Empower U.S. military members, veterans, and military spouses to succeed in the civilian workforce: Using an Analytical approach

Teradata Data Challenge

Team

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Non-Veteran Unemployment Rate 4.3%

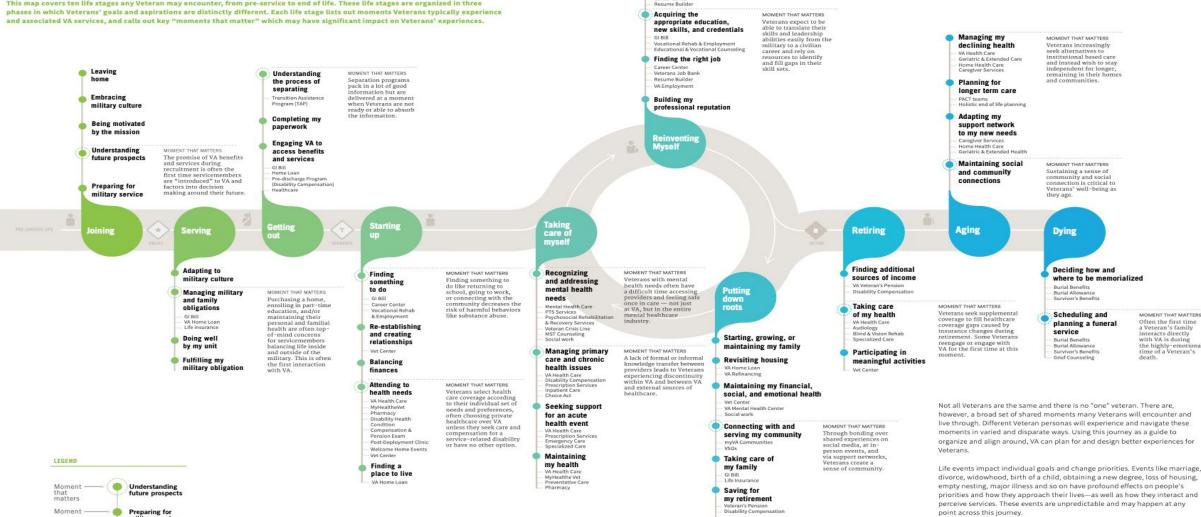


Women Veteran Unemployment Rate 2.7%

JOURNEYS OF VETERANS MAP

Life Stage

This map covers ten life stages any Veteran may encounter, from pre-service to end of life. These life stages are organized in three



Building my network Translating my military skills Skills Translator Career Center

> Engaging VA to access benefits and services

Pre-discharge Program

(Disability Compe Healthcare

GI Bill Home Loan Developed by the VA Veterans Experience Team.

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For more information contact Sarah Brooks.

LIVING CIVILIAN LIFE SERVING IN AND SEPARATING FROM THE MILITARY **RETIRING AND AGING**

Motivation

- United States is home to millions of Veteran military families
- Veterans offer a unique set of skills, experiences and leadership abilities
 - Translating these skills to the civilian workforce and marketing themselves to employers plays a major role
- HHUSA is an organization which is helping them to achieve stability in civilian life
- U.S. Department of Labor estimates that the military trains people in skills applicable in at least 962 civilian occupations*

How to get a veteran hired in the shortest span?

Out of the several challenging business questions that HHUSA provided, our focus was on the primary function of HHUSA which is nothing but getting a veteran hired

- Is there any relationship between the amount of time spent working with individual clients (time to complete an assessment, time to complete resume, # of logged activities, etc.) and how quickly are they employed?
- Is there a relationship between a client's demographic profile and getting hired?



What are we aiming to achieve of this reserach?

- To be able to provide targeted training to the unemployed veterans
- Improved hiring profiles can be established to target veteran with specific configurations of life, military, and work experience
- Tapping the pockets of filled with military talent and unique skills
 - This solution solves the moral imperative and becomes a competitive business advantage



Dataset

Description

- Data is provided by Hire Heroes USA
- Non Profit Organization
- Empowers U.S. military members, veterans, spouses
- Goal is to support military personnel to succeed in civilian workforce

Datasets

- Salesforce CRM structure
- Data provided includes 13 spreadsheets
- To answer our questions we focused on:
 - Contacts
 - Activities
 - Hired Information

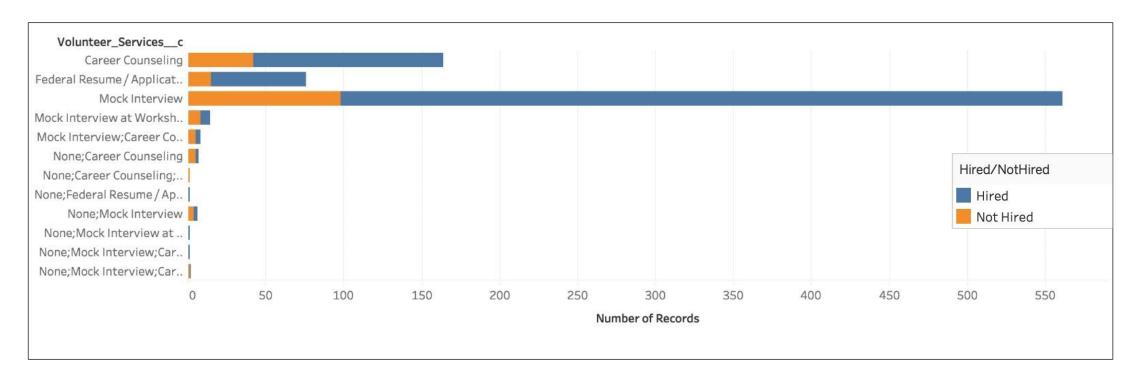


Summary

Variable	Number of Missing Values	% of Missing Values
Gender	1046	7.60
Last Rank	12685	92.26
Race	10750	78.19
Additional Service Needs	13606	98.96
Status	1075	7.81



Mow Volunteer Services influence Hiring







Approach

Explore

- Identified three most critical datasets
- Summarized the data to understand better
- Data imbalance on predictor variable
- Identified sparsely populated variables
- Identified 60 predictor variables

Prepare

- Joined identified datasets
- Derived predictor variables
- Created data dictionary for text data
- Imputed missing values using MICE
- Created train data without imbalance

Predict

- Classification model using Random Forest
- Evaluated the models using AUC
- Cross validation and hyperparameter tuning
- Used Boruta for feature elimination
- Zeroed-in on 32 predictor variables
- Conducted survival analysis based on variable importance

Dataset	Contact	Hire Info	Activities
# of Categorical variables	264	17	38
# of Continuous variables	0	1	7
# of Textual data variables	42	8	3
# of Unique identifier columns	11	2	12
# of Data / Time variables	74	7	9
Total # of Columns	391	35	69
# of Columns with > 35% 'NA' / Spaces	198	8	6
# of Highly imbalanced columns	17	3	4
# of Variables considered for prediction	36	2	4

Explore

- Data Filters
 - Indicates the account is that of a job seeker
 - Active Color Status of client
 - **Eliminated Black & Blanks**
 - If the account is active or not
- Identified the date and textual columns that can be effectively used
- Dropped columns with > 35% missing values
- Columns with heavy data imbalance were ignored
- Target variable is imbalanced Hired: 30%, Not Hired: 70%

<u>Prepare</u>

- Derived target and predictor variables
 - Days to Hire a Hired Date Record Created date
 - Resume Complete Duration
 - Initial Assessment Complete Duration
 - Number of Activities
- Data Dictionary Created data dictionary for fields such as Highest Level of Education, Desired Industry for Employment
- Imputation Columns with missing values < 35% was imputed using MICE with 5 iterations
- Under sampling was done on train dataset to make data balanced.
 Test dataset is left untouched

Predict

- Selected Random Forest classification model after comparing different classification models
- Hyper parameter tuning and Cross validation
- Models were evaluated using AUC
- Three variables were removed based on Boruta
- Final predictor variables 35
- Survival analysis was conducted on 5 variables that influences model the most

Some of the variables which were found statistically significant using Logistic Regression & Boruta

Variable	Variable Description
Gender	Gender of client
Number of Activities	Count of logged activities per client
Highest_Level_of_Education_Completed_c	Education attained
Willing_to_Relocatec	Willingness to relocation
Resume_Tailoring_Tipsc	Client acquired support to tailor resume
HHUSA_Workshop_Participantc	Client is participant of HH USA training workshop

Algorithm Employed	Parameter/Hyperparameters	Performance Measures
Random Forest	NA	AUC = 0.65, Accuracy = 57%
Logistic Regression	NA	AUC = 0.64, Accuracy = 55%

 Data Used
 Target Variable: Hired/NotHired
 Predictors Count : Around 400

Algorithm Employed	Parameter/Hyperparameters	Performance Measures	
Naive Bayes (Baseline)	10-fold Cross Validation	AUC = 0.8341	
Neural Networks	size=3, maxit=1000	AUC = 0.8891	
GLM Net	3-fold Cross Validation	AUC = 0.9053	
GBM	3-fold Cross Validation	AUC = 0.9056	
Random Forest	Default Values	AUC = 0.9195	

 Data Used
 Target Variable: Hired/NotHired
 Predictors Count : 32

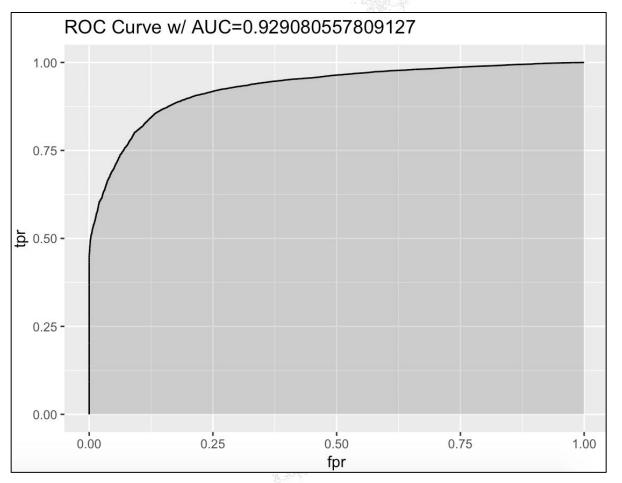
Random Forest Variants (Parameter/Hyperparameters)	Performance Measures
Fine-tuning (120 trees, 5 nodes)	$\mathbf{AUC} = 0.9182$
10 fold Repeated Cross-Validation, mtry =7, num.tree = 120	AUC = 0.9211
10- fold Repeated Cross-Validation (repeats 3 times) splitrule = gini and min.node.size = 3, metric = ROC, mtry =8 (using Tune grid)	AUC = 0.9263

Data Used	Target Variable: Hired/NotHired	Predictors Count: 32	

Final AUC & Classification Matrix

	Reference		
		Hired	Not Hired
Prediction	Hired	5713	1473
	Not Hired	1559	14136

	Requested_Supportc, Open_duration, Resume_Complete_Duration, Num Activities,
Top Features	Initial_Assessment_Complete_Duration, Genderc, Mileage_Willing_To_Commutec, Used_Volunteer_Servicesc, MyTrak_VTS_Assignedc,



Results: Important features

Staff_Assigned	Hired	Not Hired
No	31%	68%
Yes	76%	23%

Reason_for Requested Assistance (top 3)	%age of total requests	Hired	Not Hired
Mock Interviews	around 30%	66%	33%
Industry Specific Job Search	around 26%	72%	27%
LinkedIn Review	around 15%	61%	38%

Results: Important features

• Classification: Hired = If hired in <180 days, else Not-hired

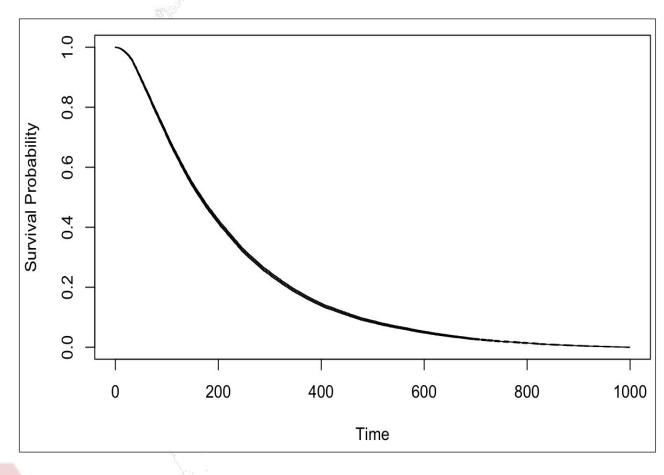
Feature	Impact	Description
Number of Activities	Positive	The larger the number of activities performed, the greater the chance of being hired
Transition Specialist Assigned	Positive	If a specialist is assigned, then a greater chance in being hired than when a specialist is not assigned
Finalized Revised Resume on File	Negative	If a final resume is not on file, then prob of getting hired decreases than when a resume is on file
Initial Assessment Complete Duration	Negative	The longer the duration in completing the initial assessment, the lower the chances of being hired
Resume Tailoring Tips	Negative	If client did not receive resume tailoring tips, then a lower chance of being hired than when client did receive resume tailoring tips
HHUSA Workshop Participant	Positive	If client received attended HHUSA workshop, then a greater chance of being hired than when client did not attend HHUSA workshop

Results: Demographic features

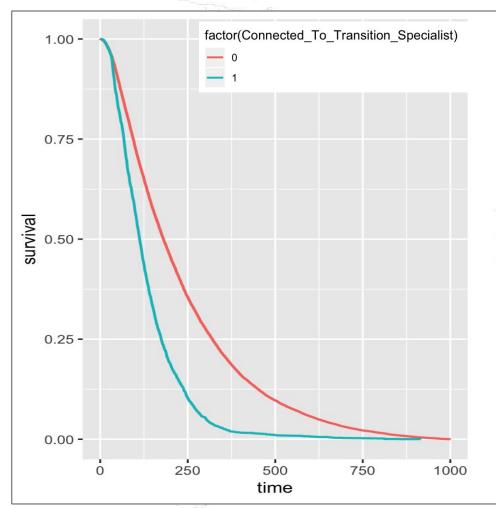
• Classification: Hired = If hired in <180 days, else Not-hired

Feature	Impact	Description
Gender**	Positive	A male client has a higher probability of getting hired than a female client
Race	Not Significant	Cannot Interpret
Highest Level of Education**	Positive	A client with a master degree has a higher probability of getting hired than a client without any degree
Disability > 60%**	Negative	A person with a disability of >60% is less likely to be hired than a person without a disability

Results: Hire Heroes Impact



Overall Survival Probability



Recommendations

- Interact more..!
- Assign a transition specialist at the earliest
- Assist client with resume tailoring tips to have a completed resume on file
- Encourage all clients to attend HHUSA workshop
 - Mock Interview top request from client
- Host employment bootcamps



Future work

- Have a framework to increase the completeness and correctness of data capture
- Analyze some of the critical fields such as Rank, Employed Position, Skillset of hired veterans to aid the unemployed veterans
- Analyze the funding capabilities and its influence
- Personalized career coaching to the client on site (as of now most of the training is virtually offered to clients)



