**Title**: Basal cell carcinoma risk prediction among long-term survivors of childhood cancer: a report from the Childhood Cancer Survivor Study (CCSS)

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**Background**: Basal cell carcinomas (BCCs) represent >50% of subsequent neoplasms among long-term survivors of childhood cancer. While general population BCC risk prediction models are available, models considering clinical features relevant to survivors are lacking.

**Methods**: To develop risk prediction models for experiencing any subsequent BCC by age 40 years, we used data from five-year survivors in CCSS. BCCs occurring ≥5 years after childhood cancer diagnosis were confirmed via pathology report or medical record review. Predictors included: demographics; childhood cancer diagnosis/age; treatment era; body region-specific cumulative radiotherapy (RT) doses; hematopoietic cell transplantation (HCT); chemotherapy drugs/doses; prior non-BCC skin neoplasms. Among the evaluated statistical learning algorithms, XGBoost was the best-performing algorithm. To assure generalizability, we assessed nested cross-validated prediction performance metrics (AUROC/AUPRC: area under the receiver operating characteristic curve/precision-recall curve). We compared our model with Children's Oncology Group Long-Term Follow-Up Guidelines (COG LTFU, v6.0) recommendations indicating yearly dermatologic screenings following any RT or HCT.

**Results**: Among 23,172 survivors, 1,164 developed BCCs (63% occurred <40 years of age). The XGBoost model showed good discrimination (AUROC=0.76, 95% CI: 0.74-0.78) and precision (AUPRC=0.28, 95% CI: 0.24-0.32). The most important predictors included diagnosis age and cumulative doses of cranial/neck RT, alkylators, and anthracyclines. The cross-validated BCC prevalence by age 40 was 4%, 17%, and 59%, respectively, in low (<10%), medium (10-49%), and high ( $\geq$ 50%) model-defined predicted risk groups. In comparison, the COG LTFU-based classification prediction performance metrics were significantly worse (AUROC=0.63, P<0.001; AUPRC=0.10, p<0.001). The BCC prevalence by age 40 years among survivors recommended for screening per COG LTFU was 11%; 63% of them were re-classified as low risk using our XGBoost model.

**Conclusion**: Once validated in the St. Jude Lifetime Cohort, implementation of our subsequent BCC risk prediction models will more accurately inform risk-/age-based screening recommendations which are currently unavailable for this high-risk population.