of 5 cm, nuclear grade, and tumor necrosis. Its bootstrap-adjusted concordance index was 83.8%. The original Zisman UISS model relied on the AJCC-stage, Fuhrman grade, and ECOG performance status. However, the authors did not provide accuracy estimates. Its stratification criteria were applied to a cohort of 1,060 patients and demonstrated concordance indices that ranged from 79% to 86%, with a mean of 83%. Patard et al¹⁹ also applied the Zisman criteria to a large (n = 4,202) multi-institutional cohort and demonstrated a bootstrap-adjusted concordance index of 80.9%. However, no external validation was performed. Relative to these studies, our nomogram predicts between 2.5% and 5.4% better. This suggests that if predictions are generated for 1,000 consecutive patients, between 25 and 54 patients may be ranked incorrectly if other models are chosen over ours. This is not negligible, especially not to those potentially incorrectly ranked patients, who deserve the most accurate survival predictions. Other advantages of our nomogram over that of previous studies reside in our sample size. Our nomogram-development cohort (n = 2,530) is the second largest and our external validation cohort (n = 1,377) is equally impressive. In addition, our cohorts are more contemporary than those used for the development and validation of the previous models. We used the most stringent statistical methodology to develop and validate our tool. Our model provides accurate predictions that span a 10-year period after nephrectomy, which exceeds the prognostic range of other models. Finally, our tool requires routinely used variables that are virtually invariably recorded. Conversely, the model developed by Frank et al⁶ requires inclusion of tumor necrosis, which is not included routinely in the pathology report. Finally, our model demonstrated higher predictive accuracy than the UISS model when both models were subjected to external validation within the current study.

Our analytic approach also distinguishes itself from previous work by accounting for competing risks. Not all patients with RC die as a result of RC. Other-cause mortality may alter the risk of RCspecific mortality. This effect should be accounted for in disease models, in which the natural history of treated disease allows other-cause mortality to claim patient lives. We complemented our analyses with competing-risks regression, which addressed the significance of the combined multivariate contribution of all risk factors to RC-specific mortality, after accounting for other-cause mortality. This analysis demonstrated that, of variables included in the reduced nomogram, only Fuhrman grade II relative to grade I was not a statistically significant predictor of RC-specific mortality. Lack of availability of commercial software that allows development of competing-risks regression-based nomograms prevented us from relying on this methodology in our prognostic tool.

Our study is not devoid of limitations. Despite having achieved accuracy that exceeded that of other existing models, our nomogram is not perfect. Indeed, 13.7% of predictions will be made incorrectly. This flaw is shared with virtually all predictive models, given that 100% correct predictions virtually are never achieved.^{5,15-17,20-22} The multiinstitutional nature of our data set may be interpreted as a limitation, given that it groups the contribution of multiple surgeons and pathologists and relies on different surgical approaches, in addition to other differences that might distinguish the five contributing centers. Alternatively, our strategy provides a unique opportunity to pool data and increase the statistical power of these outcome studies. Lack of central pathology review might represent another weakness. Central pathology review might have contributed to higher accuracy of pathologically assessed variables and could have improved the overall ability of the nomogram to predict RC-specific survival. Conversely, the use of local pathology analysis confirms the validity of the nomogram when it is used at large. Finally, patients with stage IV disease received a variety of treatments, which ranged from interferon to high-dose interleukin-2. Unfortunately, treatment details were not captured in the institutional databases and could not be included in this analysis.

In summary, we developed a highly accurate (86.3%) nomogram. Its accuracy is superior to all other survival tools, and in a comparison within this study, surpasses the accuracy of the UISS model.

AUTHORS' DISCLOSURES OF POTENTIAL CONFLICTS OF INTEREST

The authors indicated no potential conflicts of interest.

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Appendix

The Appendix is included in the full-text version of this article, available online at www.jco.org. It is not included in the PDF version (via Adobe® Reader®).