# Survival Prediction for Pediatric Brain Cancer Patients

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Patients

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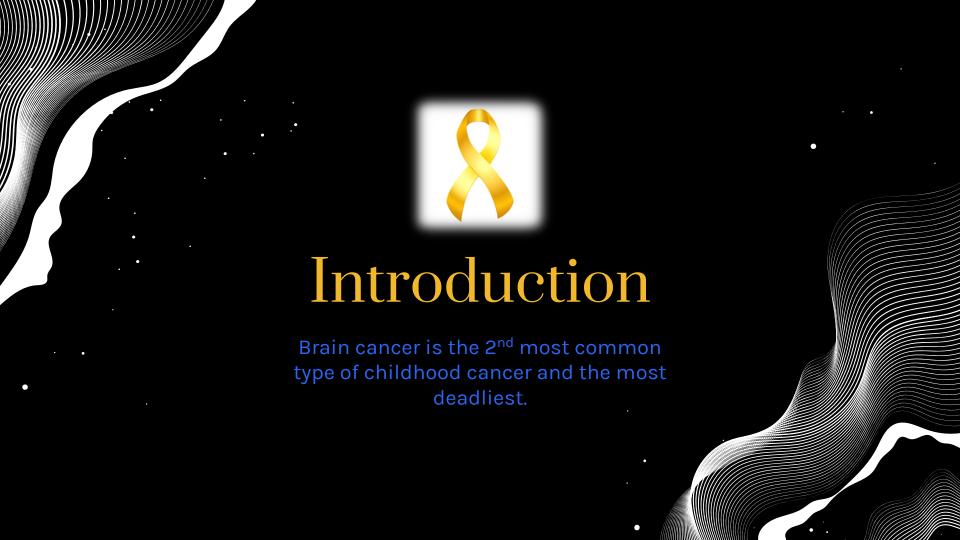
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## SEER Cancer Data



#### **Total Patients**

331,449,281 Total Cancer Patients



#### Variables

263 Variables available for analysis



#### **Data Collection**

SEER has been funded by the NCI since 1973. It collects cancer data that covers around 48% of the U.S. population.

# Patients Included in Analysis

**Total Number of Patients** 

12,164

X-Variables Considered

Age

Race

Gender

**Primary Site** 

Tumor Type

Grade

Laterality

Median Household

Income

Rural Urban Code

Y-Variable

Patient Survival

101 Patients removed due to unknown cause of death

Years

2000 - 2019

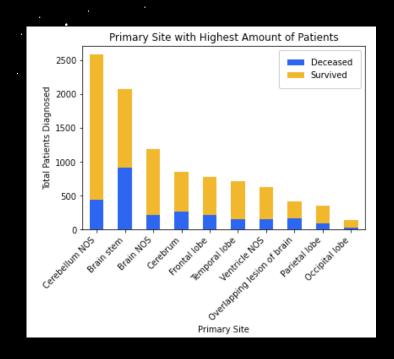
#### Deceased Survived 2000 32.59% 67.41% 2001 35.40% 64.60% 2002 35.60% 64.40% 2003 30.02% 69.98% 31.09% 2004 68.91% 2005 32.97% 67.03% 2006 29.45% 70.55% 2007 31.08% 68.92% 2008 31.05% 68.95% 2009 26.81% 73.19% 2010 25.12% 74.88% 2011 28.50% 71.50% 2012 24.75% 75.25% 2013 30.64% 69.36% 2014 25.60% 74.40% 2015 24.52% 75.48% 2016 23.57% 76.43% 2017 21.29% 78.71% 2018 15.92% 84.08% 2019 5.42% 94.58%

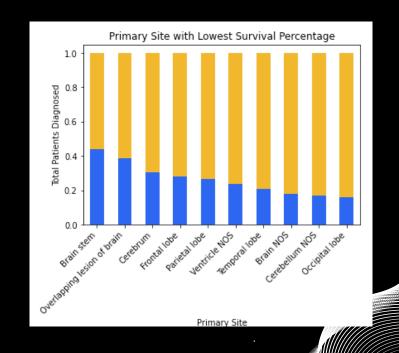
# Year Diagnosed

Reduced total Patients by removing those diagnosed before 2003 and after 2017.

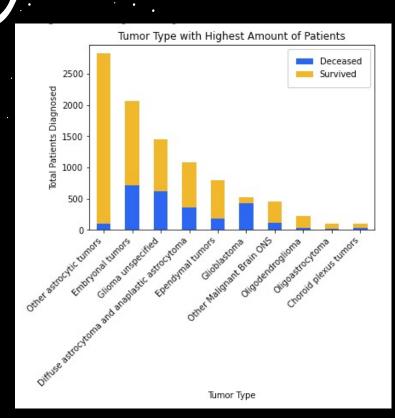
New total patients: 9,714

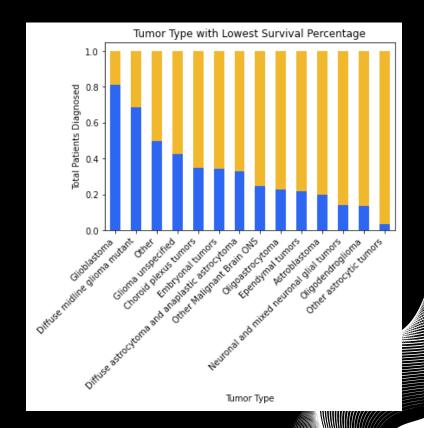
# Primary Site





# Tumor Type





# Race Origin Recode

|  | Deceased | Survived |
|--|----------|----------|
| Non-Hispanic Asian or Pacific Islander     | 33.05%   | 66.95%   |
| Non-Hispanic Black                         | 31.93%   | 68.07%   |
| Hispanic (All Races)                       | 30.87%   | 69.13%   |
| Non-Hispanic American Indian/Alaska Native | 24.32%   | 75.68%   |
| Non-Hispanic White                         | 23.91%   | 76.09%   |

Removed 102 Non-Hispanic Unknown Race patients as their survival rate was 96.08%.

## Grade

```
Unknown
Undifferentiated; anaplastic; Grade IV
Blank(s)
Moderately differentiated; Grade II
Well differentiated; Grade I
Poorly differentiated; Grade II
1.831045
```

Removed Grade variable because majority of patients had missing information.

# Age

Created bins for patients age ranging based on quartiles.

0-4

5-8

9-13

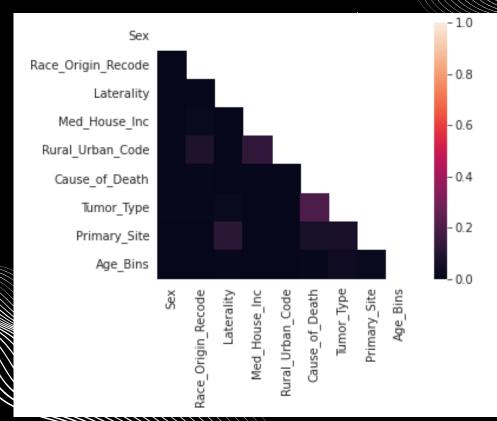
14 - 19

|   | Deceased | Survived |
|---|----------|----------|
| 1 | 30.63%   | 69.37%   |
| 2 | 30.07%   | 69.93%   |
| 3 | 24.02%   | 75.98%   |
| 4 | 23.29%   | 76.71%   |

Total Patients for Machine Learning: 9,611 Total X-Variables: 8



# Cramer's V Correlation Matrix





4 0.553822 0 0.257930 3 0.106604 2 0.073843 1 0.007800 Non-Hispanic White

Hispanic (All Races) Non-Hispanic Black

Non-Hispanic Asian or Pacific Islander

Non-Hispanic American Indian / Alaska Native

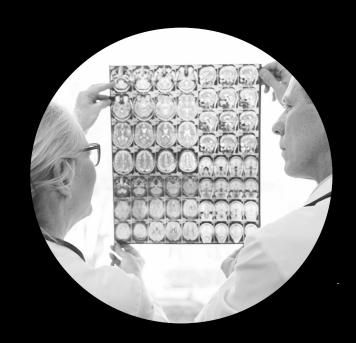


Survived: 5,608

Deceased: 2,080

Final Training Set: 11,216 rows 59 columns

Final Testing Set: 1,912 rows 59 columns



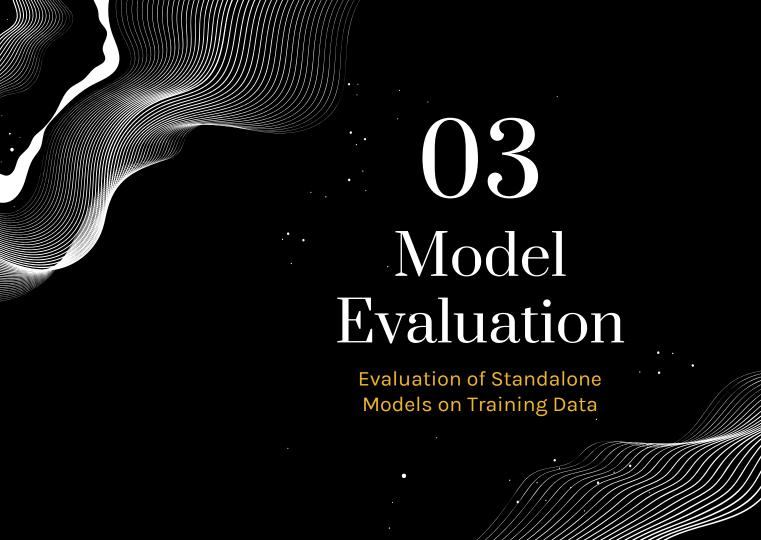
# Models

#### Classification Models Selected

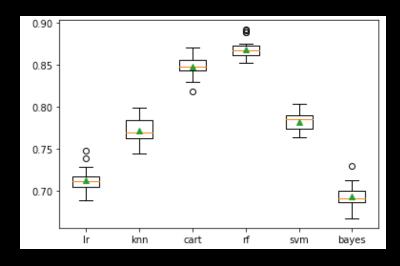
Logistic Regression
KNN
Decision Tree
Random Forest
SVC
Gaussian NB

Stratified K Fold with 10 splits









Random Forrest had the highest accuracy on the training data.

#### Accuracy

lr 0.713 (0.013) knn 0.772 (0.014) cart 0.849 (0.011) rf 0.869 (0.010) svm 0.783 (0.011) bayes 0.693 (0.013)



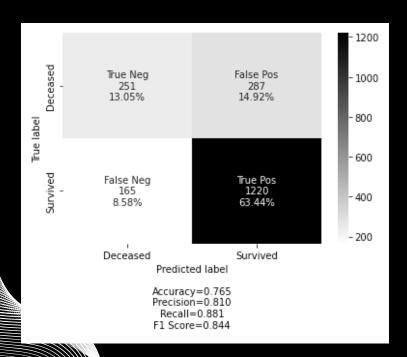


The stacking classifier function allows the strength of many classifiers to make a prediction. The output of each estimator is used as input of a final estimator.





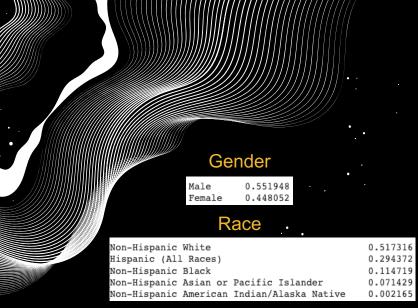
#### Results



Of the 1,923 patients, 1,471 were correctly classified.

287 of the incorrectly classified were labeled as surviving and 165 as deceased.

The model was better at identifying patients who survived than not labeling deceased patients as survived.



#### Rural Urban Code

|                          | 4 1331                              |          |
|--------------------------|-------------------------------------|----------|
| Counties in metropolitan | areas ge 1 million pop              | 0.608225 |
| Counties in metropolitan | areas of 250,000 to 1 million pop   | 0.192641 |
| Counties in metropolitan | areas of 1t 250 thousand pop        | 0.086580 |
| Nonmetropolitan counties | adjacent to a metropolitan area     | 0.062771 |
| Nonmetropolitan counties | not adjacent to a metropolitan area | 0.049784 |

#### Laterality

| Not a paired site                                     | 0.727273 |
|---|----------|
| Left - origin of primary                              | 0.121212 |
| Right - origin of primary                             | 0.121212 |
| Paired site, but no information concerning laterality | 0.010823 |
| Paired site: midline tumor                            | 0.008658 |
| Bilateral, single primary                             | 0.008658 |
| Only one side - side unspecified                      | 0.002165 |

# Distribution of the Variables from Patients Incorrectly Classified

Embryonal tumors

Astroblastoma

#### Tumor Type

#### Age Bins

| Bins_1 | 0.400433 |
|--------|----------|
| Bins_2 | 0.235931 |
| Bins_4 | 0.183983 |
| Bins_3 | 0.179654 |

| Glioma unspecified                             | 0.153680 |
|--|----------|
| Diffuse astrocytoma and anaplastic astrocytoma | 0.127706 |
| Ependymal tumors                               | 0.101732 |
| Other astrocytic tumors                        | 0.058442 |
| Other Malignant Brain ONS                      | 0.045455 |
| Glioblastoma                                   | 0.043290 |
| Choroid plexus tumors                          | 0.025974 |
| Oligodendroglioma                              | 0.015152 |
| Diffuse midline glioma mutant                  | 0.010823 |
| Oligoastrocytoma                               | 0.008658 |
| Other  | 0.006494 |
| Neuronal and mixed neuronal glial tumors       | 0.002165 |

0.398268

#### Median Household Income

| alali i i  | Cacci    | noid into | 71110                |
|------------|----------|-----------|----------------------|
| \$75,000+  |          | 0.285714  |                      |
| \$65,000 - | \$69,999 | 0.199134  |                      |
| \$60,000 - | \$64,999 | 0.164502  |                      |
| \$70,000 - | \$74,999 | 0.086580  |                      |
| \$50,000 - | \$54,999 | 0.075758  | <i>"//////</i> ///// |
| \$55,000 - | \$59,999 | 0.075758  |                      |
| \$45,000 - | \$49,999 | 0.058442  |                      |
| \$40,000 - | \$44,999 | 0.030303  |                      |
| \$35,000 - | \$39,999 | 0.015152  |                      |
| < \$35,000 |          | 0.008658  |                      |

Brain stem 0.272727 Cerebellum NOS 0.264069 Brain NOS 0.114719 Cerebrum 0.080087 Ventricle NOS 0.077922 Overlapping lesion of brain 0.054113 Frontal lobe 0.051948 Parietal lobe 0.043290 Temporal lobe 0.030303 Occipital lobe 0.010823



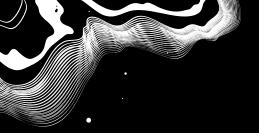


# Discussion Summary

The predicted results could be improved by analyzing which features contributed to incorrect classification by the model. It would be best to examine why the model incorrectly identified patients as survived when they did not.

Further analysis of the data could be used to predict the survival months of the patients who were classified as deceased.

Additionally, it would be beneficial to see which treatments had the best success with the different patient variables.



#### Conclusion

The model performed well with identifying patients who were predicted to survive brain cancer. It would be beneficial to have a model that performed better to identify patients who are less likely to survive so that doctors could aggressively treat the patients and additional research and analysis could be allocated to those types of cancers.



Hossain MJ, Xiao W, Tayeb M, Khan S. Epidemiology and prognostic factors of pediatric brain tumor survival in the US: Evidence from four decades of population data. Cancer Epidemiol. 2021 Jun;72:101942. doi: 10.1016/j.canep.2021.101942. Epub 2021 May 1. PMID: 33946020; PMCID: PMC8142618.

Is it time to use machine learning survival algorithms for survival and risk factors prediction instead of Cox proportional hazard regression? A comparative population-based study. Sara Morsy, Truong Hong Hieu, Abdelrahman M Makram, Osama Gamal Hassan, Nguyen Tran Minh Duc, Ahmad Helmy Zayan, Le-Dong Nhat-Nam, Nguyen Tien Huy medRxiv 2021.11.20.21266627; doi: https://doi.org/10.1101/2021.11.20.21266627

National Cancer Institute. (2022) Cancer Stat Facts: Childhood Brain and Other Nervous System Cancer (Ages 0-19). National Institutes of Health. <a href="https://seer.cancer.gov/statfacts/html/childbrain.html">https://seer.cancer.gov/statfacts/html/childbrain.html</a>



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