A novel multi-SNP model predictive of erectile dysfunction following radiotherapy in prostate cancer

Material and Methods: The study cohort comprised 40 patients who underwent neoadjuvant radiochemotherapy (N-RCT) for rectal cancer (28x1.8 Gy, 5 times weekly, concomitant with two cycles 5-FU-based chemotherapy). From each of these patients dermal fibroblasts were cultured from skin specimen gained outside of the radiotherapy planning target volume at occasion of surgery conducted about six weeks upon N-RCT completion. Acute radiation toxicity was thoroughly monitored throughout the N-RCT series and documented according to CTC classification. Maximal acute toxicity (MAT) was defined by the highest CTC grade of the four items “cystitis”, “proctitis”, “enteritis”, and “dermatitis”. MAT was grouped into grades 0/1 (n=16), 2 (n=16), and 3/4 (n=8). N-RCT was simulated in the cultured fibroblasts for five consecutive days (1.8 Gy each at d1-d5 with addition of 5-FU at a concentration reflecting clinical steady-state levels) followed by a 7-day wash-out period. Gene expression of nine candidate genes (CAT, CDKN1A, CTGF, SMAD2, SMAD3, TGFβ1, TGFβ2, SMAD7, GAPDH, and HPRT1) supposed to mediate early radiation-induced toxicity was ascertained by quantitative real time PCR. Samples for these RNA analyses were harvested at d2 and d5 (each 4 hours upon application of the radiation fraction) as well as at day 12 upon the wash-out period. GAPDH and HPRT1 transcript levels served as internal reference.

Results: MAT was related to radiation-induced expression changes of four of the considered genes in fibroblasts. The strongest impact was obtained for SMAD7 and CAT at d5. The higher the MAT score, the lower the induction of SMAD7 and CAT by radiation was (p=0.001 and 0.003). However, upon the wash-out period at d12 no statistical differences in dependence on the MAT score were seen anymore for these two genes. In contrast, a high MAT score was linked to low radiation-induced induction of CTGF (p=0.005) and to a faster decrease of the massively induced Cdkn1a (p=0.03) at d12. At d2, a trend (p=0.06) for CAT in relation to MAT in the same direction as at d5 was noticed with no correlation of any of the other genes at this early time point.

Conclusion: Radiation-induced expression changes in patient-matched fibroblasts may serve as biomarkers to predict clinical radiation toxicity. The role might also be attributed to sustained elevation of TGFbeta signalling and reactive oxygen species load thus saving normal tissue during radiotherapy. A protective role might also be attributed to sustained elevation of Cdkn1a. The link between post-radiation induction of CTGF in fibroblasts and low MAT remains to be clarified. Understanding the mechanistic basis of these findings might pave the way for better protection of irradiated normal tissue.

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Purpose or Objective: Erectile dysfunction (ED) is one of the most common complications encountered after radiotherapy in prostate cancer patients. The goal of this study was to investigate whether single nucleotide polymorphisms (SNPs) are associated with late ED in men treated with radiotherapy for prostate cancer. To this end, we designed a novel machine learning-based multi-SNP model using a genome-wide association study (GWAS) dataset.

Material and Methods: We analyzed 236 evaluable patients with at least one year of follow-up for the development of ED. The severity of ED was assessed using either the patient-administered Sexual Health Inventory for Men (SHIM) or the clinician-assigned Mount Sinai Erectile Function (MSEF) scoring schema. There were 133 patients with Grade 2 or more ED. For our analysis, the cohort was split into two groups (cases/controls: MSEF 0,1 / 2,3; cases/controls: SHIM ≥7 / ≥16). Genome-wide SNP data were available from Affymetrix Genome-Wide Human SNP Array 6.0. After a quality test including SNP missing rate<5%, minor allele frequency (MAF)>5%, and Hardy-Weinberg equilibrium (p<10E-5), 613,496 SNPs remained. For the validation purpose of our proposed model, the dataset was split into a training dataset (2/3 of samples) and a validation dataset (1/3 of samples). Our model building process was performed using the bootstrapped data from the training dataset. Our idea is to convert the binary outcomes into preconditioned continuous outcomes based on normal tissue complication probability (NTCP) using principal component analysis (PCA) and logistic regression. The preconditioned outcomes were used in the model building process using random forest regression. Then, the model was tested using the validation dataset. The final predicted outcomes were compared with the original binary outcomes to estimate the predictive performance. We iterated this process 100 times and the performance was averaged. We compared the performance of our proposed method (preconditioning random forest regression: PRFR) with other methods including preconditioning lasso (PL), lasso logistic regression (LLR), and random forest classification (RFC).

Results: Univariate analysis was performed using the training dataset. With a threshold of p<0.001, 367 SNPs remained. These SNPs were fed into our model. As shown in Figure 1, the averaged performance with the validation dataset was AUC=0.62, which is better than other methods: RFC (0.57), PL (0.58), and LLR (0.57). The 79 patients in the validation dataset were binned into 6 groups according to the predicted risk of ED. Figure 2 shows the comparison of the model-predicted incidence of ED with observed incidence.

Conclusion: Our machine learning-based multi-SNP model showed the potential to better predict the radiation-induced late ED. However, we need to validate our model using other datasets.