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Predicting the regrowth of clinically non-functioning pituitary adenoma with a statistical model

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Abstract

Background: Compared with clinically functioning pituitary adenoma (FPA), clinically non-functioning pituitary adenoma (NFPA) lacks of detectable hypersecreting serum hormones and related symptoms which make it difficult to predict the prognosis and monitoring for postoperative tumour regrowth. We aim to investigate whether the expression of selected tumour-related proteins and clinical features could be used as tumour markers to effectively predict the regrowth of NFPA.

Method: Tumour samples were collected from 295 patients with NFPA from Beijing Tiantan Hospital. The expression levels of 41 tumour-associated proteins were assessed using tissue microarray analyses. Clinical characteristics were analysed via univariate and multivariate logistic regression analyses. Logistic regression algorithm was applied to build a prediction model based on the expression levels of selected proteins and clinical signatures, which was then assessed in the testing set.

Results: Three proteins and two clinical signatures were confirmed to be significantly related to the regrowth of NFPA, including cyclin-dependent kinase inhibitor 2A (CDKN2A/p16), WNT inhibitory factor 1 (WIF1), tumour growth factor beta (TGF- β), age and tumour volume. A prediction model was generated on the training set, which achieved a fivefold predictive accuracy of 81.2%. The prediction ability was validated on the testing set with an accuracy of 83.9%. The area under the receiver operating characteristic curves (AUC) for the signatures were 0.895 and 0.881 in the training and testing sets, respectively.

Conclusion: The prediction model could effectively predict the regrowth of NFPA, which may facilitate the prognostic evaluation and guide early interventions.

Keywords: Clinically non-functioning pituitary adenoma, Regrowth, Predicting model

Background

Pituitary adenoma (PA) constitutes approximately 15% of all intracranial neoplasm [1]. NFPA is a special type of

pituitary adenoma that generally shows no clinical symptoms such as serum hormone level elevation apart from mass effect, including visual disturbance, headache and various degree of hypopituitarism [2, 3]. Surgery is the first choice of treatment for most patients and no effective drugs can ameliorate its prognosis. However, about 12–58% of patients with macro-adenoma will experience tumour regrowth within 5 years even the tumour remnants was undetectable during surgery or on post-operation magnetic resonance imaging (MRI) [4]. However, we cannot currently predict which patients will experience tumour regrowth after surgery. Although radiotherapy is effective for the reduction of recurrence, the



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potential risk for hypopituitarism or optic nerve injury limits its application. Therefore, clinical or pathological parameters for predicting such regrowth behavior are of necessity.

Few systematic reports have discussed the prediction of the regrowth possibility of NFPA using statistical prediction models. In this study, we identified 41 key proteins whose expression were related to tumorigenesis and progression of pituitary adenoma. We then established a regrowth prediction model with two clinical signatures and three tumour-associated proteins whose expression levels were varied in NFPA. The main purpose of our study was to determine whether we could build a prediction model and validate its authenticity. Our study may facilitate the prognostic evaluation and early intervention to patients with NFPA.

Materials and methods

Patients selection

We retrospectively collected regrowth and non-regrowth tissue from 295 patients diagnosed with NFPA at Beijing Tiantan Hospital from April 2006 to September 2014. All patients were performed enhanced head MRI scan before and after surgery in order to assess tumour volume, Knosp classification and tumour resection grade. The minimum follow-up time was 36 months, and the average follow-up time was 86.5 months (range 36–137 months). Regrowth of NFPA was defined as a tumour maximum diameter increases from any direction on MRI of more than 2 mm from the day of surgery to follow-up endpoint with or without the reappearance of visual disturbance, headache or hypopituitarism.

Sample preparation

NFPA samples were stored and fixed in 10% neutral-buffered formalin immediately upon removal from the sellar region and then embedded in paraffin. Specimen slides stained with hematoxylin and eosin (H&E) were viewed by two different pathologists, who marked the tumour tissues on the samples.

Tissue microarray analyses

Two 2.0-mm diameter samples were removed then transferred to a recipient paraffin block to construct TMAs using a Tissue Array MiniCore 3 (ALPHELYS, Plaisir, France). Samples (4-µm thick) were obtained from each TMA with Leica Rotary Microtome RM2135 (Wetzlar, Germany). Immunohistochemical staining was performed on Leica BOND III automated system. All samples were pretreated in a 65 °C oven for 1.5 h. Staining was performed with dewaxing and epitope retrieval. Samples were blocked with peroxide solution for 7 min. Selected antibodies (shown in Additional file 1: Table S1)

were used to incubate samples, followed by post-primary for 8 min, polymer for 8 min, diaminobenzidine (DAB) for 5 min, and hematoxylin counterstain for 2 min. Finally, the samples were dehydrated, cleared, then fixed with neutral resins. Stained samples were scanned with Leica Aperio AT2 scanner ($400\times$ magnification) and analysed by two different pathologists. Samples were scored as negative (0+), weak (1+), moderate (2+), and strong (3+) signal. The percentage of positivity was also calculated by two pathologists. The histological scores (H-Score) was calculated using following formula: H-Score = $0\times$ (percentage of negative) + $1\times$ (percentage of weak) + $2\times$ (percentage of moderate) + $3\times$ (percentage of strong). Thus, the H-Score ranges from 0 to 300 [5].

Statistical analyses

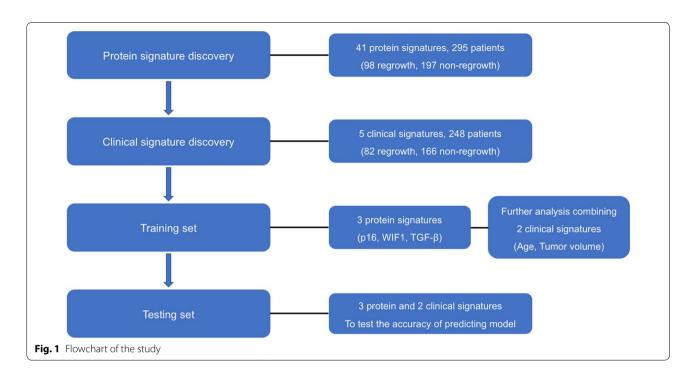
The flowchart of the study is shown in Fig. 1. For candidate protein signatures analyses, the differences in expression levels were evaluated with Wilcoxon-Mann-Whitney test. Kaplan-Meier analyses with two-sided logrank test was used to determine the significance of the survival differences between the two groups, and hazard ratio (HR) was calculated using Cox proportional hazards regression model. The clinical characteristics were analysed with univariate and multivariate logistic regression analyses. Logistic regression algorithm was employed to develop the classification model. The prediction performance of the prediction model was evaluated in terms of the discriminatory accuracy in the training and testing set respectively. The discrimination performance of biomarkers for predicting the regrowth of NFPA was assessed by generating receiver operating characteristic (ROC) curve, the area under the ROC curve (AUC) was used to evaluate the classification performance. All statistical analyses and modeling were performed in R (version 3.1) and IBM SPSS Statistics for Windows, version 21.0. The differences were considered to be statistically significant for P < 0.05.

Results

Discovering candidate protein signatures in tumour tissues

Tumour tissues from 295 cases including 98 regrowth patients and 197 non-regrowth patients were entered into this study. The expression levels of 41 tumour-related proteins in regrowth and non-regrowth tumour were determined via microarray analyses. We then compared the differential expression levels between the two groups, and selected 16 different proteins (presequence translocase-associated motor 16 [PAM16], TGF- β , WIF1, mothers against decapentaplegic homolog 3 [SMAD3(P)], Dopamine receptor D₂ [DRD2], cyclindependent kinase 4 [CDK4], CDKN2A/p16, secreted

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frizzled-related protein 4 [SFRP4], T-box transcription factor [TBX19], somatostatin receptor 1 [SSTR1], high mobility group AT-hook 1 [HMGA1], retinoblastoma protein [RB1], E2F3, cyclin-dependent kinase inhibitor 1 [CDKN1A/p21], estrogen receptor 1 [ESR1], and mouse double minute 2 homolog [MDM2]) that were valuable (P < 0.01, Fig. 2). Meanwhile, in order to assess the associations between protein expression levels with regrowth, we performed Kaplan-Meier analyses to further identify whether there are possible candidates. The event endpoint was calculated from surgical resection to the date of the first regrowth. Patients are divided into 2 groups, low-expression group and high-expression group, according to the median cut-off of the protein expression levels. 17 proteins (PAM16, TGF-β, WIF1, SMAD3(P), DRD2, CDK4, CDKN2A/p16, Ki67, SFRP4, TBX19, SSTR1, HMGA1, RB1, E2F3, CDKN1A/ p21, cyclin-D1 [CCND1], and secretogranin II [SCG2]) were valuable (P < 0.01) including the previous 14 proteins (Fig. 3). Therefore, we used the intersection of the signatures of these 14 proteins to perform further investigations. There are 203 cases expressed all 14 proteins without missing values.

Protein signatures analyses

Based on the 14 candidate protein signatures from the intersection of significant proteins in the differential analyses and survival analyses, a logistic regression analysis was used to screen the most likely candidate factors for tumour regrowth (shown in Additional file 2: Table S2).

Finally, three proteins, WIF1, p16 and TGF- β , with the p values less than 0.05 were selected as predictors for the further establishment of the prediction model (Fig. 3). There are 248 cases expressed all three proteins without missing values; therefore, we chose 248 cases to perform the clinical characteristic analyses and the establishment of the prediction model. The 248 patients were randomly grouped into training and testing sets in a 3:1 ratio using computer-generated random numbers.

Patients characteristics

We preliminarily selected sex, age (<40 years as grade 0, 40–60 years as grade 1, and \geq 60 years as grade 2), tumour volume (macro-adenoma: 10–40 mm as grade 1, and giant adenoma: >40 mm as grade 2), tumour resection rate (total resection as grade 1, subtotal resection as grade 2, partial resection as grade 3) and Knosp grade (0, 1, 2, 3, and 4) as candidate clinical predictors for the prediction of tumour regrowth. All of the selected clinical characteristics were categorical variables. The clinical factors that were statistically significantly associated with the regrowth of NFPA are shown in Table 1. Univariate logistic regression analyses showed that gender was not related to tumour regrowth (P=0.067). Age, Knosp grade, tumour volume and tumour resection rate were found to be correlated with regrowth (P<0.05).

Clinical signature analyses

To further determine the underlying significance of the selected clinical signatures, multivariate logistic Cheng et al. J Transl Med (2019) 17:164 Page 4 of 9

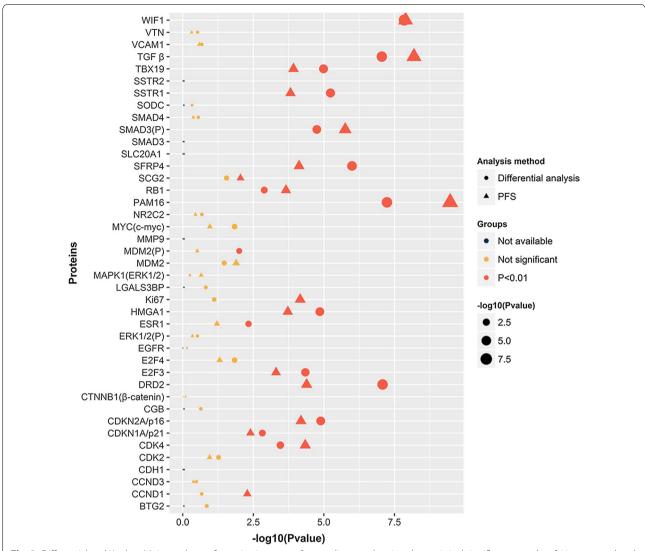


Fig. 2 Differential and Kaplan–Meier analyses of protein signatures. Scatter diagram showing the statistical significance results of 41 tumour related proteins. Sixteen and seventeen protein signatures were valuable in differential and Kaplan–Meier analyses, respectively (P < 0.01)

regression analyses was performed to identify independent factors that were statistically significantly associated with the regrowth of NFPA patients. Age (P<0.01) and tumour volume (P=0.01) were independent risk factors of regrowth (Table 1). Our analyses indicated that the regrowth inclination decreases with age, and patients younger than 40 years are more prone to regrowth compared with those aged from 40 to 60 years (OR=0.2, 95% CI=0.10 to 0.39). In addition, patients aged over 60 years are least likely to have tumour regrowth (OR=0.09, 95% CI=0.03 to 0.24). The tumour volume has a positive correlation with the risk of tumour regrowth. Compared with patients with macro-adenoma, patients with giant

adenoma were more inclined to suffer tumour regrowth (OR = 2.68, 95% CI = 1.55 to 4.62).

Establishment of the regrowth prediction model

Based on the clinical and protein signatures, we established a regrowth prediction model using logistic regression algorithm in the training set, the final prediction model was expressed by the following equation

 $Y=-0.0159\times p16-0.0067\times WIF1-0.0162\times TGF\\ -\beta-1.2260\times age~grade+1.2205\times gross~tumor~volume\\ grade+6.4351.$

The discriminant score P was determined as: $P = \exp(Y)/(1 + \exp(Y))$. In our classification model, the cut-off score was set at 0.5. When the discriminant score

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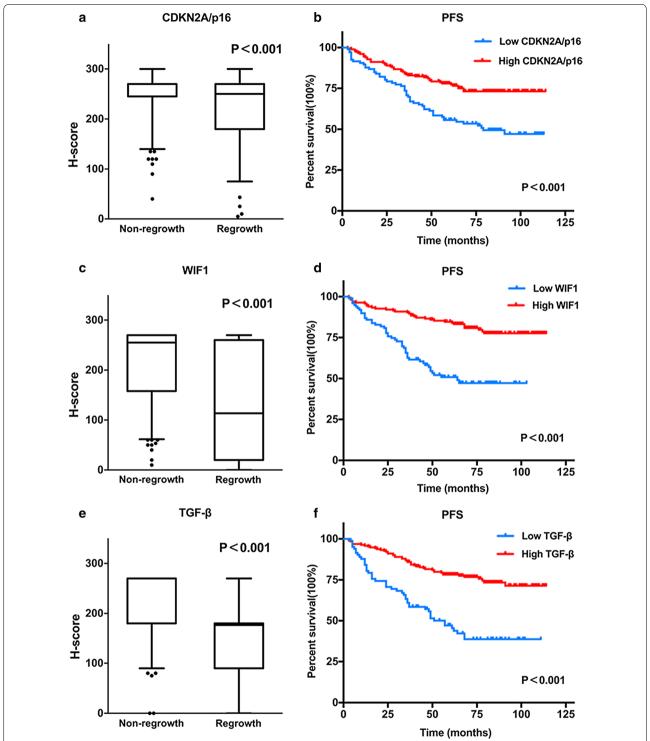


Fig. 3 Differential and Kaplan–Meier analyses of the three protein signatures in patients with NFPA. Distributions of p16, WIF1 and TGF- β protein expression levels in patients of non-regrowth and regrowth cohorts (**a**, **c**, **e**). The Kaplan–Meier curves of PFS between the low and high median protein expression level groups are also shown (**b**, **d**, **f**)

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Table 1 Clinical characteristics of the patients with NFPA

Characteristics	All patients (n = 295)	Regrowth cohort (n = 98)	Non-regrowth cohort (n = 197)	Univariate analysis		Multivariate analysis	
				OR (95% CI)	p value	OR (95% CI)	p value
Gender					0.067		-
Female	140 (47.46%)	56 (57.14%)	84 (42.64%)	1.00 (referent)		-	
Male	155 (52.54%)	42 (42.86%)	113 (57.36%)	0.61 (0.36 to 1.04)		-	
Age					< 0.001		< 0.001
< 40	61 (20.68%)	38 (38.78%)	23 (11.68%)	1.00 (referent)		1.00 (referent)	
40-60	177 (60.00%)	52 (53.06%)	125 (63.45%)	0.20 (0.10 to 0.39)		0.22 (0.11 to 0.43)	
≥60	57 (19.32%)	8 (8.16%)	49 (24.87%)	0.09 (0.03 to 0.24)		0.11 (0.04 to 0.30)	
Tumour volume					< 0.001		0.01
Macro	181 (61.36%)	46 (46.94%)	135 (68.53%)	1.00 (referent)		1.00 (referent)	
Giant	114 (38.64%)	52 (53.06%)	62 (31.47%)	2.68 (1.55 to 4.62)		2.15 (1.20 to 3.87)	
Knosp grade					0.022		0.349
0	62 (21.02%)	11 (11.22%)	51 (25.89%)	1.00 (referent)		_	
1	61 (20.68%)	21 (21.43%)	40 (20.31%)	1.71 (0.63 to 4.67)		_	
2	41 (13.90%)	13 (13.27%)	28 (14.21%)	2.31 (0.83 to 6.41)		_	
3	27 (9.15%)	12 (12.24%)	15 (7.61%)	4.29 (1.45 to 12.65)		_	
4	104 (35.25%)	41 (41.84%)	63 (31.98%)	3.56 (1.49 to 8.48)		_	
Tumour resection rate					0.003		0.254
Total	143 (48.47%)	38 (38.78%)	105 (53.30%)	1.00 (referent)		_	
Subtotal	60 (20.34%)	16 (16.32%)	44 (22.33%)	1.15 (0.55 to 2.42)		_	
Partial	92 (31.19%)	44 (44.90%)	48 (24.37%)	2.75 (1.50 to 5.04)		=	

OR, odds ratios; 95% CI, 95% confidence intervals; Micro, micro-adenoma; Macro, macro-adenoma; Giant, giant adenoma; Total, total resection; Subtotal, subtotal resection; Partial, partial resection

P was calculated to be less than 0.5, the patient was classified as a non-regrowth case, the case was predicted to be a regrowth.

During the training phase, a fivefold cross-validation accuracy was used to assess the unbiased performance of the logistic regression-based classifier in distinguishing regrowth and non-regrowth patients. For the training set of 186 fivefold cross-validation cases, the combined signature provided an overall accuracy of 81.2% (151 of 186), in which 36 of 61 regrowth cases and 115 of 125 non-regrowth cases were correctly predicted (Fig. 4a). The final classification model has obtained a better prediction accuracy in the training set, which provided an accuracy of 84.9%. In this set, 158 of 186 patients were correctly classified, of the 28 patients misclassified, 18 were false-negative and 10 were false-positive, respectively (Fig. 4b). Meanwhile, the prediction ability was validated in the testing set, with an accuracy of 83.9% (52 of 62). Correct predictions were obtained for 16 of 20 (four false-negative cases) regrowth cases and 36 of 42 (six false-positive cases) non-regrowth cases (Fig. 4c).

The ROC analyses were used to assess the sensitivity and specificity of the regrowth prediction. Our classification model based on the combination of clinical and protein signatures showed a good sensitivity and specificity for regrowth prediction with an AUC of 0.895 in the training set (Fig. 5a). We thus validated the model using the same discriminant score in the testing set. The predictive ability was remarkably stable with an AUC of 0.881. In order to explore the possibility whether proteins combined with clinical signatures could be better biomarkers for regrowth prediction, we constructed a model using the protein signatures alone and compared its predictive ability with the model integrating the protein and clinical signatures. The combined model had a significantly better predictive ability than that of the protein signature-alone model, in both the training set (AUC: 0.895 to 0.846, P<0.001, Fig. 5a) and testing set (AUC: 0.881 to 0.808, *P* < 0.001, Fig. 5b). The introduction of clinical signatures performed remarkably well.

Discussion

As a benign tumour derived from the anterior pituitary gland, some pituitary adenomas, like FPA, can induce endocrine disorders by the hypersecretion of hormone, such as adrenocorticotropic hormone (ACTH), thyroid-stimulating hormone (TSH), growth hormone (GH) and prolactin (PRL) [6]. The serum hormone levels and corresponding endocrine symptoms provides a feasible

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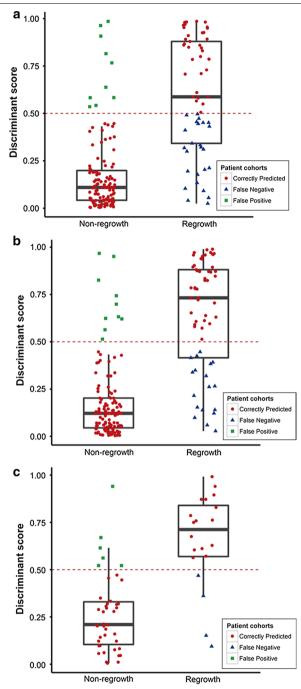


Fig. 4 Distributions of the discriminant scores predicted by the model for the different groups. Box-and-whisker plot showing the distributions of the discriminant scores of the non-regrowth and regrowth groups in the overall fivefold cross-validation cases (**a**), the training (**b**) and testing set (**c**). The three different shapes (or colours) in the boxplot respectively indicates the patients who were correctly predicted and mis-predicted (false positive and false negative)

approach to qualitative diagnosis and regrowth evaluation. Compared with FPA, NFPA lacks endocrine symptoms such as hormone hypersecretion and is mostly diagnosed and postoperatively monitored via imaging examinations, such as enhanced MRI and computed tomography. Besides the imaging techniques, there are few convenient detection approaches for early diagnosis or regrowth monitoring. The possibility of early intervention and fair prognosis may have vanished when the symptoms of mass effect appear or imaging examination reveals abnormal changes. For these reasons, we dedicate to establish an efficient prediction model for the clinical and pathological practice of NFPA. However, research on molecular markers to predict the regrowth of NFPA is far from comprehensive. Researchers had proposed that some proteins have correlation with NFPA regrowth, but few had minutely illustrated their practical application and none had put forward to aggregate analyses forecast model using these proteins and clinical signatures.

In an attempt to identify proteins that allow us to establish a reliable NFPA regrowth prediction model, we started by screening the key papers of pituitary adenoma to identify 41 tumour related proteins. We then narrowed the set to 14 proteins with differential protein expression and Kaplan-Meier analyses. The final three protein signatures including p16, WIF1 and TGF-β were validated via logistic regression analyses with the p values less than 0.05. They were adopted to build the prediction model in which low expression levels were related to regrowth in our analyses. As far as their biological functions are concerned, low expression level of p16 and WIF1 is observed in pituitary tumorigenesis and is necessary for tumour progression, the methylation of their promoters indicates a poor prognosis [7–11]. The TGF- β signaling pathway is a key player in tumour development whose activity is reportedly lower in NFPA which results in infaust outcomes [12, 13]. The combining effect of these genes leads to the proliferation of tumor cells and the increase of tumor volume.

Previous researchers have taken attempt to predict regrowth with clinical features, but they have not reached a consensus in this respect. Tumour residue has always been considered as a risk factor for regrowth because either trans-sphenoidal surgery or open surgery has difficulties in removing tumour invading cavernous sinus and most patients faced with regrowth in around 5 years [4, 14]. However, in our study, multivariate regression analyses showed that tumour resection rate has no relation with regrowth. This may indicate that regrowth may be associated with the instinct

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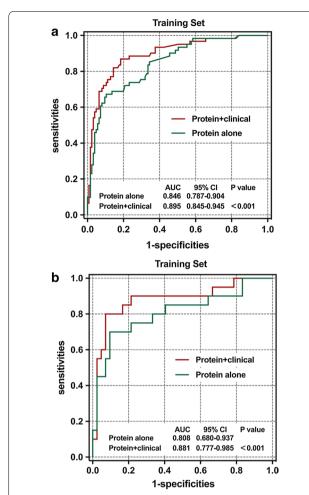


Fig. 5 The regrowth prediction model in the training and testing sets. Two ROC curves showing the comparisons of the sensitivity and specificity for the prediction of regrowth in protein combining clinical signature and protein signature alone in the training (**a**) and testing set (**b**). The combination of protein and clinical signatures shows a better prediction accuracy than that of protein signature alone (P < 0.001)

molecular biology characteristic of pituitary adenoma, which also result in the progression of tumour volume, instead of the surgical intervention.

Knosp grade is a pivotal factor in clinical treatment because of its instruction to surgical resection. However, its significance in tumour regrowth is quite controvertible [15–17]. Our research shows that Knosp grade has no influence on tumour regrowth. This may be rational because its potential is revealed by inducing surgical obstacles and directly result in tumour residue which is not an influential factor in our model.

The potential of age as a risk factor for tumour regrowth deserves some consideration. Several studies have reported that younger age to be strongly correlated with regrowth either in relation to residue or as an independent risk factor [16, 18, 19]. However, other researchers hold that age may not show an underlying tendency to regrowth [4, 20]. Our analyses indicates that age is an independent protective factor in patients with NFPA, and younger age indicates a higher regrowth rate.

In recent years, new genetic markers researches including cell-free nucleic acids or long noncoding RNA have been increasingly reported in a variety of tumour, suggesting a promising new class of molecular markers for tumour diagnosis and prognostic evaluation [21-23]. However, these markers have not been fully adapted for clinical use because of their low sensitivity and complicated process. Furthermore, there have been few similar studies in pituitary adenoma except for some research on molecular mechanism [24, 25]. For these reasons, our model combining the clinical features with pathological signatures is a convenient and feasible way to evaluate the postoperative regrowth of NFPA. However, the number of patients in our study was not large enough which restricts us to evaluate the accuracy of our model in an additional validating set.

Conclusion

To our knowledge, our study is the first to establish a promising statistical model to predict the regrowth of NFPA with clinical and protein signatures. This work provides a novel insight into prognostic evaluation and may help patients to benefit from accurate early intervention.

Additional files

Additional file 1: Table S1. Antibody information involved in the research.

Additional file 2: Table S2. P values of 41 tumour related proteins in differential and Kaplan–Meier analyses.

Abbreviations

PA: pituitary adenoma; FPA: clinically functioning pituitary adenoma; NFPA: clinically non-functioning pituitary adenoma; CDKN2A/p16: cyclin-dependent kinase inhibitor 2A; WIF1: WNT inhibitory factor 1; TGF-β: tumour growth factor beta; ROC: receiver operating characteristic; AUC: area under the receiver operating characteristic curves.

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Everyone that has contributed to this program has been listed as an author.

Authors' contributions

SC, JQW, CYL, XMY and YZZ were responsible for the research design. CZL, SC, YFL and CHL participated in the sample collection and clinical data collection. YFL, CHL and GLL participated in performing the experiment. SC and CZL contributed to the manuscript writing and submission. JQW, SFH, WJL and XMY were responsible for the statistical analyses and figure formatting. CZL and YZZ participated in the funding application and management of the project. All authors have contributed to manuscript. All authors read and approved the final manuscript.

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Availability of data and materials

The datasets during and/or analyzed during the current study available from the corresponding author on reasonable request.

Ethics approval and consent to participate

This study was approved by the medical ethics committee of Beijing Tiantan Hospital.

Consent for publication

All the authors in this paper consent to publication of the work.

Competing interests

The authors declare that they have no competing interests.

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References

- Ostrom QT, Gittleman H, Liao P, Vecchione-Koval T, Wolinsky Y, Kruchko C, Barnholtz-Sloan JS. CBTRUS statistical report: primary brain and other central nervous system tumors diagnosed in the United States in 2010–2014. Neuro Oncol. 2017;19:v1–88.
- Ebersold MJ, Quast LM, Laws ER Jr, Scheithauer B, Randall RV. Long-term results in transsphenoidal removal of nonfunctioning pituitary adenomas. J Neurosurg. 1986;64:713–9.
- Snyder PJ, Bashey HM, Phillips JL, Gennarelli TA. Comparison of hormonal secretory behavior of gonadotroph cell adenomas in vivo and in culture. J Clin Endocrinol Metab. 1985;61:1061–5.
- Brochier S, Galland F, Kujas M, Parker F, Gaillard S, Raftopoulos C, Young J, Alexopoulou O, Maiter D, Chanson P. Factors predicting relapse of nonfunctioning pituitary macroadenomas after neurosurgery: a study of 142 patients. Eur J Endocrinol. 2010;163:193–200.
- Budwit-Novotny DA, McCarty KS, Cox EB, Soper JT, Mutch DG, Creasman WT, Flowers JL, McCarty KS Jr. Immunohistochemical analyses of estrogen receptor in endometrial adenocarcinoma using a monoclonal antibody. Cancer Res. 1986;46:5419–25.
- Kovacs K, Horvath E, Vidal S. Classification of pituitary adenomas. J Neurooncol. 2001;54:121–7.
- Cao L, Gao H, Gui S, Bai G, Lu R, Wang F, Zhang Y. Effects of the estrogen receptor antagonist fulvestrant on F344 rat prolactinoma models. J Neurooncol. 2014;116:523–31.
- Cao L, Gao H, Li P, Gui S, Zhang Y. The Wnt/beta-catenin signaling pathway is involved in the antitumor effect of fulvestrant on rat prolactinoma MMQ cells. Tumour Biol. 2014;35:5121–7.

- Chesnokova V, Zonis S, Rubinek T, Yu R, Ben-Shlomo A, Kovacs K, Wawrowsky K, Melmed S. Senescence mediates pituitary hypoplasia and restrains pituitary tumor growth. Cancer Res. 2007;67:10564–72.
- Machiavelli G, Cotignola J, Danilowicz K, Carbonara C, Paes de Lima A, Bruno OD, Basso A, Szijan I. Expression of p16(INK4A) gene in human pituitary tumours. Pituitary. 2008;11:71–5.
- 11. Quereda V, Malumbres M. Cell cycle control of pituitary development and disease. J Mol Endocrinol. 2009;42:75–86.
- 12. Massague J. TGFbeta in Cancer. Cell. 2008;134:215-30.
- Zhenye L, Chuzhong L, Youtu W, Xiaolei L, Lei C, Lichuan H, Hongyun W, Yonggang W, Fei W, Yazhuo Z. The expression of TGF-beta1, Smad3, phospho-Smad3 and Smad7 is correlated with the development and invasion of nonfunctioning pituitary adenomas. J Transl Med. 2014;12:71.
- Cury ML, Fernandes JC, Machado HR, Elias LL, Moreira AC, Castro M. Nonfunctioning pituitary adenomas: clinical feature, laboratorial and imaging assessment, therapeutic management and outcome. Arq Bras Endocrinol Metabol. 2009;53:31–9.
- Chang EF, Zada G, Kim S, Lamborn KR, Quinones-Hinojosa A, Tyrrell JB, Wilson CB, Kunwar S. Long-term recurrence and mortality after surgery and adjuvant radiotherapy for nonfunctional pituitary adenomas. J Neurosurg. 2008:108:736–45.
- Losa M, Mortini P, Barzaghi R, Ribotto P, Terreni MR, Marzoli SB, Pieralli S, Giovanelli M. Early results of surgery in patients with nonfunctioning pituitary adenoma and analysis of the risk of tumor recurrence. J Neurosurg. 2008;108:525–32.
- Roelfsema F, Biermasz NR, Pereira AM. Clinical factors involved in the recurrence of pituitary adenomas after surgical remission: a structured review and meta-analysis. Pituitary. 2012;15:71–83.
- Losa M, Franzin A, Mangili F, Terreni MR, Barzaghi R, Veglia F, Mortini P, Giovanelli M. Proliferation index of nonfunctioning pituitary adenomas: correlations with clinical characteristics and long-term follow-up results. Neurosurgery. 2000;47:1313–8 (discussion 1318–1319).
- Watts AK, Easwaran A, McNeill P, Wang YY, Inder WJ, Caputo C. Younger age is a risk factor for regrowth and recurrence of nonfunctioning pituitary macroadenomas: results from a single Australian centre. Clin Endocrinol (Oxf). 2017;87:264–71.
- Tampourlou M, Ntali G, Ahmed S, Arlt W, Ayuk J, Byrne JV, Chavda S, Cudlip S, Gittoes N, Grossman A, et al. Outcome of nonfunctioning pituitary adenomas that regrow after primary treatment: a study from two large UK centers. J Clin Endocrinol Metab. 2017;102:1889–97.
- 21. Chen X, Dai M, Zhu H, Li J, Huang Z, Liu X, Huang Y, Chen J, Dai S. Evaluation on the diagnostic and prognostic values of long non-coding RNA BLACAT1 in common types of human cancer. Mol Cancer. 2017;16:160.
- Wu Y, Wang YQ, Weng WW, Zhang QY, Yang XQ, Gan HL, Yang YS, Zhang PP, Sun MH, Xu MD, Wang CF. A serum-circulating long noncoding RNA signature can discriminate between patients with clear cell renal cell carcinoma and healthy controls. Oncogenesis. 2016;5:e192.
- Yamoah K, Johnson MH, Choeurng V, Faisal FA, Yousefi K, Haddad Z, Ross AE, Alshalafa M, Den R, Lal P, et al. Novel biomarker signature that may predict aggressive disease in african American men with prostate cancer. J Clin Oncol. 2015;33:2789–96.
- 24. Li Z, Li C, Liu C, Yu S, Zhang Y. Expression of the long non-coding RNAs MEG3, HOTAIR, and MALAT-1 in non-functioning pituitary adenomas and their relationship to tumor behavior. Pituitary. 2015;18:42–7.
- Robertson AM, Heaney AP. Molecular markers in pituitary tumors. Curr Opin Endocrinol Diabetes Obes. 2016;23:324–30.

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