

Estimating *TP53* Mutation Carrier Probability in Families with Li-Fraumeni Syndrome Using LFSPRO

Gang Peng¹, Jasmina Bojadzieva², Mandy L. Ballinger³, Jialu Li¹, Amanda L. Blackford⁴, Phuong L. Mai⁵, Sharon A. Savage⁵, David M. Thomas³, Louise C. Strong², and Wenyi Wang¹



Abstract

Background: Li-Fraumeni syndrome (LFS) is associated with germline *TP53* mutations and a very high lifetime cancer risk. Algorithms that assess a patient's risk of inherited cancer predisposition are often used in clinical counseling. The existing LFS criteria have limitations, suggesting the need for an advanced prediction tool to support clinical decision making for *TP53* mutation testing and LFS management.

Methods: Based on a Mendelian model, LFSPRO estimates *TP53* mutation probability through the Elston-Stewart algorithm and consequently estimates future risk of cancer. With independent datasets of 1,353 tested individuals from 867 families, we evaluated the prediction performance of LFSPRO.

Results: LFSPRO accurately predicted *TP53* mutation carriers in a pediatric sarcoma cohort from MD Anderson Cancer Center in the United States, the observed to expected ratio (OE) = 1.35

(95% confidence interval, 0.99–1.80); area under the receiver operating characteristic curve (AUC) = 0.85 (0.75–0.93); a population-based sarcoma cohort from the International Sarcoma Kindred Study in Australia, OE = 1.62 (1.03–2.55); AUC = 0.67 (0.54–0.79); and the NCI LFS study cohort, OE = 1.28 (1.17–1.39); AUC = 0.82 (0.78–0.86). LFSPRO also showed higher sensitivity and specificity than the classic LFS and Chompret criteria. LFSPRO is freely available through the R packages LFSPRO and BayesMendel.

Conclusions: LFSPRO shows good performance in predicting *TP53* mutations in individuals and families in varied situations.

Impact: LFSPRO is more broadly applicable than the current clinical criteria and may improve clinical management for individuals and families with LFS. *Cancer Epidemiol Biomarkers Prev*; 26(6); 837–44. ©2017 AACR.

Introduction

Li-Fraumeni syndrome (LFS) is an autosomal dominant cancer predisposition syndrome (1, 2) commonly manifested as soft tissue and bone sarcomas, breast cancer, brain tumors, leukemia, and adrenal cortical carcinoma (3–5). Clinical criteria were established to describe families with classic LFS (ref. 3; Supplementary Fig. S1).

Germline mutations in the tumor-suppressor gene *TP53* are present in about 70% of families meeting classic LFS criteria (refs. 6, 7; Supplementary Fig. S1). In these families, the lifetime cancer

penetrance for *TP53* mutation carriers is 90% (8, 9), increasing to nearly 100% for women by age 70 years, mostly because of breast cancer (10). Given this extremely high lifetime cancer risk, tools to identify who to test for germline *TP53* mutations may be useful. One study demonstrated significantly improved survival of *TP53* mutation carriers using clinical surveillance and whole-body imaging (11). While several surveillance studies are underway (12–16), whole-body imaging is recommended in U.S. clinical management guidelines (17). Emerging evidence supports improved outcomes with targeted risk management (18).

Classic LFS criteria have high specificity for predicting *TP53* mutations, but low sensitivity. Further criteria assist in identifying potential germline *TP53* mutation carriers (7, 19–22), but are relatively insensitive to *de novo* mutations. Chompret criteria (Supplementary Fig. S1; refs. 20–22) have evolved to identify *de novo* *TP53* mutations, which may affect up to 20% of carriers (23). They have higher sensitivity than classic LFS criteria, but decreased specificity.

Widely used, classic LFS and Chompret criteria provide binary results and only apply to affected counselees and families with LFS-spectrum cancers. To address these limitations, we developed LFSPRO, which uses a Mendelian risk prediction model (24) that has been used successfully in familial breast and ovarian (25), bowel (26), and pancreatic (27) cancers and melanoma (28). We introduced *de novo* mutations into the Mendelian model to incorporate this relatively frequent occurrence in LFS individuals (29) and validated our model using three independent datasets.

¹Department of Bioinformatics and Computational Biology, The University of Texas MD Anderson Cancer Center, Houston, Texas. ²Department of Genetics, The University of Texas MD Anderson Cancer Center, Houston, Texas. ³The Kinghorn Cancer Center and Garvan Institute of Medical Research, Darlinghurst, New South Wales, Australia. ⁴Johns Hopkins Sidney Kimmel Comprehensive Cancer Center, Baltimore, Maryland. ⁵Clinical Genetics Branch, Division of Cancer Genetic and Epidemiology, National Cancer Institute, Bethesda, Maryland.

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Corresponding Author: Wenyi Wang, The University of Texas MD Anderson Cancer Center, 1400 Pressler St Unit 1410, FCT4.5038, Houston, TX 77030. Phone: 713-792-5377; Fax: 713-563-4242; E-mail: wwang7@mdanderson.org

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Materials and Methods

Model development

LFSPRO estimates the probability of any designated family member carrying a *TP53* mutation based on a detailed family cancer history (Supplementary Table S1) and values of two sets of parameters, the prevalence and penetrance for *TP53*, respectively. The prevalence of *TP53* mutations is expressed as $\Pr(G)$, where G denotes genotype, which could be wildtype (denoted as 0), a heterozygous mutation (denoted as 1), or a homozygous mutation (denoted as 2). $\Pr(G)$ for the three genotypes can be derived from the prevalence of mutated alleles using the Hardy-Weinberg equilibrium. The penetrance is the probability of developing cancer at a given age, for *TP53* mutation carriers and noncarriers, respectively. LFSPRO further estimates the individual's future cancer risk if currently asymptomatic, using the estimated carrier probabilities and the penetrance values.

The probability of the counselee's genotype G_0 given the family cancer history is estimated via the following formula (24):

$$\begin{aligned} & \Pr(G_0|H_0, H_1, \dots, H_n) \\ &= \frac{\Pr(G_0) \Pr(H_0, H_1, \dots, H_n|G_0)}{\sum_{G_0} \Pr(G_0) \Pr(H_0, H_1, \dots, H_n|G_0)} \\ & \Pr(H_0, H_1, \dots, H_n|G_0) \\ &= \sum_{G_1 \dots G_n} \Pr(H_0, H_1, \dots, H_n|G_0, G_1, \dots, G_n) \Pr(G_1, \dots, G_n|G_0) \\ &= \sum_{G_1 \dots G_n} \left[\prod_{i=0}^n \Pr(H_i|G_i) \right] \Pr(G_1, \dots, G_n|G_0) \end{aligned}$$

Here, H_i denotes the cancer history for individual i , including the cancer status, diagnosed as having cancer, healthy, or dead; and age, age at diagnosis, at last contact or at death. We use n to denote the total number of counselee's relatives in a family. We may consider the calculation of $\Pr(G_0|H)$ as updating the population prevalence $\Pr(G_0)$ by incorporating family cancer history. The term $\Pr(H_0, H_1, \dots, H_n|G_0)$ is the probability of the phenotypes for the whole pedigree given the genotype of the counselee. Because this is complex to evaluate directly, it is computed as a weighted average of the probabilities of family history given each possible genotype configuration of all relatives, where the weights are the probabilities of the genotype configuration based on Mendelian transmission. The conditional probabilities are products of the individual probability distributions of penetrance when we assume conditional independence. The posterior probability calculation is performed using the Elston-Stewart, or peeling, algorithm (30). This algorithm splits the whole pedigree into anterior and posterior parts according to the individual of interest. The anterior part relates to the parents of the individual, and the posterior part relates to the child/children of the individual. The probability of the anterior and posterior parts can be estimated recursively, such that the posterior genotype probability is calculated according to the probability of the anterior part and the posterior part. The peeling algorithm characterizes Mendelian transmission using a transmission matrix $\Pr(G_i|G_{fi}, G_{mi})$, which is the probability of the genotype for individual i , given the genotypes of his/her father and mother. When Mendel's law is strictly followed, as in previous Mendelian models (26–28, 31), $\Pr(G_i = 1 \text{ or } 2|G_{fi} = 0, G_{mi} = 0) = 0$. However, LFSPRO accounts for the occurrence of *de novo* mutations: when neither parent carries a *TP53* mutation, we still consider a transmission prob-

ability of $2a$ that their child may become a carrier of a *TP53* germline mutation, where a denotes the probability of obtaining a new mutated allele.

The net risk of developing any invasive cancer is calculated as a weighted sum of the penetrance, where weight is the estimated probability of each genotype configuration. We convert these net risks to clinically relevant absolute risks by accounting for the chance of dying of other causes (32), for which we obtain estimates from the SEER 2008–2010 report (33).

Penetrance and prevalence are inputs for LFSPRO

We used previous penetrance estimates for *TP53* mutation carriers and noncarriers from six large pediatric sarcoma families at the University of Texas MD Anderson Cancer Center (MDACC), which are not included in the test cohort (10). For $\Pr(G = 1, 2)$, the prevalence of *TP53* germline mutations, we found estimates from three sources. (i) Based on reports from two population studies (29, 34), we used 0.0001 as a reasonable estimate of the general population prevalence. However, recently available commercial tests for genes that predispose to hereditary cancer syndromes have produced new data on *TP53* germline mutation prevalence. (ii) A study by Myriad Genetics gave a frequency of 1 in 1,600 (82 of 135,609) in a population (affected or unaffected with cancer) undergoing genetic testing (35). (iii) A study by Ambry Genetics found 1 in 300 (69 of 22,226) in a cancer patient population undergoing genetic testing (36). To compare the true prevalence of *TP53* mutations in clinic populations to that in the general population, we modeled LFSPRO outputs, assuming the above prevalence estimates. Gonzalez and colleagues (23) estimated the proportion of *de novo* mutations within germline mutations of *TP53* as 7% to 20%, with paternal or maternal origin. We used both 7% and 20% to calculate the prevalence of *de novo TP53* mutations from the three sets of prevalence data described above. Using three independent datasets (Tables 1 and 2), we evaluated the performance of LFSPRO with inputs of the six combinations of prevalence estimates (Supplementary Table S2). In the LFSPRO software, the user can modify the inputs of penetrance and prevalence.

Test cohorts

We tested our method using three unique datasets, none of which were used in model development (Table 1 and 2). (i) Patients with childhood soft-tissue sarcoma or osteosarcoma treated at MDACC (Houston, TX) from 1944 to 1983 and their extended family members comprised the "pediatric sarcoma cohort." Details on data collection, specific cancers, and germline testing have been published (8, 9, 37–39). The pediatric sarcoma cohort comprised 183 unrelated families with 2,553 individuals. Eleven of the 183 kindreds had at least one individual with a confirmed germline *TP53* mutation; the rest had no confirmed mutation carriers. Due to the nature of ascertainment through probands with pediatric sarcoma, there are more cancer patients in the carrier families than in noncarrier families (average of 14 vs. 1.5, including the proband), resulting in a higher average age at diagnosis in the former set of families (Tables 1 and 2). (ii) The International Sarcoma Kindred Study (ISKS) dataset (40) is a prospective cohort of patients with adult-onset sarcoma and their extended families, recruited from six treatment centers across Australia. This "adult sarcoma cohort" comprised 582 separate families with 16,977 individuals. Nineteen of the kindreds had at least one individual with a confirmed *TP53* mutation. (iii) The

Table 1. Summary of demographic information for families from three study cohorts

Among all families	MDACC pediatric sarcoma		ISKS adult sarcoma		NCI LFS		Total
	Number of carriers > 0	Number of carriers = 0	Number of carriers > 0	Number of carriers = 0	Number of carriers > 0	Number of carriers = 0	
Number of families	11 (6.0%)	172 (94.0%)	19 (3.3%)	563 (96.7%)	78 (76.5%)	24 (23.5%)	867
Number of individuals	1,256 (49.2%)	1,297 (50.8%)	591 (3.5%)	16,386 (96.5%)	1,857 (69.4%)	819 (30.6%)	22,206
Average age at first diagnosis	46.4 (SD = 23.9)	26.2 (SD = 24.3)	46.3 (SD = 21.8)	54.7 (SD = 18.7)	39.9 (SD = 22.1)	52.5 (SD = 22.0)	—
Number of individuals with number of cancers = 1	127 (10.1%)	265 (20.4%)	78 (13.2%)	1,788 (10.9%)	336 (18.1%)	159 (19.4%)	2,753
Number of individuals with number of cancers = 2	19 (1.5%)	0 (0%)	7 (1.2%)	165 (1.0%)	91 (4.9%)	22 (2.7%)	304
Number of individuals with number of cancers > 2	7 (0.6%)	0 (0%)	8 (1.4%)	48 (0.3%)	51 (2.7%)	11 (1.3%)	125

Abbreviation: SD, standard deviation.

NCI LFS cohort (NCT01443468) is from a long-term prospective, natural history study that started in 2011, and includes individuals meeting classic LFS or modified criteria (19), having a pathogenic germline *TP53* mutation or a first- or second-degree relative with a mutation, or a personal history of choroid plexus carcinoma, adrenocortical carcinoma, or at least 3 primary cancers. A detailed family history questionnaire was obtained, including birth date, vital status, date or age at death if deceased, history of cancer, and if positive, type and year of diagnosis or age at diagnosis, for all first-, second-, and third-degree relatives and any other extended family members for whom the information was available. There were 2,676 individuals from 102 families included in the NCI LFS cohort (see Supplementary Materials for details on the datasets). We defined case subjects as affected only if diagnosed with malignant tumors, excluding nonmelanoma skin cancers, but including tumors of the adrenal cortex, choroid plexus, ovary (granulosa), and breast carcinoma *in situ*. Cancers were classified by site: LFS spectrum (osteosarcomas, soft-tissue sarcomas, breast, brain, adrenal, or lung cancers and leukemia) and non-LFS spectrum (prostate, colon, kidney, or thyroid cancers and others).

Mutation testing

Peripheral blood samples were collected after obtaining informed consent. The probands' *TP53* status was determined by PCR sequencing of exons 2 to 11 at MDACC (8). In addition, high-resolution melt analysis and multiplex ligation-dependent probe amplification had been performed in the ISKS and NCI cohort data to detect large deletions or genomic rearrangements (40).

At MDACC, if a *TP53* mutation was identified, then all first-degree relatives of the proband (affected and unaffected by cancer) and any other family member at risk of carrying the familial mutation were tested. Extending germline testing based on mutation status and not on phenotype of family members should not introduce ascertainment bias during analysis (8, 10). Individuals unavailable for testing (largely deceased) linked to or between confirmed mutation carriers were considered obligate mutation carriers. No other family member was tested when the proband tested negative.

Within the NCI cohort, for individuals who had testing prior to enrollment, copies of the clinical *TP53* test reports were obtained and verified by the study team. For individuals actively participating in the protocol and not previously tested, clinical genetic testing was performed after enrollment. All at-risk family members of individuals who tested positive for a mutation (either prior to enrollment or on study) were offered the option of having site-specific testing through the study. No testing was offered to relatives if the proband tested negative for a *TP53* mutation. Obligate carriers were considered mutation-positive regardless of whether testing was done.

Within the ISKS cohort, *TP53* testing was performed on probands, and on family members whoever are willing to participate in the study. The test procedures were as described previously (40).

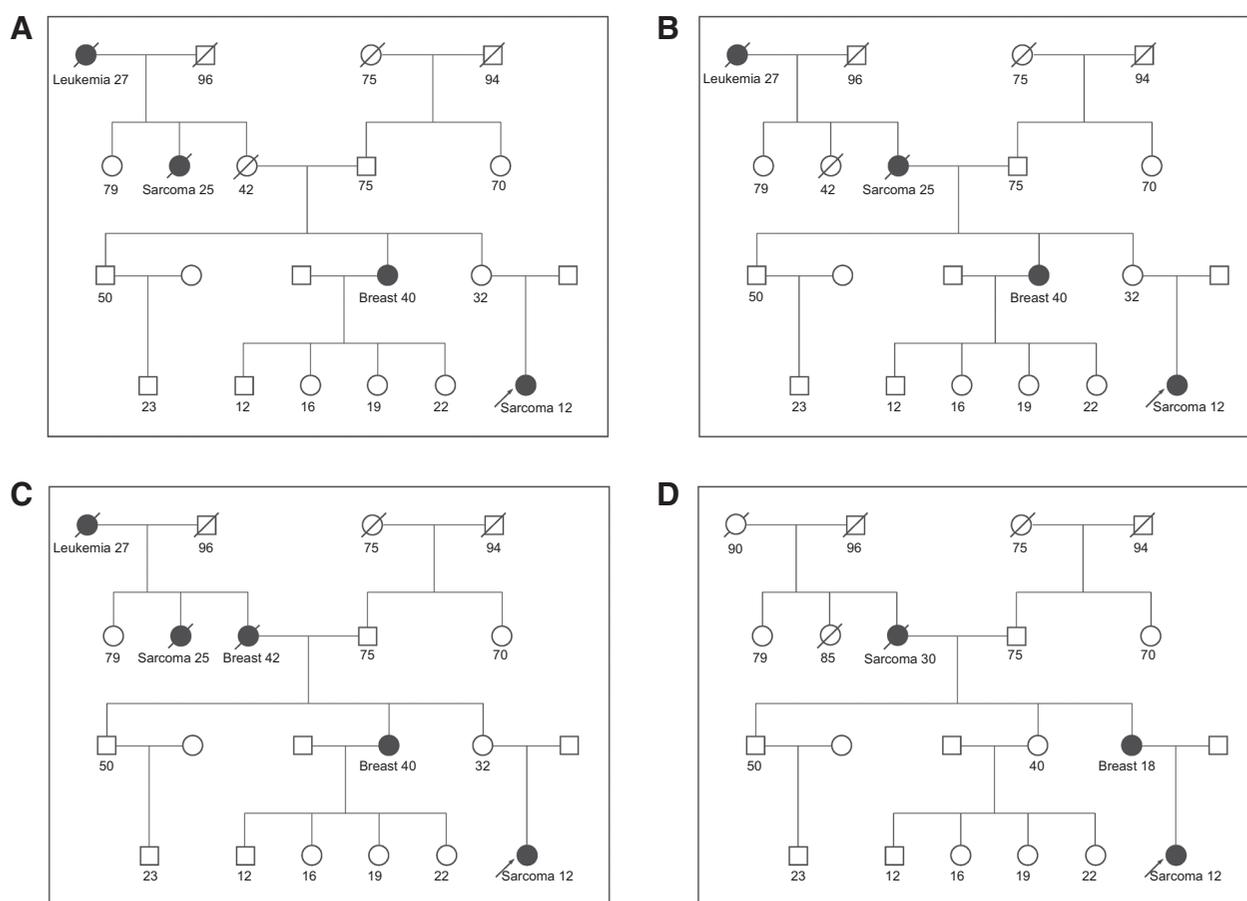
Evaluation of LFSPRO

We used LFSPRO to calculate the carrier probabilities of *TP53* mutations for all tested individuals (*TP53* positive and negative). We used observed estimated ratios (OE) to evaluate the calibration and ROC curves to evaluate our model's discrimination ability. A high area under the ROC curve (AUC) indicates that we can find a point on the ROC curve for determining the *TP53* mutation carrier with a high true-positive rate and low false-positive rate. The OE is the ratio between the number of observed *TP53* mutation carriers and the summation of the estimated probability of *TP53* mutation carriers. The ideal situation is when the number of observed carriers equals the number estimated (OE = 1). When there is more than one family member tested, we do not include mutation information of anyone in the LFSPRO calculation for each family member. We estimated 95% confidence intervals (CI) on our summary measures based on 10,000 bootstrap samples, using the family instead of the individual as the sample unit to account for dependence in outcomes among individuals within a family. For the same individuals, we applied classic LFS and Chompret criteria and compared binary outcomes (mutation carrier or not) with the test results. All analyses were performed using the open-source environment R (<http://cran.r-project.org>).

Table 2. Summary of individuals tested for *TP53* mutation status from three study cohorts

Among <i>TP53</i> tested individuals	MDACC pediatric sarcoma		ISKS adult sarcoma		NCI LFS		Total
	Carriers	Noncarriers	Carriers	Noncarriers	Carriers	Noncarriers	
Number of individuals	32 (11.4%)	249 (88.6%)	25 (4.1%)	588 (95.9%)	210 (45.8%)	249 (54.2%)	1,353
Average age at first diagnosis	25.4 (SD = 24.4)	13.0 (SD = 12.7)	34.4 (SD = 14.3)	46.0 (SD = 17.7)	29.0 (SD = 16.0)	42.1 (SD = 15.3)	—

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**Figure 1.**

A hypothetical family pedigree to illustrate the utility of LFSPRO for genetic counseling. An arrow indicates the counselee, for whom the *TP53* mutation probability is calculated by LFSPRO on the basis of the family history of cancer. Cancer type and age at diagnosis (years) in affected family members are indicated.

A-D, Four variations in clinical scenarios.

Evolution of family history in the pediatric cohort

Information on the 183 kindreds from the pediatric sarcoma cohort at MDACC had been collected over a range of 22 to 62 years (Supplementary Table S3). We arranged the test data, starting when the proband was first diagnosed with invasive cancer, and including only family members, their ages, and their cancer status that existed during that period. We repeated our analysis at 10-year intervals based on the projected family history of that period, which resulted in 4 time points: the proband's ascertainment date was 0, then 10, 20, and the last contact date, with the length of the last time interval varying across families. These time points are relative and thus corresponded to different calendar years for each family, as illustrated in Supplementary Fig. S2, in a subset of families. We applied LFSPRO to the family history "collected" at each time point. In this analysis, we calculated the 95% CIs for the differences in AUCs by first taking the differences in AUCs for each bootstrap sample and then calculating the means and 95% CIs.

Results

Clinical illustration for counseling

Four scenarios of a hypothetical pedigree illustrate how LFSPRO estimates *TP53* mutation probability (Fig. 1, Table 3). Fig-

ure 1A shows four family members diagnosed with early-onset cancers in each of four generations, suggesting a germline *TP53* mutation. The counselee is a potential mutation carrier by Chompret but not classic LFS criteria. Both criteria require the counselee to be a cancer patient. LFSPRO gives the counselee a carrier probability of 57% (Table 3), compared with 0.06% as observed by Myriad Genetics for a genetic testing population without knowledge of family history (ref. 35; see Materials and Methods). Given the high mutation probability, when asymptomatic, the counselee has an elevated risk of developing cancer compared with the population: 31% versus 2% at age 40 years and 43% versus 9% at age 70 years, respectively. In Fig. 1B, the number and types of cancer are constant, but the status of the counselee's grandmother (sarcoma at age 25) is exchanged with that of the grandmother's sister (healthy until death from other causes at age 42). Early cancer in a grandmother within the counselee's direct lineage provides stronger evidence that the counselee may have inherited a germline mutation in *TP53*, and her carrier probability increases to 82%. By contrast, the clinical prediction based on classic LFS or Chompret criteria, which use binary outcomes, is not sensitive to this change. In Fig. 1C, more family members have cancer, further increasing the counselee's carrier probability to

Table 3. Clinical illustration of LFSPRO

	Carrier probability	Counselee's absolute risk of developing cancer ^a		Classic criteria	Chompret criteria
		By age 40 years	By age 70 years		
From a baseline population (34)	0.06%	2%	9%	—	—
Shown in Fig. 1A (scenario 1)	57%	31%	43%	N	Y
Exchange the status of counselee's grandmother and the grandmother's sister (scenario 2, Fig. 1B)	82%	40%	52%	N	Y
Counselee's grandmother is also affected at age 42 years (scenario 3, Fig. 1C)	91%	43%	56%	N	Y
Only the counselee and her mother and grandmother are affected (scenario 4, Fig. 1D) ^b	60% 2%*	31% 3%*	43% 10%*	Y	Y

^aAssuming the counselee has not developed cancer before age 35 years.

^bAll carrier probabilities are calculated allowing for *de novo* mutations, except for scenario 4, where we calculated the carrier probability (*) at *de novo* mutation rate of 0.

91%, whereas the clinical criteria still do not recognize the family as LFS. In these three scenarios, the Mendelian transmission of a germline mutation in *TP53* likely occurred via the counselee's maternal line. Figure 1D demonstrates how LFSPRO captures a *de novo* mutation. Both of the counselee's maternal great-grandparents stayed healthy until death at an advanced age, whereas the counselee's mother had early-onset breast cancer, likely due to *de novo TP53* mutation. Because LFSPRO accounts for probabilities of *de novo* mutations, the counselee's carrier probability is 60%, rather than the 2% obtained from a strictly Mendelian model (24).

LFSPRO predictive performance

Three independent cohorts, comprising 867 families and 22,206 individuals, and 1,353 individuals tested for *TP53* mutations, were used to assess the performance of LFSPRO (Tables 1 and 2) using six settings of the prevalence values for deleterious germline and *de novo TP53* mutations (Supplementary Table S2). LFSPRO performed well in predicting the mutation carrier status under different scenarios of ascertainment. Incorporating *de novo* mutations into LFSPRO substantially improved its predictive power across all prevalence settings (Supplementary Figs. S3–S5). We selected one setting to demonstrate LFSPRO performance as follows. Clinical populations like our cohorts are likely to have enriched *TP53* mutations compared with that in the general population. The closest data we can reliably use to estimate a population undergoing genetic testing are that provided by Myriad Genetics (35), which showed the prevalence of germline mutations in *TP53* as 0.0006 (see Materials and Methods). We set the percentage of *de novo* mutations to be 20% (23), as a similar level of *de novo* mutations was observed in the MDACC dataset (unpublished data). We also use this setting as the default in the LFSPRO software if the user has not otherwise specified this value. Under this setting, the AUCs for the three datasets, MDACC, ISKS, and NCI, are 0.85 (95% CI, 0.75–0.93), 0.67 (95% CI, 0.54–0.79), and 0.82 (95% CI, 0.78–0.86), respectively. The OE ratios are 1.35 (0.99–1.80), 1.62 (1.03–2.55), and 1.28 (1.17–1.39), respectively (Supplementary Table S2).

Figure 2A compares the predictive performance of classic LFS and Chompret criteria and LFSPRO (also Supplementary Figs. S6–S8). LFSPRO provides probabilities rather than binary outcomes and generates an ROC curve with changing sensitivities and specificities as the probability cutoff changes. For the MDACC dataset, classic LFS criteria were highly specific at predicting *TP53* mutation carrier status, but had 12.5% sensitivity; whereas Chom-

pret criteria had higher sensitivity (50%), with a slightly higher false-positive rate (7%). LFSPRO achieved higher sensitivity and specificity than the Chompret and classic criteria (Supplementary Figs. S3 and S6). For ISKS and NCI datasets, Chompret criteria and LFSPRO achieved similar sensitivity and specificity (Supplementary Figs. S4, S5, S7, and S8). However, providing probabilities further allows LFSPRO to reach higher sensitivity while losing a little specificity by lowering the cutoff value (moving along the curve toward the upper right), or to reach higher specificity while losing a little sensitivity by increasing the cutoff value (moving toward the lower left).

The cutoffs for BRCA1/2 genetic counseling using BRCAPRO (31) typically lie between 0.05 and 0.2. In the MDACC data, the respective values of sensitivity and specificity for LFSPRO are 0.84 and 0.86 at a cutoff of 0.05; 0.78 and 0.89 at a cutoff of 0.1; and 0.63 and 0.94 at a cutoff of 0.2. We demonstrate LFSPRO's ability to classify individuals using a cutoff of 0.2, and how frequently it corrects the two criteria's classification of individuals as mutation carriers or noncarriers (Table 4). For the MDACC dataset, LFSPRO classifies more individuals correctly than both criteria. For the ISKS dataset, LFSPRO adds correct classifications of noncarriers (higher specificity) as compared with the Chompret criteria but not the classic LFS criteria. This is likely due to the low number of total mutation carriers observed in ISKS (Tables 1 and 2). Classic LFS criteria's low sensitivity makes it difficult to detect *TP53* mutation carriers, but its high specificity reduces false positives. For the NCI dataset, which contains the highest number of mutation carriers of the three datasets, LFSPRO added correct classifications of mutation carriers (higher sensitivity) but also reclassified some noncarriers incorrectly as carriers (reduced specificity), as compared with both criteria.

Evolution of family history in the pediatric cohort

To better mimic the application of LFSPRO when family data are collected in genetic counseling clinics, we determined the performance of LFSPRO over time following the initial diagnosis of the MDACC pediatric sarcoma dataset. This analysis accounted for the diagnosis of new cancer cases during follow-up, and estimated the predictive power at various time points. Figure 2B shows LFSPRO ROC curves for four time points while rolling forward. The AUCs for LFSPRO increased steadily over time, from 0.50 (no predictive power) at the date of ascertainment to 0.86 (strongly predictive) at the date of last contact. Supplementary Table S4 shows that the differences in AUCs in 10-year intervals are greater than 0 and are statistically significant when the time of

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Table 4. Reclassification of *TP53* mutation carriers using LFSPRO

LFSPRO	Carriers		Noncarriers		Reclassification rate ^a
	Pr ≥ 0.2	Pr < 0.2	Pr ≥ 0.2	Pr < 0.2	
MDACC					
Classic					
+	4	0 ^b	0	1 ^c	11.0%
-	16 ^c	12	14 ^b	234	
Chompret					
+	11	3 ^b	3	11 ^c	12.1%
-	9 ^c	9	11 ^b	224	
ISKS					
Classic					
+	2	0 ^b	1	0 ^c	2.6%
-	4 ^c	19	12 ^b	575	
Chompret					
+	3	7 ^b	2	32 ^c	8.6%
-	3 ^c	12	11 ^b	543	
NCI					
Classic					
+	23	0 ^b	1	0 ^c	49.7%
-	146 ^c	41	82 ^b	166	
Chompret					
+	94	18 ^b	6	8 ^c	38.8%
-	75 ^c	23	77 ^b	158	

^aReclassification rate is defined as the percentage of individuals that were classified differently by LFSPRO (categories b and c) as compared with the other criteria.

^bThe clinical criteria corrected classifications of carrier or noncarrier, as compared with LFSPRO.

^cLFSPRO corrected classifications of carrier or noncarrier, as compared with clinical criteria.

follow-up is more than 10 years. Importantly, the Chompret criteria performed similarly to LFSPRO at each time point and then performed a little worse than LFSPRO at the last time point (Fig. 2B). This result provides empirical evidence to the importance of collecting family history and following up with these families over the years, as is being practiced by genetic counselors at high-risk cancer clinics.

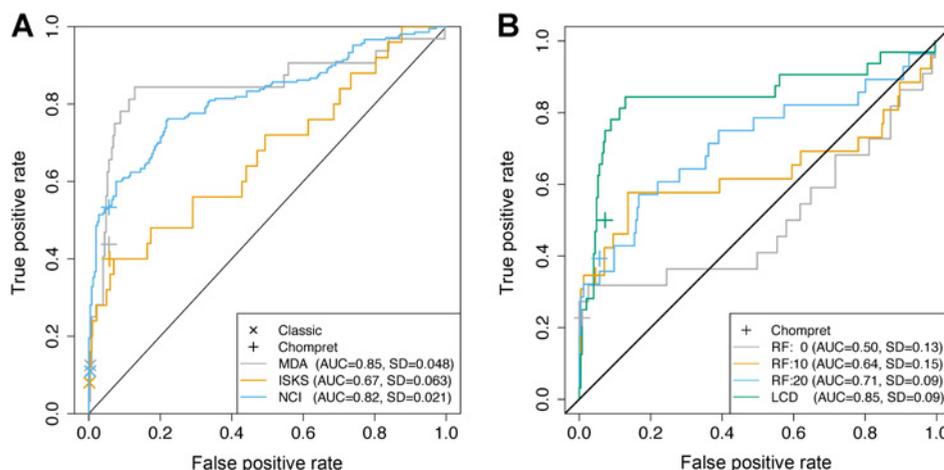
Discussion

We have developed a probabilistic prediction algorithm, LFSPRO, for *TP53* germline mutations and personalized cancer risk associated with LFS (Supplementary Table S1 describes LFSPRO software). We demonstrated the strength of LFSPRO in risk prediction and clinical utility using datasets that represent

varied ascertainment and biological mechanisms related to *TP53* and mixtures of families seen by clinical geneticists and genetic counselors. The associated stand-alone R software is freely available at <http://bioinformatics.mdanderson.org/main/LFSPRO>, and in the widely used BayesMendel R package version 2.1-3, freely available at <http://bcf.dfci.harvard.edu/bayesmendel/>. The LFSPRO software contains functions for classic LFS and Chompret criteria so that users can obtain results from those criteria and LFSPRO for the same data.

Ideally, information on affected family members should improve the sensitivity of *TP53* mutation probability prediction by LFSPRO, whereas information about unaffected family members should lower high false-positive rates. This was evident when we added *TP53*-negative families into our test cohorts. In many *TP53*-negative families, the proband and close relatives also had early-onset LFS-spectrum cancers. The high AUCs on multiple datasets support LFSPRO's ability to discriminate carriers and noncarriers. The OE ratios being close to 1 (with the 95% CIs covering 1 or very close to 1) further support that the carrier probabilities estimated by LFSPRO are consistent with the truth (Supplementary Table S2). We further performed subgroup analysis of OE, dividing the individuals into three risk groups, with estimated probabilities in low risk: 0–0.33; medium risk: 0.34–0.67; and high risk: 0.68–1 (Supplementary Table S5). The OE ratios show good calibration for all risk subgroups in the MDACC dataset, and for the medium- and high-risk subgroups in the NCI dataset. We observed a slight underestimation of carrier probabilities in the low-risk subgroup in the NCI dataset. In the ISKS dataset, there is underestimation in the low-risk subgroup (statistically significant) and overestimation in the medium- and high-risk subgroups (not statistically significant).

The difference in ascertainment may account for the difference in the performance levels that result when using the model on the various cohorts. LFSPRO performs best with the MDACC cohort (ascertainment through a specific tumor type and age, i.e., pediatric sarcoma) and the NCI cohort (ascertainment through a specific type of family history, i.e., families of LFS type), as *TP53* germline mutations are the dominant cause of cancer in both cohorts. Therefore, applying LFSPRO to families in high-risk clinics that are recruited in a similar way will help counsel the family and provide further guidance on screening. In contrast, in the ISKS dataset, an adult sarcoma cohort, many sarcomas common in adults are rare in children, and some, such as well-

**Figure 2.**

Validation results. **A**, ROC curves of LFSPRO for three datasets. Also shown are true positive and false positive rates of classic LFS and Chompret criteria for the three datasets. **B**, ROC curves of the MDACC dataset at different times of follow-up to evaluate the effect of ascertainment of family history in real clinical settings. The corresponding result from the Chompret criteria is also shown. RF, roll forward; LCD, last contact date. In this analysis, the prevalence of germline mutation in *TP53* is 0.0006, and the percentage of *de novo* mutations among all *TP53* germline mutations is 20%.

differentiated or de-differentiated liposarcoma, myxoid liposarcoma, and synovial sarcoma, are not typically associated with somatic or germline *TP53* mutations (40, 41). The ISKS cohort is population-based so the family history of cancer is not taken into consideration when recruiting families. Many of these families did not display the classic LFS family history of cancer, rather subsequent sequencing identified a group of participants that carried germline *TP53* mutations (42). In this case, LFSPRO may still be used, but in conjunction with risk prediction models for other genes when possible [e.g., BRCAPRO (31) for BRCA1/2 mutations in early-onset breast cancer], in order to help identify portions of the phenotype that are caused by *TP53* mutations or other mutations, as people are increasingly identified as carrying *TP53* mutations by nontraditional methods.

The practice of testing for *TP53* mutations has changed over time and is not always 100% accurate. The MDACC dataset was tested more than 10 years ago. The NCI and ISKS datasets were tested more recently, and therefore some might have included deletions or duplications (del/dups). For the NCI dataset, some participants had testing performed prior to enrollment, which might not have included del/dups. We consider the deleterious del/dups in *TP53* to be rare, and the difference in sensitivity across these datasets is likely to be small. All test results are expected to be 100% specific, with sensitivity between 0.95 and 0.99. We decided not to impose an arbitrary number on sensitivity of genetic testing in our analysis. Mathematically, if we did impose a sensitivity value less than 1, our ROC results would not be affected, but the OE ratios would be multiplied by 1/sensitivity.

Classic LFS and Chompret criteria start with the counselee being a cancer patient and are therefore not applicable for directly predicting a healthy individual's carrier status. They also provide a binary prediction, rather than a probability between 0 and 1. Hence, LFSPRO can provide complementary information on LFS risk assessment. Previous risk prediction models like BRCAPRO also provide probabilities between 0 and 1, with cutoffs for BRCA testing chosen by genetic counselors. We project the cutoffs of LFSPRO will converge to a certain range as this prediction software is used together with clinical criteria in clinical settings. The LFSPRO software currently provides an individual's future risk of developing any invasive cancer. In the future, this will be expanded to predict the risk of developing a specific cancer type over the years. Both functions will be evaluated in large cohorts, preferably in conjunction with screening studies of LFS to further improve the practice of genetic counselors and the management of LFS families.

De novo TP53 germline mutations have high prevalence in LFS (23), occurring in 7% to 20% of families. LFSPRO is the first Mendelian model that allows for *de novo* mutations in order to accommodate this relatively frequent event. We show this improved LFSPRO performance compared with that of a strict Mendelian model (Supplementary Figs. S3–S5). LFSPRO well discriminated mutation carriers from noncarriers (evaluated by AUCs) under six sets of assumed prevalence values for germline and *de novo TP53* mutations. Because these values likely depend on ascertainment, users can select these in the LFSPRO software to optimize performance under different clinical scenarios.

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Family history changes over time. The long follow-up in the MDACC pediatric sarcoma cohort allows us to visualize the evolution of family history over time. This pediatric cohort did not look like LFS at the time of ascertainment by any clinical criteria, but slowly became more LFS-like over the years (Fig. 2B). This observation supports the recognized importance among genetic counselors in ascertaining families over a period of years, especially when they are working with pediatric cancer syndromes.

As with BRCAPRO (25) and BOADICEA (43), algorithms that integrate clinical information to guide molecular testing in familial cancer may considerably affect clinical practice. As more clinical genetic testing with large gene panels becomes routine, LFSPRO will potentially aid in the interpretation of genetic variants in *TP53* of unknown significance and prove useful in the management of LFS.

Disclosure of Potential Conflicts of Interest

No potential conflicts of interest were disclosed.

Authors' Contributions

Conception and design: G. Peng, L.C. Strong, W. Wang

Development of methodology: G. Peng, W. Wang

Acquisition of data (provided animals, acquired and managed patients, provided facilities, etc.): J. Bojadzieva, M.L. Ballinger, P.L. Mai, S.A. Savage, D.M. Thomas, L.C. Strong, W. Wang

Analysis and interpretation of data (e.g., statistical analysis, biostatistics, computational analysis): G. Peng, M.L. Ballinger, J. Li, A.L. Blackford, D.M. Thomas, W. Wang

Writing, review, and/or revision of the manuscript: G. Peng, J. Bojadzieva, M.L. Ballinger, A.L. Blackford, P.L. Mai, S.A. Savage, D.M. Thomas, L.C. Strong, W. Wang

Administrative, technical, or material support (i.e., reporting or organizing data, constructing databases): L.C. Strong, W. Wang

Study supervision: L.C. Strong, W. Wang

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Gang Peng, Jasmina Bojadzieva, Mandy L. Ballinger, et al.

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