

# Predictive Analysis of Hospital Costs: A Comparative Study of Statistical Learning Techniques



Group 3  
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# Motivation

What influences a hospital's total operating costs?

- Amount of employees? Location? Number of inpatients?
- Collected data from Centers for Medicare and Medicaid Services with over 5000 hospitals during 2020 fiscal year
- Performed analyses to assess predictors' relative importance

# Exploratory Data Analysis

- Response total costs is very right-skewed
- Most relationships between predictors and response are linear
- Notable collinearity between several of the predictors
  - Example: salaries and number of employees

# Quantitative Response Results

- Most Important Predictors: Salaries and Number of Employees
- However, each predictor is associated with the response
- Random Forest has lowest test MSE, but test error is similar across all methods

# Simple Linear Regression Results:

method	cv_error	test_error	coef_est	p_value
Marginal LR number_of_beds	0.253	0.226	0.861	0.000
Marginal LR fte_employees_on_payroll	0.132	0.129	0.956	0.000
Marginal LR total_days	0.205	0.191	0.891	0.000
Marginal LR total_discharges	0.298	0.264	0.835	0.000
Marginal LR total_income	0.939	0.993	0.403	0.000
Marginal LR total_assets	0.731	0.409	0.553	0.000
Marginal LR salaries	0.103	0.229	0.979	0.000
Marginal LR inpatients	0.200	0.189	0.893	0.000
Marginal LR control_bin_Governmental	0.996	1.037	-0.025	0.480
Marginal LR control_bin_Proprietary	0.955	1.003	-0.441	0.000
Marginal LR provider_bin_Specialized	0.980	1.023	-0.337	0.000
Marginal LR rural	0.992	1.039	-0.129	0.000
Marginal LR duplicate	0.996	1.038	-0.121	0.188

# Comparison of MSE For All Methods

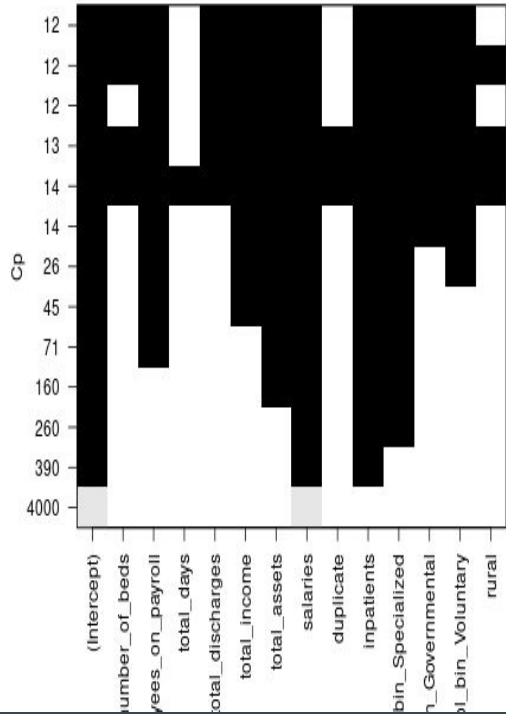
method	cv_error	test_error
Linear Regression (Main Effects)	0.071	0.106
Linear Regression (Transformations)	0.114	0.081
Regression Tree	0.144	0.100
Bagging	0.071	0.125
Random Forest	0.071	0.053
Boosting	0.092	0.079
Neural Network	0.072	0.088

# Variable Selection Results

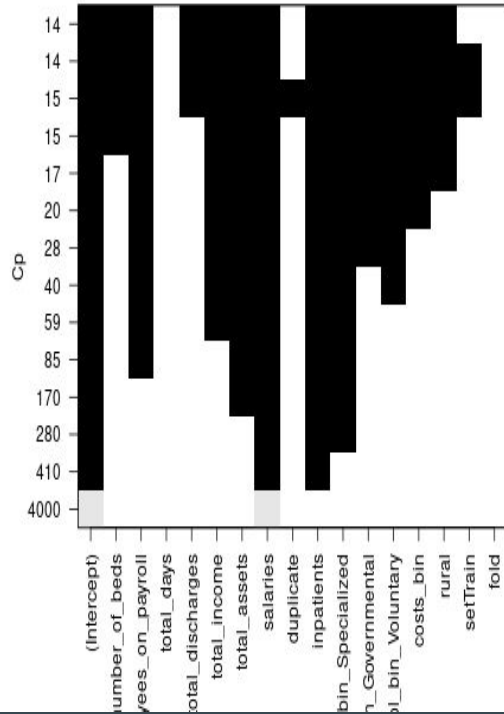
We used the regsubsets library

- Exhaustive
- Forward stepwise
- Backward stepwise
- In all cases, we got 8 variables as the optimal model size according to cross-validated test MSE.

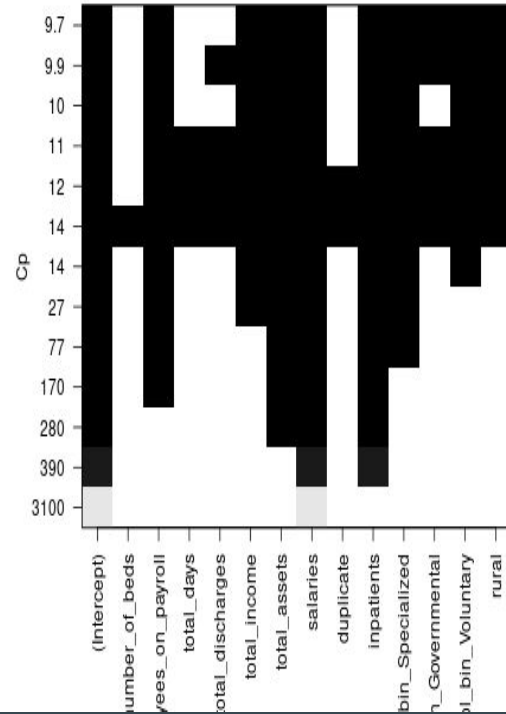
### Exhaustive



### Forward Stepwise



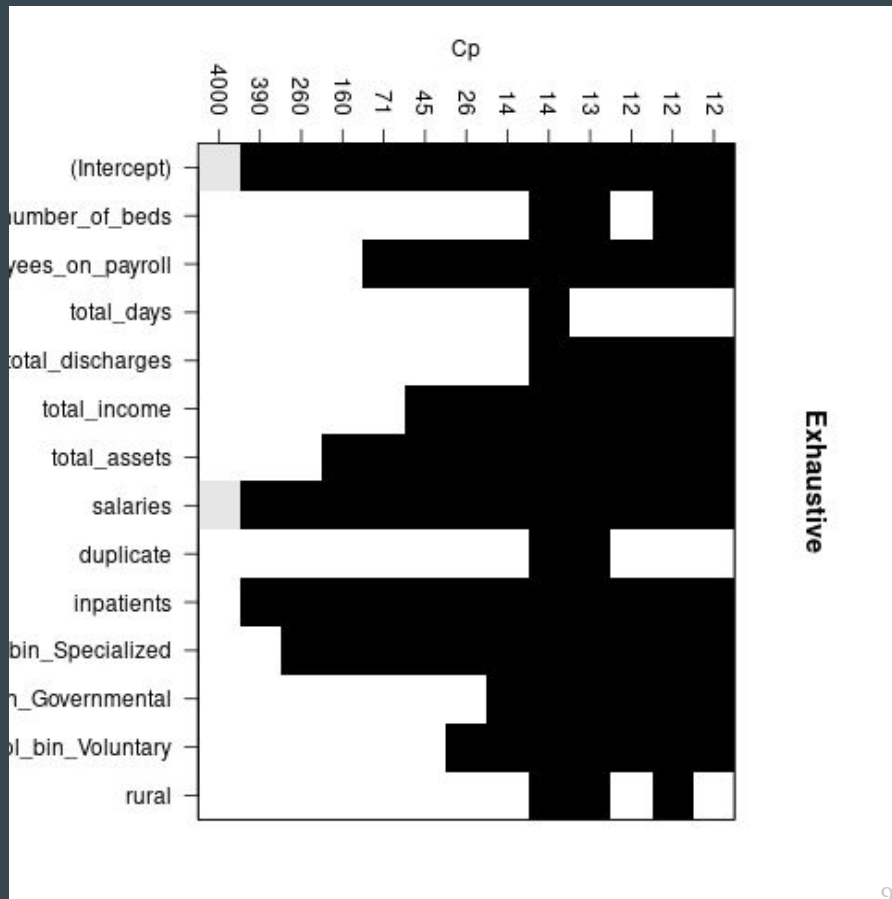
### Backward Stepwise



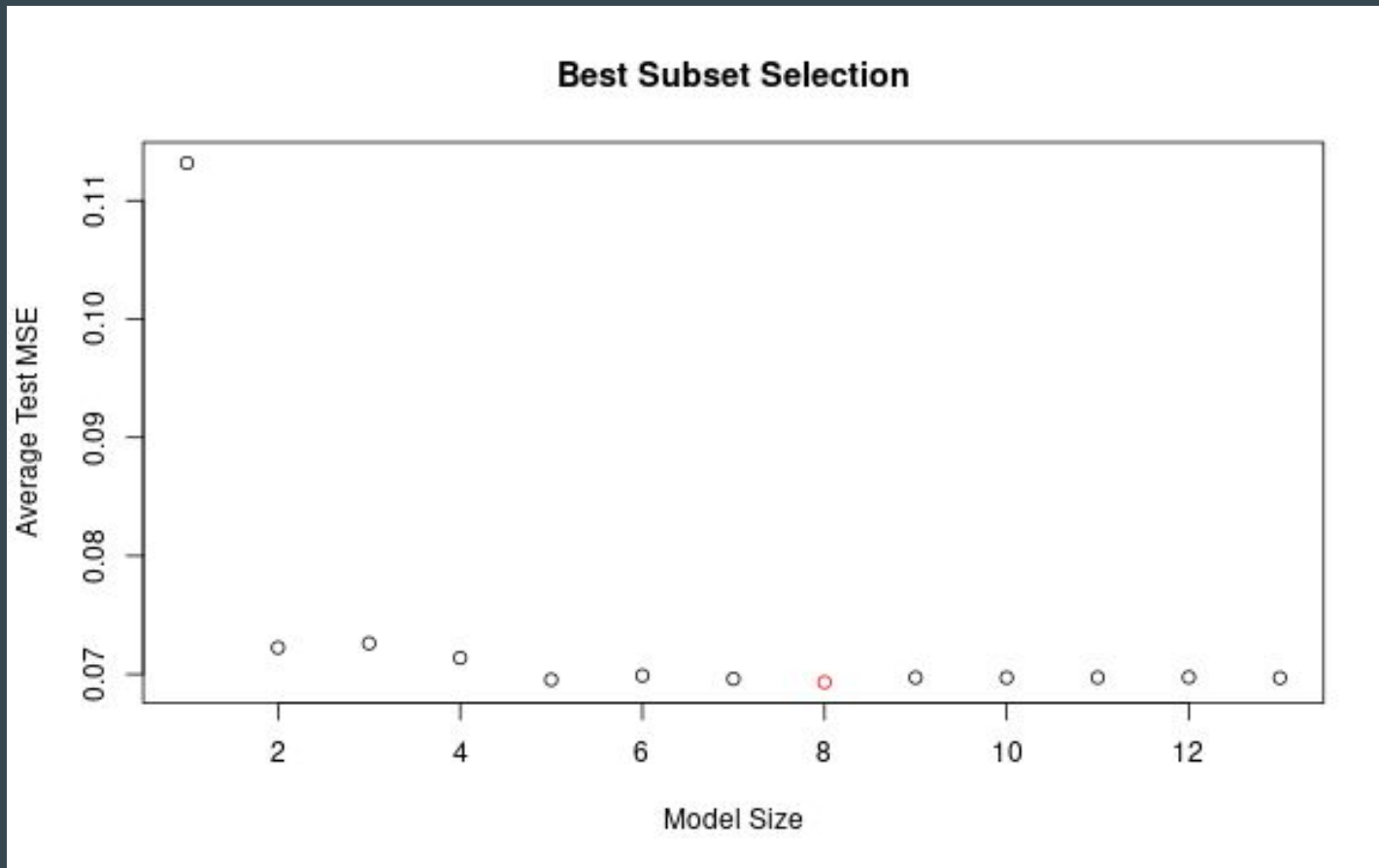


# Exhaustive

- Salaries, inpatients, and the dummy variable for whether a hospital is specialized or not seem to play a major role.
- Conversely, total days and the duplicate dummy variable in our dataset did not seem to play a major role in our models.



## 10-fold cross validated best model by smallest test MSE



# Lasso selection results

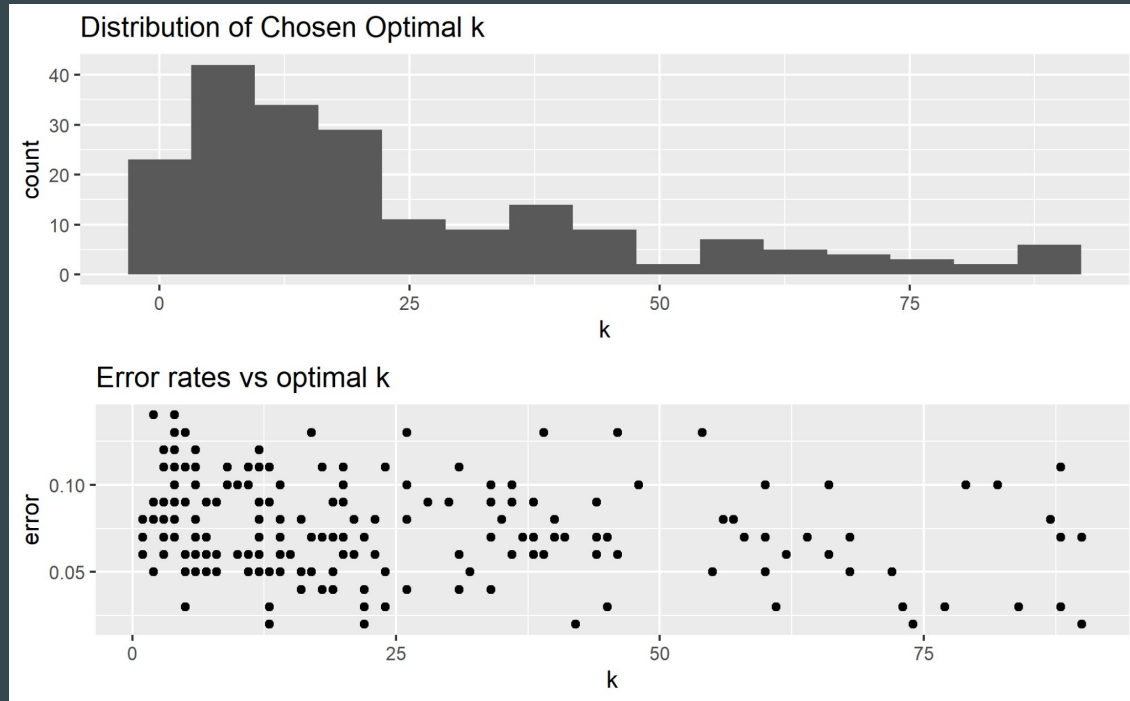
- Lasso selects 10 variables to be nonzero, dropping three in the process.
- This model size is somewhat close to the model size selected by our exhaustive and stepwise methods.
- The variables dropped are the duplicate dummy variable, the governmental dummy variable, and the number of beds.

# Qualitative Response Results

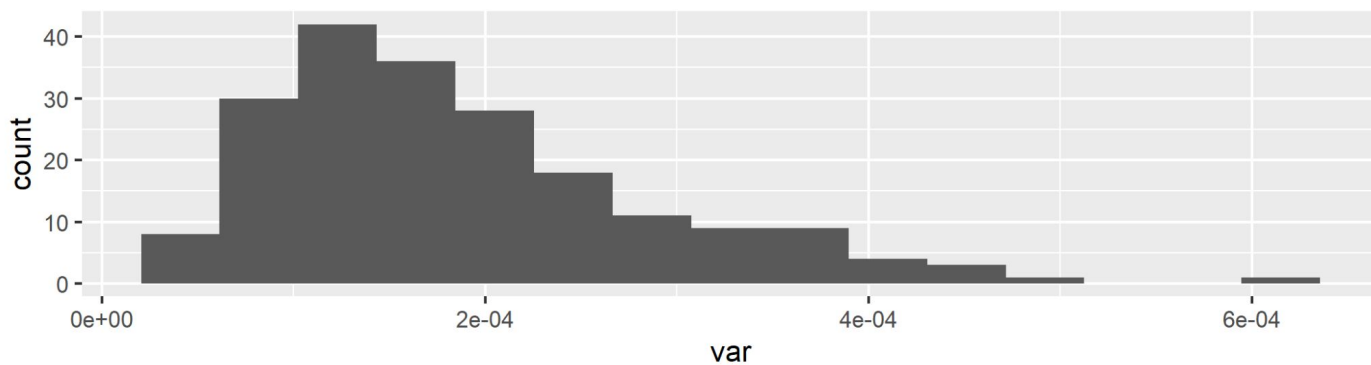
- Class labels: classified hospitals' as above/below the median total costs
- Salaries consistently the most important predictor
- Multiple logistic regression, bagging, and random forest produced similar misclassification rates around 3%
- High false negative rates
  - Difficulty identifying hospitals' with total costs above the median

# Simulation Study

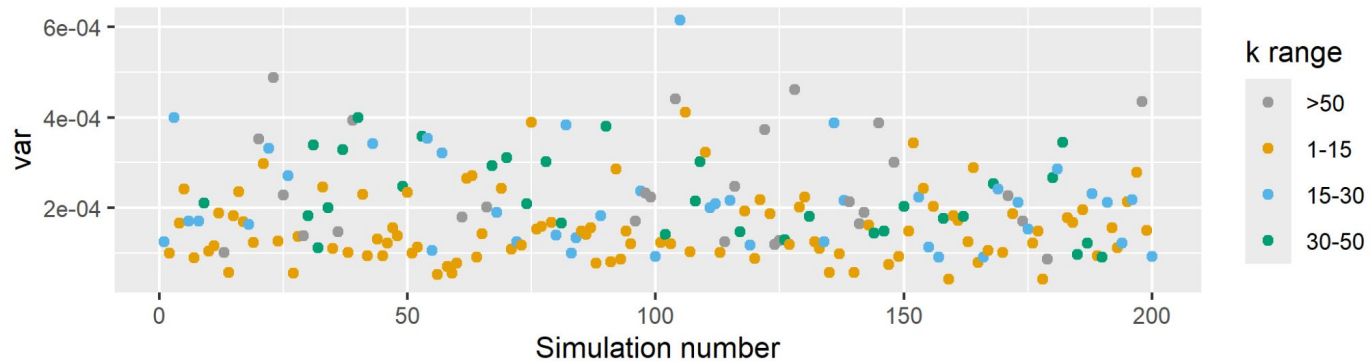
What is the optimal  $k$  in KNN with 200 simulated datasets?



### Distribution of Variances



### Error variance across k



# Conclusions

- Salaries is a very strong predictor for total costs, supported across many different analyses.
- Simple models don't perform that much worse than the more complex models.
- For KNN, error rate does not change very much for different values of  $k$ ; variance in optimal  $k$  is quite high.