ADM Analysis

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```
import numpy
           from IPython.display import display, HTML
           %matplotlib inline
In [704]: #Globals
           #Directory and statistic locations
admStats = "/active_duty_military_DetailStats.csv"
           cwopStats = "/civilian_workers_on_post_DetailStats.csv"
           admFile = "active_duty_military"
           cwopFile = "civilian_workers_on_post"
           admControlDir = "Control/active_duty_military_Individual_Mean_Stats/"
admDir = "/active_duty_military_Individual_Mean_Stats/"
           cwopControlDir = "Control/civilian_workers_on_post_Individual_Mean_Stats"
           cwopDir = "/civilian_workers_on_post_Individual_Mean_Stats"
           #Population sizes
           admCount = 27663
           cwopCount = 15912
            #Interventions of note
           #ve70vtd17milworkers/ate25at130ap130/sq90sqg25sqtd10sq128', u've70vtd38milworkers/ate2
           strongImpStrongEffect = "ve70vtd17allonpost/ate87at130ap130/sq90sqg25sqtd10sq128"
           strongImpWeakEffect = "ve30vtd17allonpost/ate25at130ap130/sq90sqg25sqtd10sq128"
           weakImpStrongEffect = "ve70vtd38active/ate87atl10apl30/sq90sqg25sqtd10sq17
weakImpWeakEffect = "ve30vtd38active/ate25atl10apl30/sq90sqg25sqtd10sq17"
           #Variables of note
           controlled = ['v','vtd','atdr','atti','attd','atl','apl','sq','sqg','sqtd','sql']
           unused = ['v','atdr','atti','attd','sq','sqg','sqtd']
natural = ['ve','ate']
experimental = filter(lambda x:x not in unused, controlled)
           epistats = ['attackRate','peakDay','peakNumber']
            #Percentage text formatting
           def percent (number):
                return "{0:.3f}%".format(number * 100)
In [705]: #Loader functions
            #Pulls attack rate from dataframe
           def getAR (meanData, cell):
```

return meanData.ix[cell]['attackRate']

def loadSingleCell(folder, subpop, cell, label):

#Loads curve for target cell

In [703]: import pandas as pd

```
fileIn = folder + '/' + subpop + '_'*(len(subpop) > 0) + 'MeanPlots.csv'
                        data = pd.read_csv(fileIn, index_col='index')
                        toReturn = data.ix[cell]
                        toReturn.name = label
                        return toReturn
                  #Loads whole experiment from GetSlice.py output
                 def getExperiment(experimentCSV):
                        data = pd.read_csv(experimentCSV, skipinitialspace=True, index_col='directory')
                        means = data[data['iteration'] == 'Mean'].replace(-1,NaN).replace('Mean',NaN).dropna iterations = data[data['iteration']!='Mean'].replace(-1,NaN).replace('Mean',NaN).dropna iterations = data[data['iteration']!='Mean'].dropna iterations = data[data['iteration
In [706]: #Stats derivation
                 def getCorrelations(iterations, natural, experimental):
                        attackNatural = iterations[natural].corrwith(iterations['attackRate'])
                        numberNatural = iterations[natural].corrwith(iterations['peakNumber'])
                        dayNatural = iterations[natural].corrwith(iterations['peakDay'])
                        attackExperimental = iterations[experimental].corrwith(iterations['attackRate'])
                        numberExperimental = iterations[experimental].corrwith(iterations['peakNumber'])
                        dayExperimental = iterations[experimental].corrwith(iterations['peakDay'])
                        naturalStats = pd.concat([attackNatural, dayNatural, numberNatural], join='outer',
                        experimentalStats = pd.concat([attackExperimental, dayExperimental, numberExperime
                        naturalStats.columns = experimentalStats.columns = ['attackRate','peakDay','peakNu
                        print "Experimental Variable v. Epistat Correlation coefficients:\n"
                        print "Pharmaceutical Effectiveness\n", naturalStats
                        print "\nIntervention Sequence\n", experimentalStats,'\n\n'
                        return {'naturalStats':naturalStats,'experimentalStats':experimentalStats}
In [707]: #Mean intervention efficacy maxima and minima detection
                 def getMaxima(means):
                        weakest = means.ix[means['attackRate'].idxmax()]
                        strongest = means.ix[means['attackRate'].idxmin()]
                        weakRef = weakest.name; strongRef = strongest.name
weakest.name = 'Weakest Intervention'; strongest.name = 'Strongest Intervention'
                        maxMin = pd.concat([weakest, strongest], axis= 1)
                        print "Attack Rate Extrema by Intervention Parameters\n\n", maxMin,'\n\n'
return {'weakRef':weakRef,'strongRef':strongRef,'maxMin':maxMin}
In [708]: #Plot best, worst, and control curves
                 def getCurves(expDir, weakRef, strongRef, admFile, end):
                        admBestCurve = loadSingleCell(expDir,admFile,strongRef,'Strongest')[0:end]
                        admWorstCurve = loadSingleCell(expDir,admFile,weakRef,'Weakest')[0:end]
                        seedNumber = str(admBestCurve[1])
                        admControlCurve = loadSingleCell('Control','','seeds' + seedNumber,'Control')[0:en
                        admCurves = pd.DataFrame(zip(admBestCurve,admWorstCurve,admControlCurve), columns=
                        cumsumCurves = admCurves.cumsum()
                        admCurves.plot()
                        cumsumCurves.plot()
In [709]: #Find percent attack rate
                 def getRankedInterventions(ADMMeanStats, CWOPMeanStats):
                        SISE_ADM_AttackRate = getAR(ADMMeanStats, strongImpStrongEffect)
                        SISE_CWOP_AttackRate = getAR(CWOPMeanStats, strongImpStrongEffect)
```

```
WISE_ADM_AttackRate = getAR(ADMMeanStats, weakImpStrongEffect)
               WISE_CWOP_AttackRate = getAR(CWOPMeanStats, weakImpStrongEffect)
               WIWE_ADM_AttackRate = getAR(ADMMeanStats, weakImpWeakEffect)
               WIWE_CWOP_AttackRate = getAR(CWOPMeanStats, weakImpWeakEffect)
               SISE_ADM_Percent = float(SISE_ADM_AttackRate) / admCount
               SISE_CWOP_Percent = float(SISE_CWOP_AttackRate) / cwopCount
               SIWE ADM Percent = float(SIWE ADM AttackRate) / admCount
               SIWE_CWOP_Percent = float(SIWE_CWOP_AttackRate) / cwopCount
               WISE ADM Percent = float (WISE ADM AttackRate) / admCount
               WISE_CWOP_Percent = float(WISE_CWOP_AttackRate) / cwopCount
               WIWE_ADM_Percent = float(WIWE_ADM_AttackRate) / admCount
               WIWE_CWOP_Percent = float(WIWE_CWOP_AttackRate) / cwopCount
               print "Theoretical strongest and weakest pharmacuetical intervention (PI) & logist
               print "\nActive Duty Military on base:", admCount
               print "Civilian Workers on post", cwopCount
               print "\n\t\tActive Duty\tCivilian Workers"
               print "Strong PI, Strong LI\t", percent(SISE_ADM_Percent), '\t', percent(SISE_CWOP
print "Weak PI, Strong LI\t", percent(SIWE_ADM_Percent), '\t', percent(SIWE_CWOP_P
print "Strong PI, Weak LI\t", percent(WISE_ADM_Percent), '\t', percent(WISE_CWOP_P
print "Weak PI, Weak LI\t", percent(WIWE_ADM_Percent), '\t', percent(WIWE_CWOP_Percent))
               return 'null'
In [710]: #Main action
          def pullExperiment (expDir, end):
               text = 'Experiment ID: ' + expDir + '
                                                             Length: ' + str(end)
               display (HTML("<b>" + text + <math>"</b>"))
               admExperimentStats = getExperiment(expDir+admStats)
               cwopExperimentStats = getExperiment(expDir+cwopStats)
               correlationData = getCorrelations(admExperimentStats['iterations'], natural, experim
               maximaData = getMaxima(admExperimentStats['means'])
               curves = getCurves(expDir, maximaData['weakRef'], maximaData['strongRef'], admFile, en
               ranks = getRankedInterventions(admExperimentStats['means'], cwopExperimentStats['me
               print '\n\n'
In [711]: pullExperiment('FtLewis5',50)
          <IPython.core.display.HTML at 0x111be8c90>
          Experimental Variable v. Epistat Correlation coefficients:
          Pharmaceutical Effectiveness
                attackRate
                              peakDay peakNumber
                 -0.279254 -0.025465 -0.187938
          ve
                 -0.058289 -0.020835
                                         -0.025321
          Intervention Sequence
                  attackRate
                                peakDay
                                             peakNumber
          vtd 3.406174e-01 -0.011575 4.527300e-01
          atl 2.571136e-18 0.000000 5.402729e-18
          apl -2.281914e-01 -0.060962 -7.351200e-03
          sql -2.953325e-01 0.074852 -8.026916e-02
```

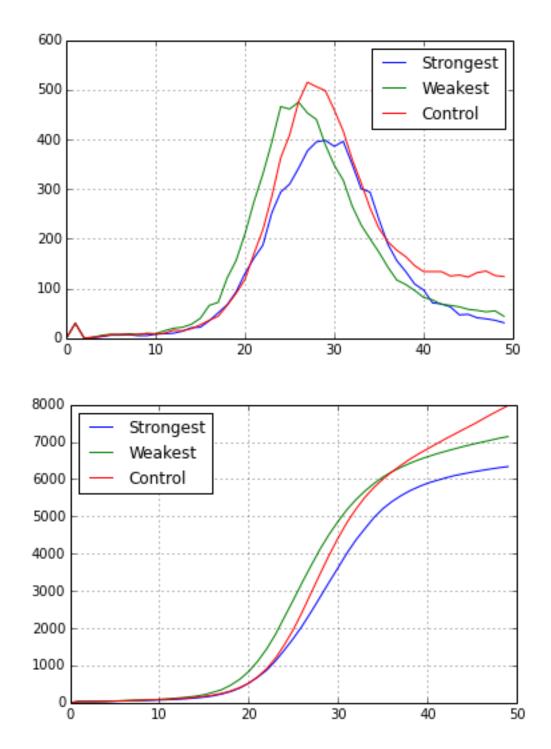
SIWE_ADM_AttackRate = getAR(ADMMeanStats, strongImpWeakEffect)
SIWE_CWOP_AttackRate = getAR(CWOPMeanStats, strongImpWeakEffect)

Attack Rate Extrema by Intervention Parameters

	Weakest	Intervention	Strongest	Intervention
ve		30		70
vtd		17		17
ate		25		25
atl		10		10
apl		10		30
sq		90		90
sqg		25		25
sqtd		10		10
sql		7		28
other		allonpost		active
attackRate		7199		6367
peakDay		26		29
peakNumber		475		398
isEpidemic		-1		-1
leftBound		15		16
rightBound		47		46

Theoretical strongest and weakest pharmacuetical intervention (PI) & logistical implementation (LI):

Active	Duty	Civilian Workers
Strong PI, Strong LI	24.426%	0.038%
Weak PI, Strong LI	24.889%	0.075%
Strong PI, Weak LI	24.310%	0.031%
Weak PI, Weak LI	25.431%	0.044%



In [712]: pullExperiment('FtLewis6',50)

<IPython.core.display.HTML at 0x111b411d0>

Experimental Variable v. Epistat Correlation coefficients:

```
ve -0.025739 -0.025801 -0.047267
ate -0.251619 -0.284967 -0.263670
```

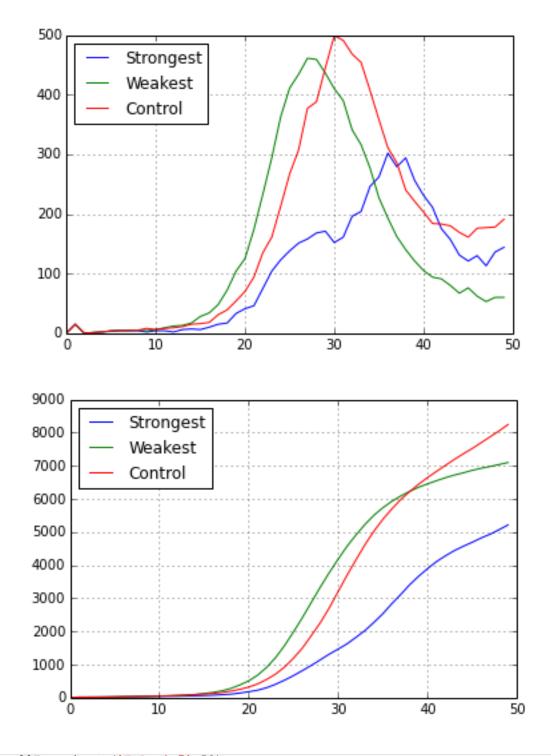
	attackRate	peakDay	peakNumber
vtd	0.058645	-2.310544e-02	7.469193e-02
atl	0.000000	7.661454e-19	9.222918e-18
apl	0.034913	3.311780e-02	5.445559e-02
sql	-0.016052	-4.659597e-02	-1.019704e-02

Attack Rate Extrema by Intervention Parameters

	Weakest	Intervention	Strongest	Intervention
ve		70		30
vtd		38		17
ate		87		87
atl		10		10
apl		10		30
sq		90		90
sqg		25		25
sqtd		10		10
sql		7		28
other		allonpost		allonpost
attackRate		7164		5398
peakDay		27		36
peakNumber		461		302
isEpidemic		-1		-1
leftBound		17		19
rightBound		48		50
sqtd sql other attackRate peakDay peakNumber isEpidemic leftBound		10 7 allonpost 7164 27 461 -1		10 28 allonpost 5398 36 302 -1

Theoretical strongest and weakest pharmacuetical intervention (PI) & logistical implementation (LI):

Active	Duty	Civilian Workers
Strong PI, Strong LI	24.607%	0.044%
Weak PI, Strong LI	24.159%	0.013%
Strong PI, Weak LI	24.444%	0.019%
Weak PI, Weak LI	24.166%	0.038%



In [713]: pullExperiment('FtLewis7',50)

<IPython.core.display.HTML at 0x112285510>

Experimental Variable v. Epistat Correlation coefficients:

```
ve -0.075978 -0.075682 -0.093934
ate 0.200194 0.316148 0.227400
```

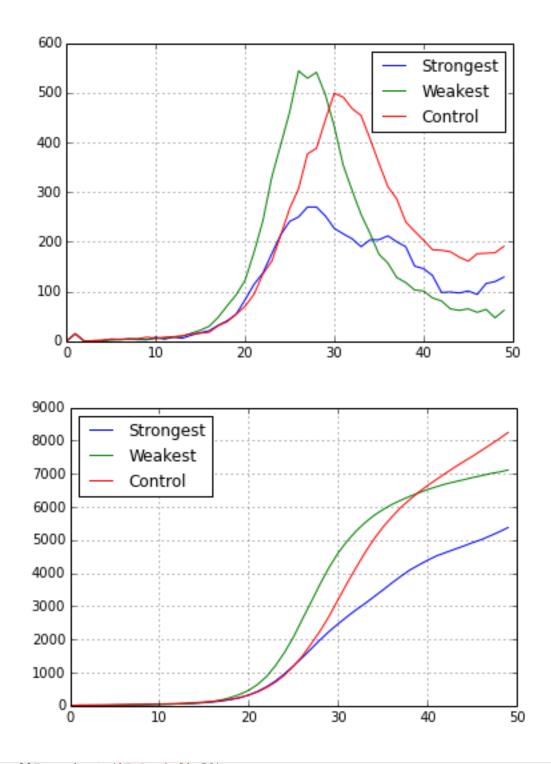
	attackRate	peakDay	peakNumber
vtd	-0.017642	-4.734763e-02	3.631568e-02
atl	0.000000	2.543056e-18	8.127693e-18
apl	-0.010265	-3.802723e-02	-1.430956e-02
sql	-0.036417	-3.094373e-02	-2.554720e-02

Attack Rate Extrema by Intervention Parameters

	Weakest	Intervention	Strongest	Intervention
ve		70		70
vtd		38		38
ate		25		87
atl		10		10
apl		10		10
sq		90		90
sqg		25		25
sqtd		10		10
sql		7		28
other		allonpost		active
attackRate		7172		5537
peakDay		26		27
peakNumber		544		270
isEpidemic		-1		-1
leftBound		18		17
rightBound		47		50

Theoretical strongest and weakest pharmacuetical intervention (PI) & logistical implementation (LI):

Active	Duty	Civilian Workers
Strong PI, Strong LI	24.000%	0.019%
Weak PI, Strong LI	24.571%	0.013%
Strong PI, Weak LI	24.079%	0.031%
Weak PI, Weak LI	24.632%	0.031%



In [714]: pullExperiment('FtLewis8',50)

<IPython.core.display.HTML at 0x111b41f50>

Experimental Variable v. Epistat Correlation coefficients:

```
ve -0.006836 -0.028484 -0.029189
ate -0.274242 -0.266242 -0.269018
```

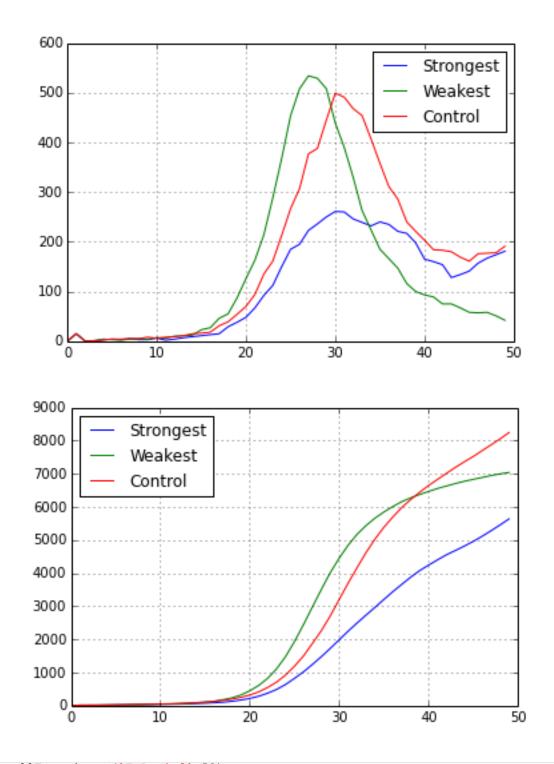
	attackRate	peakDay	peakNumber
vtd	5.587836e-03	-1.032527e-01	4.553079e-02
atl	-6.646898e-18	-2.698500e-18	1.150571e-17
apl	-1.055713e-02	-3.956043e-03	-4.027322e-03
sql	2.674206e-02	2.373626e-03	2.466300e-02

Attack Rate Extrema by Intervention Parameters

	Weakest	Intervention	Strongest	Intervention
ve		70		30
vtd		38		17
ate		87		87
atl		10		10
apl		10		10
sq		90		90
sqg		25		25
sqtd		10		10
sql		7		7
other		allonpost		active
attackRate		7094		5831
peakDay		27		30
peakNumber		534		261
isEpidemic		-1		-1
leftBound		17		19
rightBound		47		50

Theoretical strongest and weakest pharmacuetical intervention (PI) & logistical implementation (LI):

Active	Duty	Civilian Workers
Strong PI, Strong LI	24.621%	0.031%
Weak PI, Strong LI	24.028%	0.063%
Strong PI, Weak LI	23.826%	0.013%
Weak PI, Weak LI	24.141%	0.050%



In [715]: pullExperiment('FtLewis9',50)

<IPython.core.display.HTML at 0x112285550>

Experimental Variable v. Epistat Correlation coefficients:

```
ve -0.073782 0.008363 -0.064721 ate -0.261181 -0.348168 -0.280434
```

	attackRate	peakDay	peakNumber
vtd	8.023694e-02	-0.020600	1.136680e-01
atl	4.861192e-18	0.000000	6.784788e-18
apl	-2.953274e-02	-0.051195	-2.939462e-02
sql	-2.038740e-02	-0.031207	-1.055222e-02

Attack Rate Extrema by Intervention Parameters

	Weakest	Intervention	Strongest	Intervention
ve		30		70
vtd		38		17
ate		87		87
atl		10		10
apl		10		10
sq		90		90
sqg		25		25
sqtd		10		10
sql		7		7
other		allonpost		allonpost
attackRate		7144		5547
peakDay		26		27
peakNumber		489		432
isEpidemic		-1		-1
leftBound		18		17
rightBound		47		49

Theoretical strongest and weakest pharmacuetical intervention (PI) & logistical implementation (LI):

Active	Duty	Civilian Workers
Strong PI, Strong LI	24.560%	0.019%
Weak PI, Strong LI	24.173%	0.057%
Strong PI, Weak LI	23.045%	0.214%
Weak PI, Weak LI	24.600%	0.101%

