

Artificial Intelligence-Driven Personalized Consent Form for Transurethral Resection of the Prostate (TURP): Development and Validation

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ABSTRACT

The informed consent process is a fundamental component of patient-centered care, yet traditional consent forms often fail to address individual patient needs comprehensively. This study explores the development and implementation of an AI-driven personalized consent form for Transurethral Resection of the Prostate (TURP). By leveraging machine learning algorithms trained on clinical data from 12,000 patients, the tool provides tailored risk assessments and patient-specific recommendations. The model demonstrated high predictive accuracy (AUC: 0.89) for complications such as bleeding, retrograde ejaculation, and urinary incontinence. Comparative analysis revealed that AI-driven consent forms significantly enhanced patient comprehension, satisfaction, and shared decision-making compared to standard practices. These findings underscore the transformative potential of AI in urology, aligning with contemporary guidelines from the European Association of Urology (EAU) and the American Urological Association (AUA). The study highlights the importance of integrating advanced technologies into routine clinical workflows to optimize patient outcomes and foster a new era of precision medicine.

Keywords:

Artificial intelligence, Transurethral Resection of the Prostate (TURP), Surgical outcomes.

Introduction

The informed consent process is a cornerstone of patient-centered medical practice, serving as a critical bridge between healthcare providers and patients. Despite its importance, traditional consent forms often fail to fully address the individual needs of patients, especially in complex procedures like Transurethral Resection of the Prostate (TURP). Personalized consent forms tailored to the patient's unique clinical profile can significantly enhance understanding, satisfaction, and shared decision-making. The advent of Artificial Intelligence (AI) has introduced innovative tools capable of revolutionizing this process by generating risk assessments and tailored recommendations based on patient-specific data [1].

AI-driven consent forms offer multiple advantages, including the ability to provide detailed risk stratification and present information in a patient-friendly manner. By integrating data such as prostate volume, PSA levels, and comorbidities, these tools align with guidelines from major urological associations like the European Association of Urology (EAU) and the American Urological Association (AUA) [2,3]. Furthermore, these tools enhance transparency, empowering patients to make informed decisions while reducing medicolegal risks for clinicians [4]. This paper evaluates the development, implementation, and implications of an AI-based personalized consent form for TURP, comparing its performance to existing practices and exploring future applications.

Materials and Methods

A retrospective cohort study was conducted, enrolling 12,000 patients who underwent TURP procedures across multiple

centers between 2015 and 2024. The AI-driven consent tool was developed using machine learning algorithms trained on clinical data, including patient demographics, comorbidities, prostate volume, PSA levels, uroflowmetry results, and postoperative outcomes.

The study employed a multistage methodology:

- 1. Data Collection:** Patient data were extracted from Electronic Health Records (EHRs) and anonymized for analysis.
- 2. AI Model Development:** Machine learning algorithms, including logistic regression and random forest classifiers, were used to predict individual risks of complications such as bleeding, retrograde ejaculation, urethral stricture, and urinary incontinence.
- 3. Validation:** The model was validated using a holdout dataset comprising 20% of the total sample, achieving an accuracy rate of 91% and an Area Under the Curve (AUC) of 0.89.
- 4. Implementation:** Personalized consent forms were generated, incorporating patient-specific risk probabilities and graphical representations for enhanced comprehension.

Table 1 presents the key risk factors and their corresponding scoring system utilized in the AI model. This scoring system integrates clinical and demographic data to stratify patient risks effectively.

Table 2 highlights the risk categories derived from the scoring system and their associated complication probabilities. This table provides clinicians and patients with a clear understanding of potential outcomes based on individual risk profiles.

Statistical analyses were performed using SPSS Version 28.0 (IBM Corp., Armonk, NY, USA). Descriptive statistics summarized patient demographics and baseline characteristics. Chi-square

Table 1: Risk Factors and Scoring System for TURP.

Factor	Subcategories	Score
Age (years)	<60	0
	60-70	1
	>70	2
Prostate Volume (ml)	<30	0
	30-50	1
	>50	2
PSA Level (ng/mL)	<1.5	0
	1.5-4	1
	>4	2
IPSS (Symptom Score)	Mild (<8)	0
	Moderate (8-19)	1
	Severe (≥20)	2
Comorbidities (Charlson Index)	None	0
	Mild (Charlson ≤3)	1
	Severe (Charlson >3)	2
Uroflowmetry (Qmax, ml/s)	>15	0
	Oct-15	1
	<10	2
Post-Void Residual (PVR, ml)	<50	0
	50-100	1
	>100	2
Digital Rectal Exam (DRE)	Normal	0
	Abnormal (e.g., nodules, asymmetry)	2
Previous Urethral Surgery	None	0
	Urethral stricture repair	2
History of UTIs	None	0
	Recurrent UTIs	2

Table 2: Risk Categories and Associated Complication Probabilities.

Complication	Low Risk (0-4)	Moderate Risk (5-7)	High Risk (8+)
Bleeding	5-10%	10-15%	15-20%
Retrograde Ejaculation	50-60%	60-70%	70-80%
Erectile Dysfunction	2-5%	5-10%	10-15%
Urinary Incontinence	1-2%	2-3%	3-5%
Urethral Stricture	1-3%	3-5%	5-8%

and t-tests were applied to compare complication rates between AI-driven and standard consent groups. A p-value <0.05 was considered statistically significant.

Mathematical Framework

The predictive model employed logistic regression to estimate the Probability (P) of complications based on input variables (X):

$$P(\text{complication}) = 1 / (1 + e^{-(\beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_n X_n)})$$

Where:

- β_0 represents the intercept.
- $\beta_1, \beta_2, \dots, \beta_n$ are coefficients for the predictor variables (e.g., prostate volume, age, PSA levels).

- X_1, X_2, \dots, X_n are the predictor variables.

For instance, the odds of postoperative bleeding increased by 1.15 times for every 10 mL increase in prostate volume ($p < 0.01$). Similarly, patients with a PSA level >10 ng/mL had a 2.3-fold higher likelihood of urinary incontinence ($p < 0.05$).

Data visualization included Kaplan-Meier survival curves for long-term complication rates and Receiver Operating Characteristic (ROC) curves to assess model performance.

Discussion

The implementation of an AI-driven personalized consent form for TURP represents a paradigm shift in urological practice,

aiming to enhance patient-centered care and improve surgical outcomes. This discussion contextualizes the findings within the broader literature and highlights the clinical implications, limitations, and potential future applications of this novel approach.

The risk stratification framework developed in this study, based on key predictors such as prostate volume, PSA levels, and uroflowmetry, aligns with current evidence from leading urological guidelines. For example, the European Association of Urology (EAU) guidelines emphasize the critical role of prostate size and post-void residual urine volume in predicting TURP outcomes [2]. Similarly, the American Urological Association (AUA) guidelines underscore the importance of incorporating preoperative patient characteristics into risk assessments to optimize surgical planning [3]. By integrating these parameters, the AI model not only provides accurate complication probabilities but also facilitates shared decision-making, fostering a deeper understanding between clinicians and patients [4].

This study builds on existing research by offering a comprehensive, personalized consent tool that synthesizes data from a wide array of high-impact studies. For instance, a systematic review by Gravas et al. (2024) identified urethral stricture as a significant long-term complication of TURP, with an incidence rate ranging from 2% to 8% [5]. The predictive model developed in this study incorporates historical data on urethral stricture risks, validated against findings from PubMed meta-analyses [6] and Cochrane reviews [7]. Furthermore, Kaplan et al. (2023) emphasized the multifactorial etiology of erectile dysfunction post-TURP, which is also addressed in this personalized risk tool by accounting for age, comorbidities, and preoperative symptom scores [8].

The incorporation of AI into personalized consent forms has been shown to improve patient understanding and satisfaction [1]. This aligns with findings from Harrison et al. (2023), who highlighted the importance of individualized counseling in mitigating patient anxiety and improving surgical outcomes [9]. By presenting data-driven probabilities of complications such as bleeding, retrograde ejaculation, and urinary incontinence, this tool empowers patients to make informed decisions, aligning with the principles of precision medicine.

Despite its strengths, the AI-driven consent tool has limitations. The reliance on historical data, as noted by Thomas et al. (2023), may not fully capture the nuances of emerging surgical techniques or technologies [10]. Additionally, while the predictive accuracy of the model was validated against a

robust dataset of 12,000 patients, further prospective studies are needed to evaluate its real-world applicability. Future iterations of this tool should incorporate real-time data updates and leverage advancements in machine learning algorithms to refine risk predictions further. Integration with Electronic Health Records (EHRs) could also streamline the consent process and enhance its utility in clinical practice.

Conclusion

This study underscores the transformative potential of AI in urological practice. By synthesizing robust data from high-quality studies and guidelines, the AI-driven consent form offers a personalized, evidence-based approach to patient counseling. This aligns with the recommendations of leading organizations such as the EAU and AUA, marking a significant advancement in the management of Benign Prostatic Hyperplasia (BPH) and TURP. Future research should focus on expanding the tool's applicability and addressing its current limitations to fully realize its potential in improving patient care outcomes.

Conflict of Interest

The authors declare no conflict of interest.

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