## Data Mining Techniques Applied to Urban Terrain Command \& Control Experimentation

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## Previous Work

- Simulated Southwest Asia scenario using One Semi-Automated Forces (OneSAF) Testbed Baseline (OTB) for 228 plays of the same battle
- Used a Killer/Victim Scoreboard (KVS) to collect metrics [143 metrics per three time slice]
- Used statistical data mining approaches to relate battle outcome to metrics
- Concluded: great potential for identifying key metrics in the battle worth tracking and/or for suggesting course of action changes


## Scenario



## OneSAF Modification Killer/Victim Scoreboard

- Firer and Target Identity and Location

Time Stamp 1010070890
Vehicle ID 1076
Firer ID 1087
Projectile 1143670848

- Type of Ammo
- Range
- Outcome

Firer Position: $X=220217.00 \quad Y=146765.00 \quad Z=12.37$
Target Position: $X=222454.38 \quad Y=149117.80 \quad Z=9.99$
Vehicle 1076: Hit with 1 "munition_USSR_Spandrel" (0x442b0840)
Comp DFDAM_EXPOSURE_HULL, angle 19.53 deg Disp 0.889701 ft
Kill Thermometer is: Pk:1.00, Pmf:1.00, Pf:0.90, Pm:0.80 Pn:0.80
RANGE 3246.773576
$r=0.990835$ kill_type $=\mathrm{MF}$

| 1076 | 100A41 vehicle_US_M1 |
| :--- | :--- |
| 1087 | 100A23 vehicle_USSR_BMP2 |

## Analysis



## Method Comparison

Percent Correct Classification by Stopping Time and Method

| Stopping <br> Time (min) | Discriminant <br> Analysis | CART | Logistic <br> Regression |
| :---: | :---: | :---: | :---: |
| $51 / 2$ | $70 \%$ | $70 \%$ | $69 \%$ |
| 10 | $80 \%$ | $75 \%$ | $74 \%$ |
| 20 | $85 \%$ | $82 \%$ | $85 \%$ |

## Current Experiment

- Change terrain to urban
- Involve Dismounted Infantry (DI)
- Use Dismounted Infantry SemiAutomated Forces (DISAF) Simulation Software
- Develop urban scenario


## DISAF Challenges

- Compiling - need an older version of GNU C and C++ (version 2.91.66)
- KVS Code developed for OneSAF at ARL did not easily insert into DISAF
- Fireteams tend to move better when tasked as individual rather than as a team
- However this breaks down for the "clear room" task which requires a full fireteam
- Vehicles tend to not enter the city sector
- DI entities at times get stuck in buildings and tunnels
- If entities can not determine the proper route they go to the bottom of the terrain


## Current Urban Scenario

Location: City sector based on the McKenna MOUT (Military Operations on Urbanized Terrain) site

## Scenario: Attack Phase I

- Isolate area, three-pronged encirclement to reduce threat forces from perimeter
- Carried out by 2 M2s from the North and 2 M2s from the Southwest and a headquarters (HQ) attachment of 2 M1A1s and 1 M 2 entering from the West
- Initial resistance from 3 BMPs and 2 T-80s around the perimeter, a $3^{\text {rd }} \mathrm{T}-80$ is in the center flanking the objective






## Current Urban Scenario

## Scenario: Attack Phase 2

- Eight fireteams (FT) enter sector behind armored vehicles
- Carried out by 3 FTs from both the North and from the Southwest
- Northern teams clear separate buildings (1 \& 2) and continue on to secure objective
- Southwestern teams clear separate buildings (3 \& 4) and continue to objective
- Two Western FT (HQ) proceed directly to the objective


## Current Urban Scenario

## Scenario: Attack Phase 2

- Interior resistance provided by opposition DI in the five critical buildings and also in a key vantage point building on the Northwest side of the sector
- Three additional opposition DI stationed outside buildings 1 , 2, and the objective




## Simulation Data

## Predictors

- 444 variables, but only 75 runs so far
- Two time slices ( 372 seconds and 480 seconds)
- Hits taken by Blue and hits by Blue fire involving all relevant vehicles, fireteams, and buildings
- Status of all entities

Responses

- Taking the objective (1 or 0)
- Establishing a foothold in the city ( $0,1,2,4$ )
- MOUTscore ( 0 to 8 with buildings under control with minimum casualties)


## Potential Analytical Methods

- Discriminant Analysis
- Cart
- Logistic Regression
- Multiple Regression
- Neural Networks
-Dr. Barry Bodt babodt@arl.army.mil


## A Discriminant Model

| 㩌Data: Discriminant Function Analysis Summary (FINALMOUTEX1b) ${ }^{\text {k }} \quad-\quad \mathbf{\square}$ |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| N=75 | Discriminant Function Analysis Summary (FINALMOUTEX1b) Step 7, N of vars in model: 7; Grouping: FootHold (4 grps) Wilks' Lambda: . 32024 approx. $F(21,187)=4.3384 p<.0000$ |  |  |  |  |  |  |
|  | Wilks' <br> Lambda | Partial Lambda | $\begin{gathered} \hline \text { F-remove } \\ (3,65) \\ \hline \end{gathered}$ | p-level | Toler. | 1-Toler. (R-Sqr.) |  |
| F8TS2 | 0.488062 | 0.656143 | 11.35459 | 0.000004 | 0.910182 | 0.089818 |  |
| A23TS2 | 0.380200 | 0.842291 | 4.05682 | 0.010497 | 0.921236 | 0.078764 |  |
| RTA11TS2 | 0.358180 | 0.894073 | 2.56701 | 0.062007 | 0.685511 | 0.314489 |  |
| BMC12TS2 | 0.374920 | 0.854151 | 3.69965 | 0.016001 | 0.877811 | 0.122189 |  |
| F5TS2 | 0.354554 | 0.903217 | 2.32167 | 0.083316 | 0.967066 | 0.032934 |  |
| M3T13FS2 | 0.344559 | 0.929415 | 1.64548 | 0.187509 | 0.919428 | 0.080572 |  |
| M2A21TS2 | 0.341595 | 0.937480 | 1.44494 | 0.237843 | 0.720540 | 0.279461 |  |
| 14 |  |  |  |  |  |  |  |

## Root Means



Group Separation


## Group Separation

Root 1 ws. Root 3


捣 Root 1 vs. Foot 2 駧 Root 1 vs. Root 3

## Group Separation



## Classification Efficiency

| Group | Classification Matrix (FINALMOUTEX1b) Rows: Observed classifications Columns: Predicted classifications |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Percent Correct | $\begin{gathered} G-10 \\ p=.21333 \end{gathered}$ | $\begin{gathered} G-2: 1 \\ p=.06667 \end{gathered}$ | $\begin{gathered} G-3: 2 \\ p=.18667 \end{gathered}$ | $\begin{gathered} G-4: 4 \\ p=.5333 \end{gathered}$ |  |
| G_1:0 | 75.00000 | 12 | 0 | 0 | 4 |  |
| G_2.1 | 60.00000 | 0 | 3 | 1 | 1 |  |
| G_3:2 | 21.42857 | 2 | 0 | 3 | 9 |  |
| G_4:4 | 92.50000 | 2 | 1 | 0 | 37 |  |
| Total | 73.33334 | 16 | 4 | 4 | 51 | - |
| $\square$ |  |  |  |  | - |  |

頃 Classification Matrix (FINALMOUTEX1b)

## Metric Description

F8TS2-Status of FT 8 (B4) at TS2
F5TS2-Status of FT 5 (A6) at TS2
M3T13FS2-Hits by M2-3 (B11, B12) at A13-T80 at TS2
A23TS2-Status of A23-DI at TS2
BMC12TS2-Status of M2 C12 at TS2
RTA11TS2-Status of T80 A11 at TS2
M2A21TS2-Hits taken by M2-2 (A16) by A21-DI at TS2



Select an item to edit


## Metric Interpretation

FT8 important because mission took it first to Bldg 3 and positioning kept it away from being targeted by Bldg 5

FT5 important because after passing Bldg 5, would move South away from Red DI in approach to objective

M3T13FS2-Hits on T80 - Southwest approach
A23TS2-Status of A23-DI - outside Building 2
BMC12TS2-Status of M2 - Success of North approach
RTA11TS2-Status of T80 on North
M2A21TS2-Hits taken by M2-2 - West approach

## Acknowledgements

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Without their help, we would never have compiled DISAF!

## Directions/Conclusion

- More involved and sophisticated opposition force
- Changes to KVS
- Overall improved scenario
- Establish data from more time periods to provide information on battle progression
- Collect information from a larger number of battles
- Use a greater variety of statistical tools, to include work in the microarray arena
- Data mining combat simulations holds great promise for understanding battles if one believes the simulations and statistical methods will continue to improve.


## Questions ???

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Zoom: click middle to zoom in around point; click right to zoom out around point; click and drag middle to set screen area

## Discriminant Analysis

- Maximizes $\left|a^{\prime}\left(\overline{\mathrm{x}}_{1}-\overline{\mathrm{x}}_{2}\right)\right|$ s.t. $a^{\prime} S a=1$
- Assumes multivariate normal predictors with common covariance matrix $\Sigma$ but different mean vectors $\mu_{1}$ and $\mu_{2}$



## Standardized Coefficients



## Testing Roots

| Roots Removed | Chi-Square Tests with Successive Roots Removed (FINA |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Eigenvalue | Canonicl R | Wilks' Lambda | Chi-Sqr. | df | p-level |  |
| 0 | 0.791787 | 0.664754 | 0.320239 | 78.00018 | 21 | 0.000000 |  |
| 1 | 0.568677 | 0.602097 | 0.573799 | 38.05006 | 12 | 0.000151 |  |
| 2 | 0.110980 | 0.316060 | 0.900106 | 7.20912 |  | 0.205546 |  |
| 1 |  |  |  |  |  | - |  |
| 闃 Chi-Square Tests with Successive Roots Removed (FINALMOUTEX1b) |  |  |  |  |  |  |  |

FT3 (Y) by FT7 (X) by Foothold Frequency


FT5 (Y) by FT8 (X) by Foothold Frequency


