



# Let's Talk Analytics

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## Role of Data Analysts in NSHA

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May 30, 2019

Please be advised that we are currently in a controlled vendor environment for the One Person One Record project.

Please refrain from questions or discussion related to the One Person One Record project.

# Informatics...

“Utilizes health information and health care technology to enable patients to receive best treatment and best outcome possible.”

# Clinical Informatics...

“is the application of informatics and information technology to deliver health care.”

AMIA. (2017, January 13). Retrieved from <https://www.amia.org/applications-infomatics/clinical-informatics>

# Analytics...

“is the discovery, interpretation, and communication of meaningful patterns in data.”

“relies on the simultaneous application of analysis, statistics, computer programming and operations research to quantify performance.”

# Objectives

At the conclusion of this activity, participants will be able to...

- Identify what knowledge and skills health care providers will need to use information now and in the future.
- Prepare health care providers by introducing them to concepts and local experiences in Informatics.
- Acquire knowledge to remain current with new trends, terminology , studies, data and breaking news.
- Cooperate with a network of colleagues establishing connections and leaders that will provide assistance and advice for business issues, as well as for best-practice and knowledge sharing.

# Conflict of Interest Declaration

- I do not have an affiliation (financial or otherwise) with a pharmaceutical, medical device, health care informatics organization, or other for-profit funder of this program.

# NSHA Role of Analytics

**The specific objectives of this session are to help you understand the role of Analysts and Analytics in NSHA**



Performance and Analytics



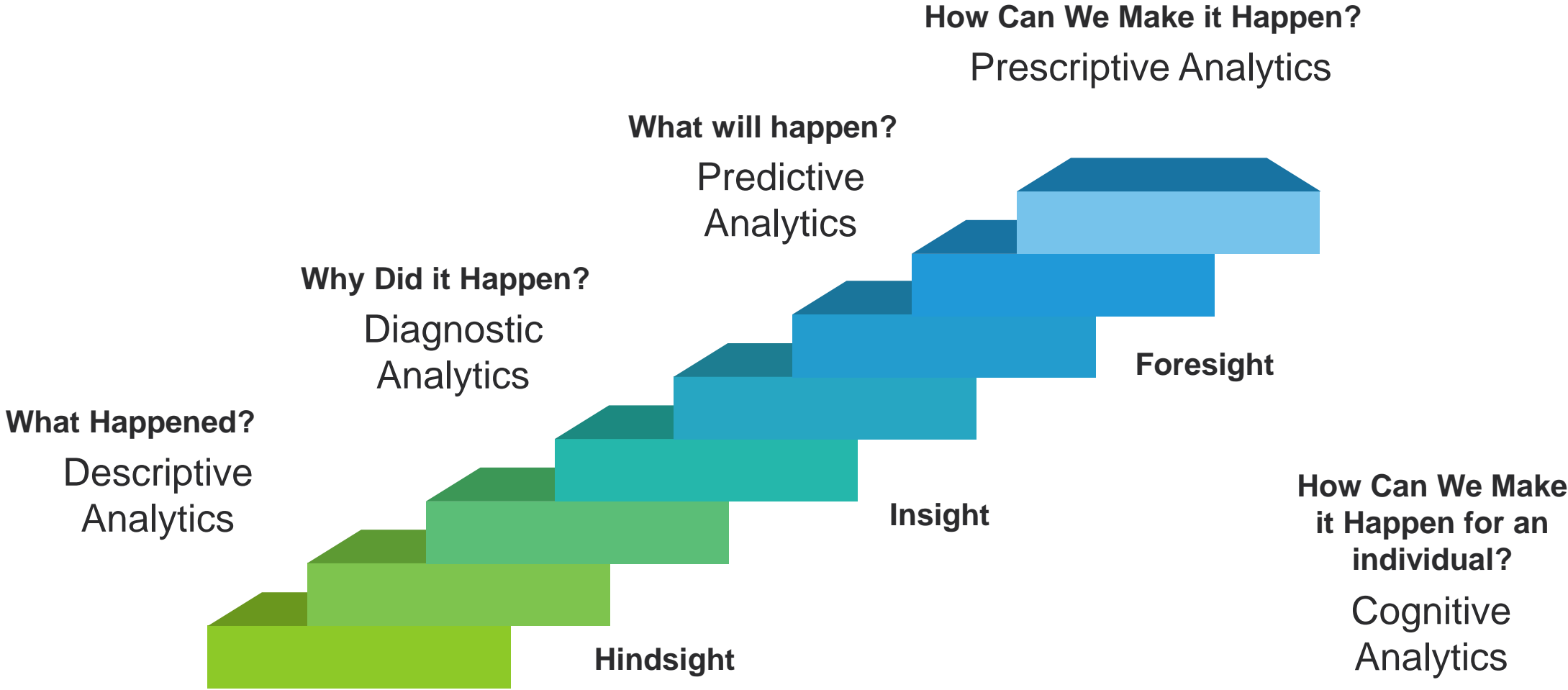
Analytics/Data Science



Applied Analytics - Case Studies




# Steps of the Analytics Continuum



# Performance & Analytics Team

A team of 25 analysts spread across the province, endeavoring to provide data to drive decision making throughout NSHA



“Supporting evidence and data based decision making”

The team is made up of a mix of roles with team members having varied backgrounds and skill sets to support a wide range of analytical services and projects.

- Senior Decision Support Analysts
- Data Analysts
- Decision Support Analysts
- Research and Statistics Officers
- Health Records Analysts
- MIS Statistical Coordinators

# What Do We Do

## Performance & Analytics Team

### Data Extraction and Staging

Data extraction, aggregation and data mining methods organize the data and make it possible to identify patterns and relationships in it that would not otherwise be visible.



### Data Linkage

Allows related information from one data source to be linked to information from another data source. Using the linked data makes it possible to gain a more comprehensive understanding.



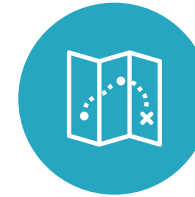
### Descriptive Analytics

Summary of historical data to yield useful information and possibly prepare the data for further analysis. Reporting and data visualization may be applied to yield more insight.



### Inferential Analytics

Examines data and relationships to answer the question “Why did it happen?”, or “What might happen?”.



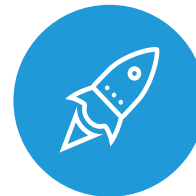
### Performance Measurement

Regular measurement of outcomes and results, which generates reliable data on access, efficiency and outcomes for NSHA programs



### Applied Analytics

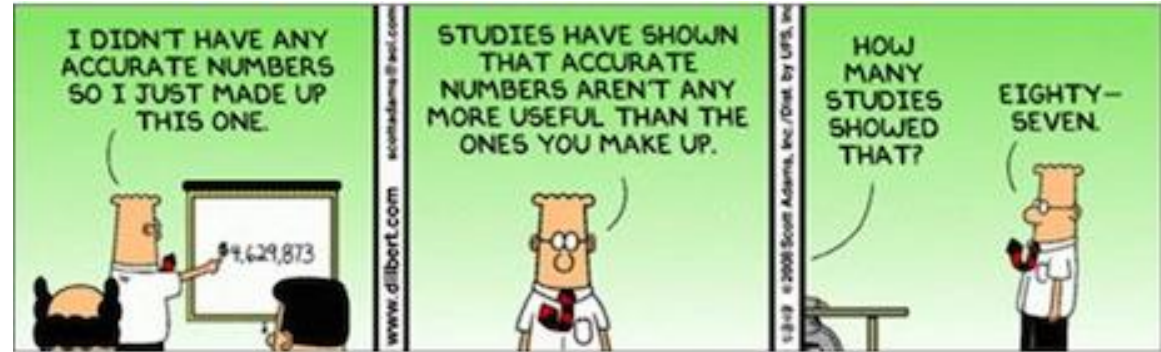
Utilizing advanced statistical methods and innovative tools to create iterative, scalable and operational solutions, applying theory to the practice of health system planning and management.



# What We Don't Do

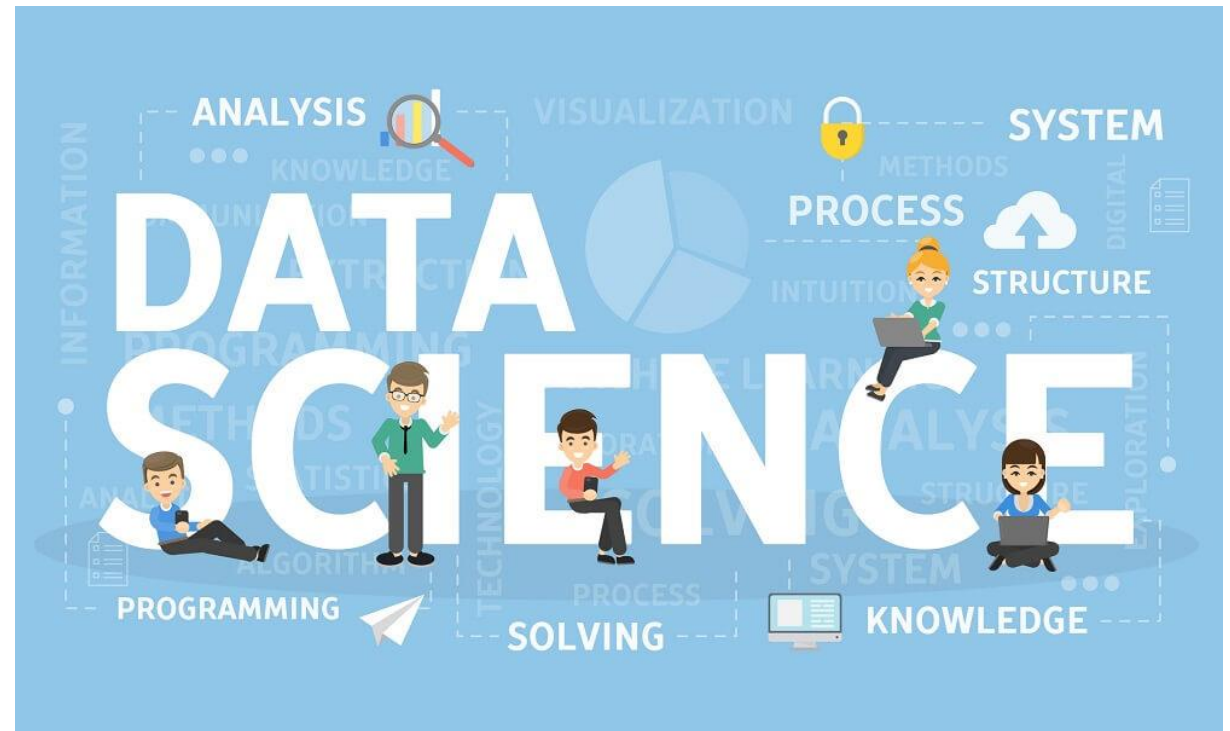
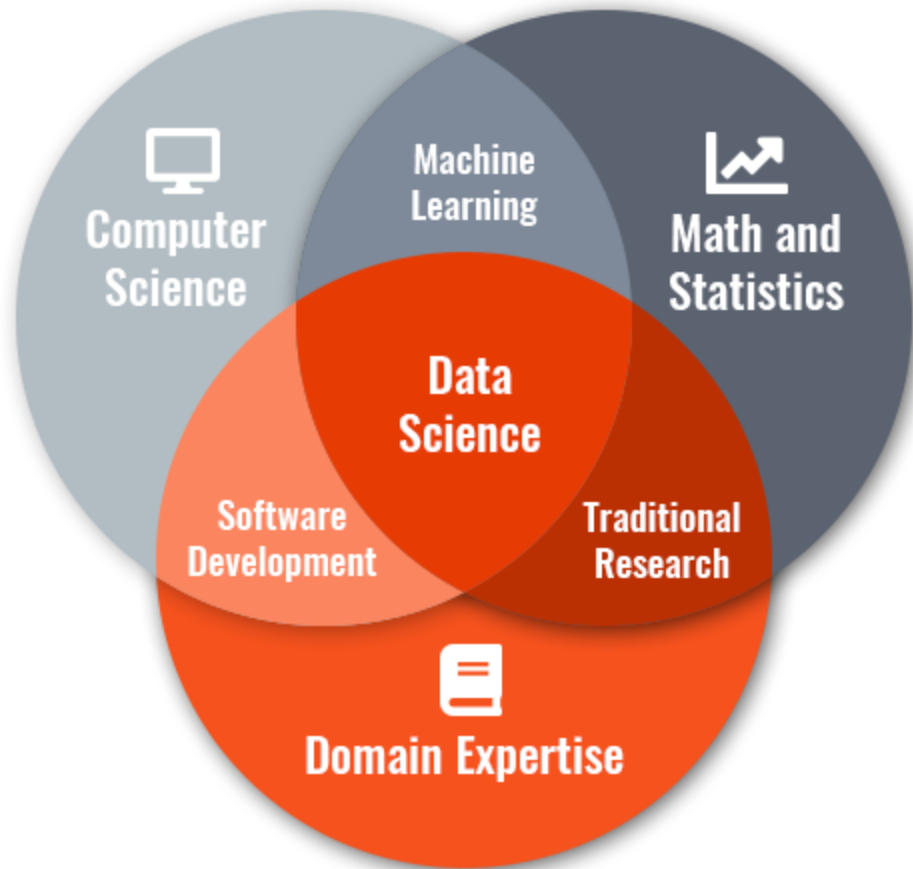


"After analyzing all your data, I think we can safely say that none of it is useful."



# What is Data Science

and, what it isn't



# Analytics

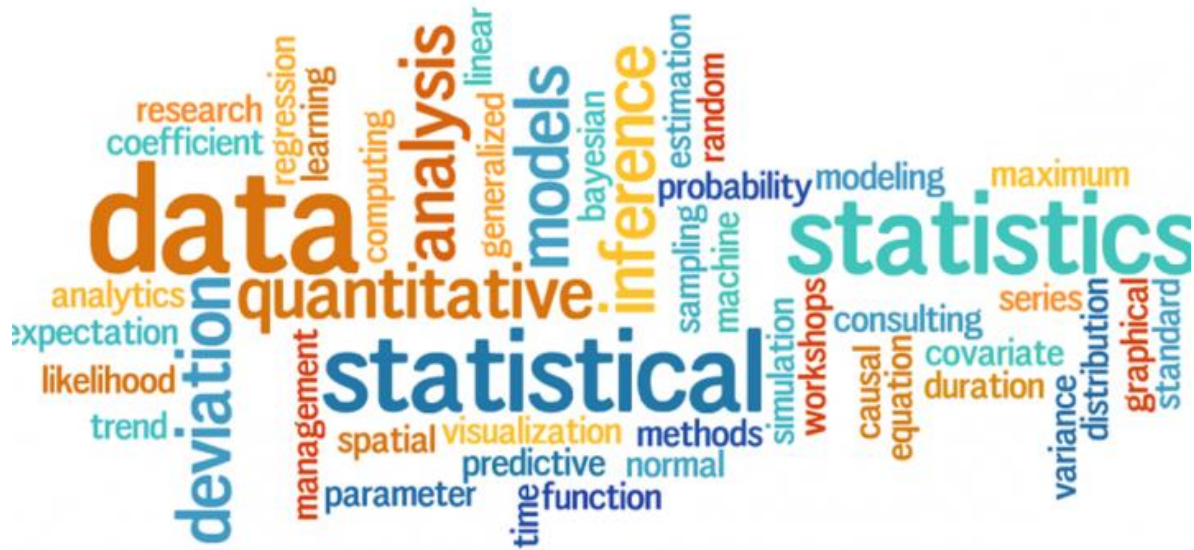
Telling the Story in the Data

- Transforming data into insight, rapid turnaround, often reactive.
- Analysts are data storytellers.
- Summarizing interesting facts and using data for inspiration.
- Domain knowledge and understanding/translation of analytical/statistical concepts to support decision-makers.



# Statistics

Determining which data and conclusions are trustworthy

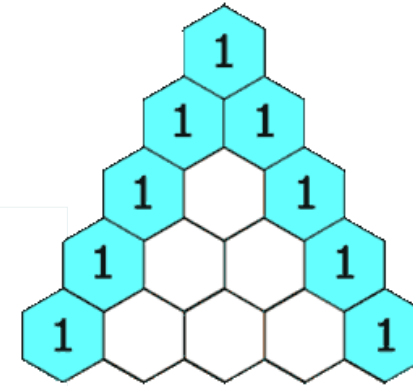
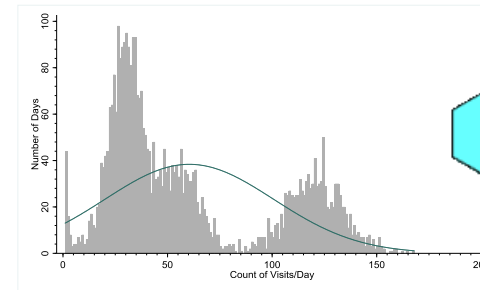


- Statistics allows us to come to conclusions beyond our data (inferences).
- Ensuring the methods applied are right for the problem and inferences are valid from the information at hand.
- People add the new information to their own pre-conceptions of the data (bias).

# Statistics

## Identifying and Understanding Patterns

- Understanding the patterns (distributions) and relationships in our data.
- Using these patterns to guide decisions and understand the past and predict the future
- Sorting out the signal from the noise (meaningful patterns)

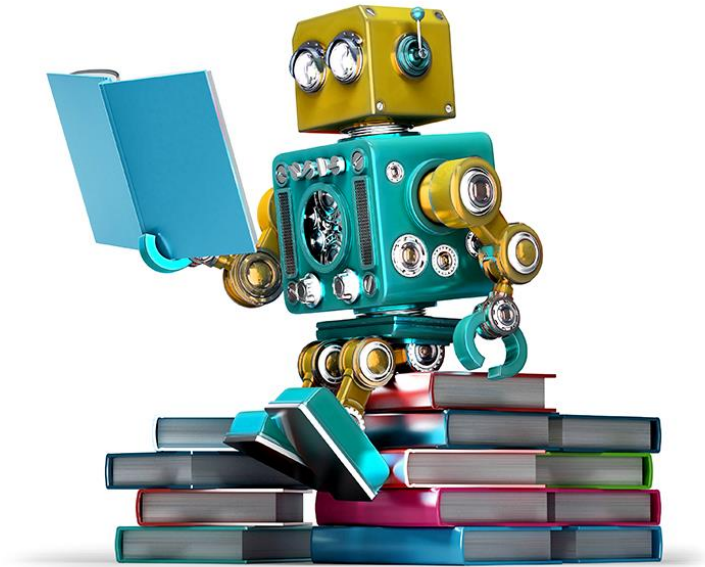




# Machine Learning

## Automation and Performance

- Machine learning is a method of analytics that automates analytical model building. It is a branch of artificial intelligence based on the idea that systems can learn from data, identify patterns and make decisions with minimal human intervention.
- Performance means reliable, scalable, and easy-to-maintain models that perform well in practice.



# Wide versus Deep

Avoiding the “rabbit hole”

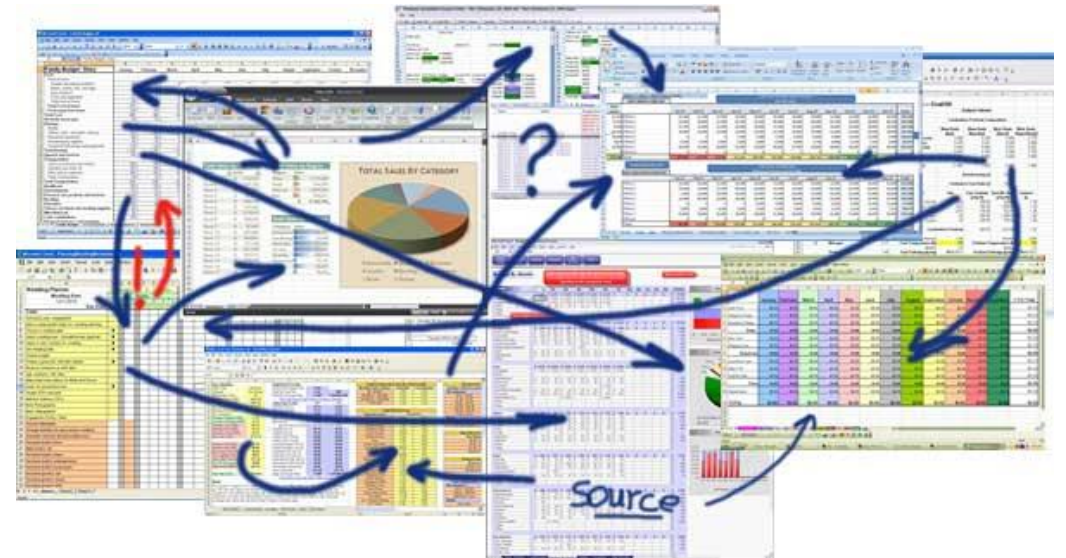


- The previous two domains of analytics both provide high-effort solutions to specific problems. If the problems they tackle aren't worth solving, you end up wasting time and resources.
- Statistics and machine learning are narrow-and-deep pursuits — the shape of a rabbit hole, incidentally — it's important to tackle problems that deserve the effort.

# Applying Data and Theory to Problems

Transform Patient Data into Intelligent Action

- As an organization we're constantly producing data; numbers, reports, trend lines, graphs, spreadsheets and more.
- But, do we know if all this data leads to better decisions?



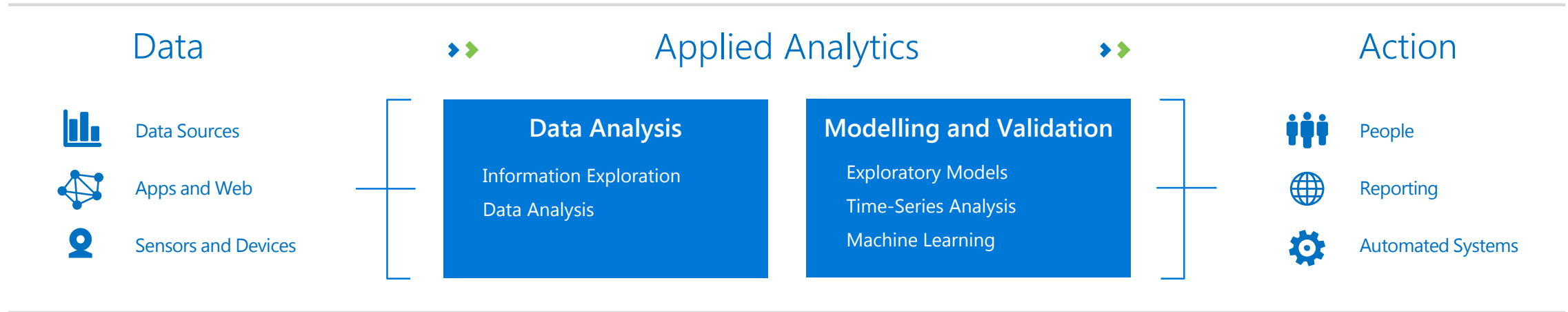
# Applied Analytics

Applying theory to the practice of health system management



- Multi-disciplinary approach focused on the pain points of health system management.
- Utilizing advanced analytical and statistical methods and innovative tools
- Develop scalable and operational solutions for identified problems

# Applied Analytics Methodology | Transform Patient Data into Intelligent Action



## Data Requirement

e.g. emergency department visits, surgical cases, outpatient visits, midnight stays history

## Modeling

Use multiple models to produce forecasts/projections

## Improved Operations

<b>Potential Short-Term Forecasting</b>	<b>Potential Long Term Forecasting</b>
Utilization	Demographics
Weekly cycles	Industry trends
Daily actuals	Policy trends

# Analytical Case Studies

**The specific objectives of this session are to help you understand the role of Performance and Analytics in NSHA and specific analyst roles**



Utilization Forecasts



Utilization - Structural Equation Modelling

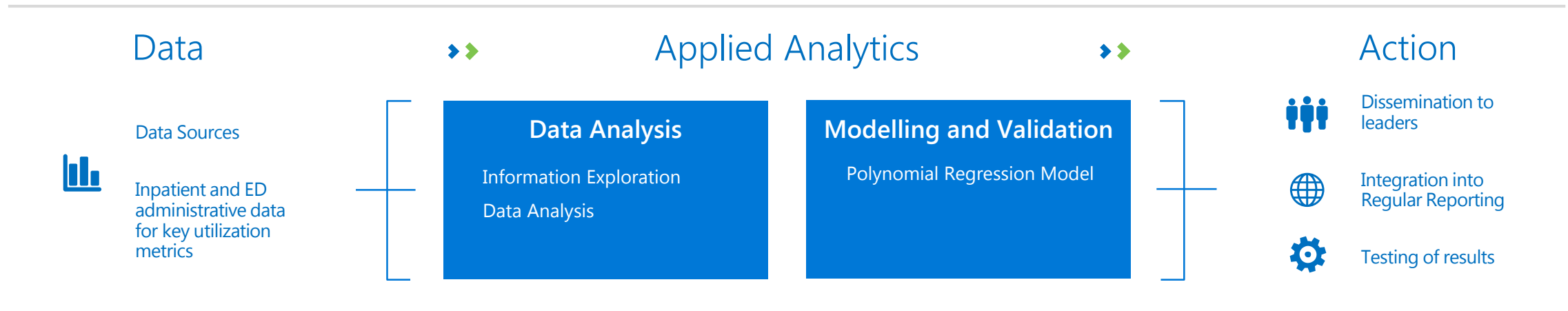
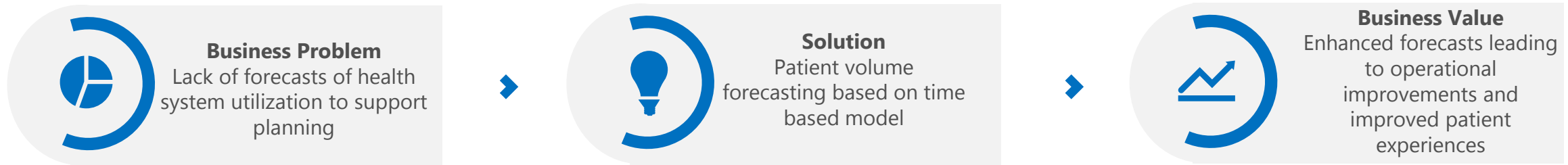


Population Based Projections



Predicting Patient Flow

# Applied Analytics | Utilization Volume Modelling and Forecasting



## Data Requirement

**Patient Volume History**  
e.g. emergency department visits, surgical cases, outpatient visits, etc.

## Modeling

**Examination of multiple models to produce forecasts**

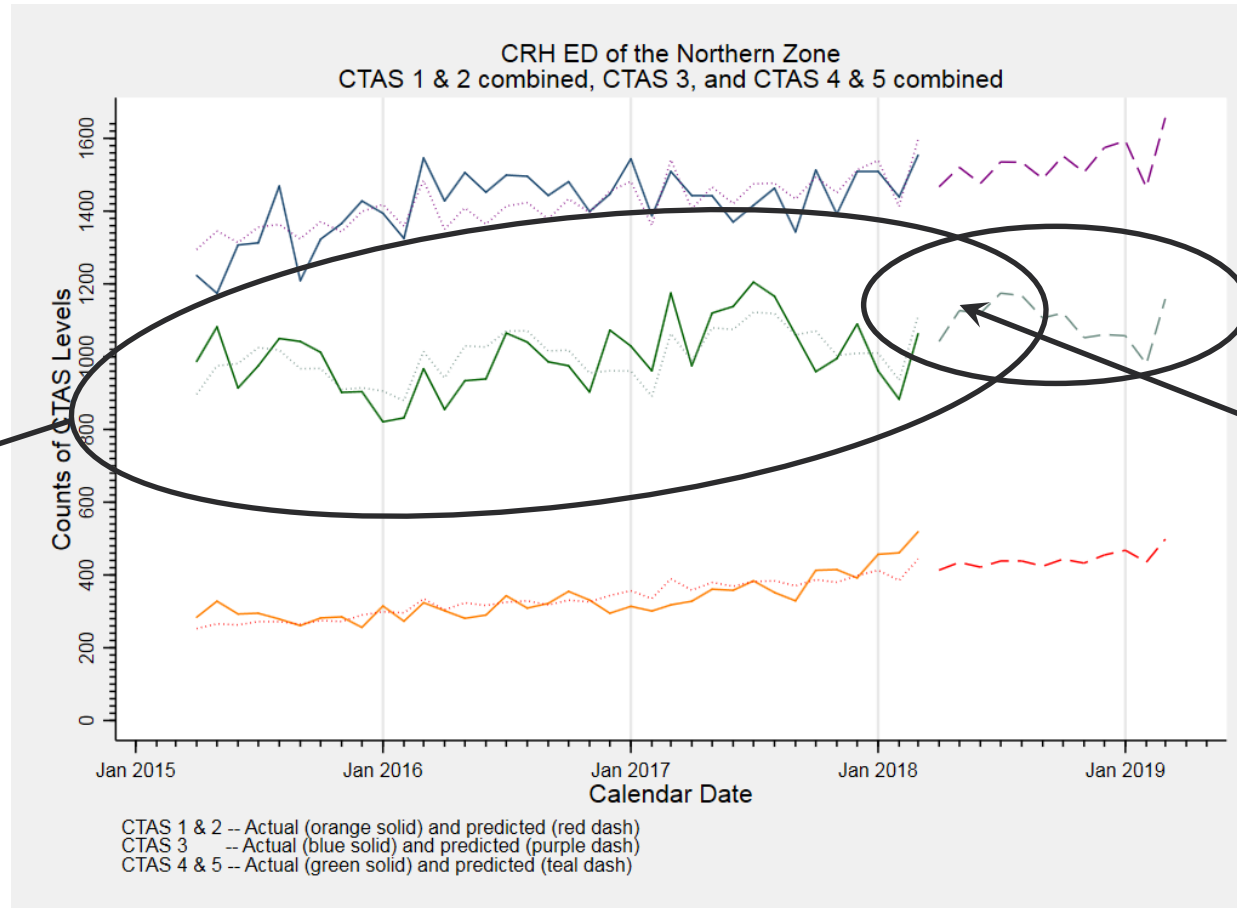
- Time Series
- Polynomial Regression
- Linear Regression
- Autoregressive–moving-average (ARMA) models

## Improved Forecasting

<b>Potential Short-Term Forecasting</b>	<b>Potential Long Term Forecasting</b>
Scheduling	Demographics
Weekly cycles	Utilization trends
Daily actuals	

# Utilization Modelling and Forecasting

Understanding the Past to Predict the Future



Modelling  
patterns  
cause and  
effects  
associations

Forecasting

testing

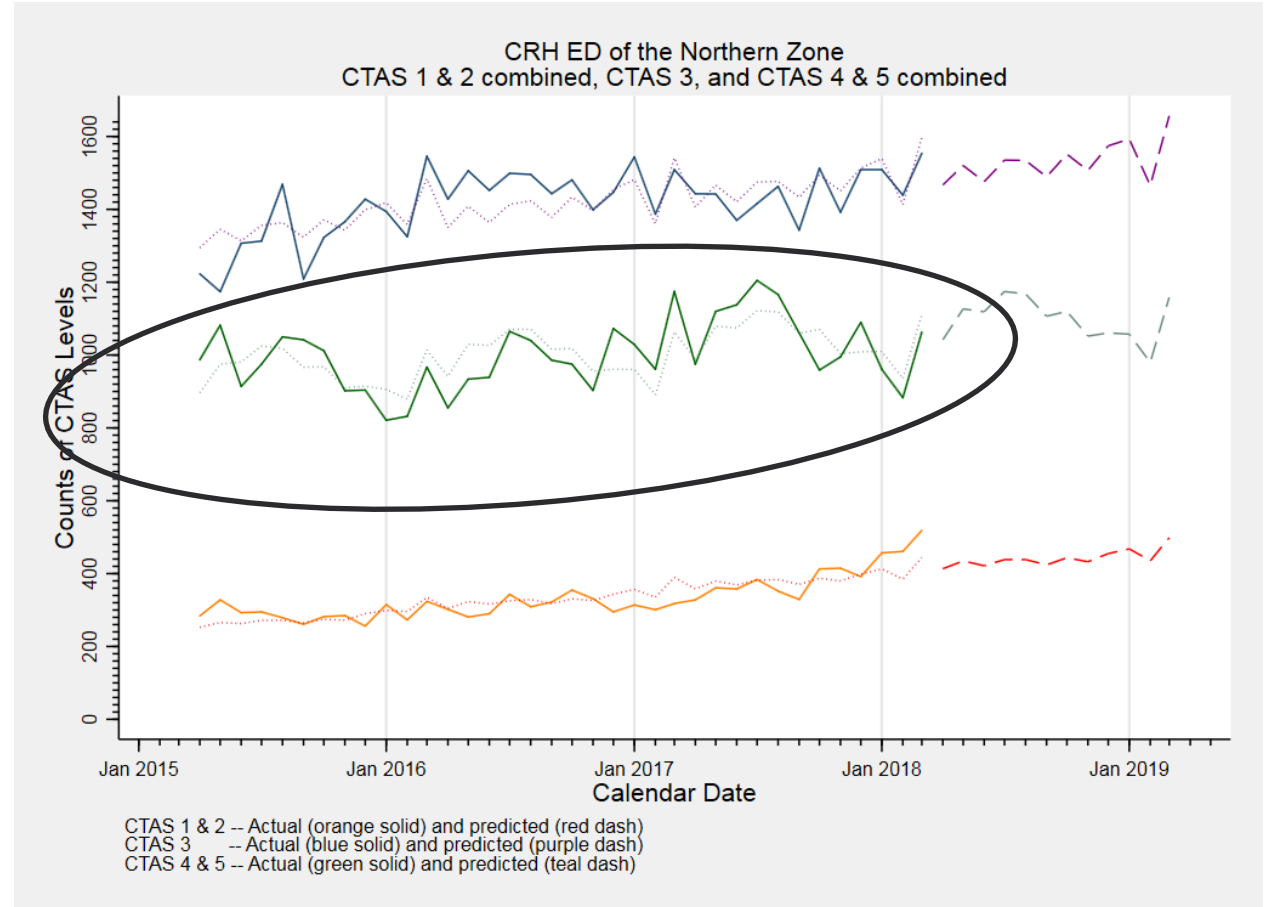
- Forecasting is key, but the quality (utility) of the forecast depends on the quality of the model
- The quality of the model depends on the choice of variables, and the quality of measurement



# Modelling - Understanding the Past

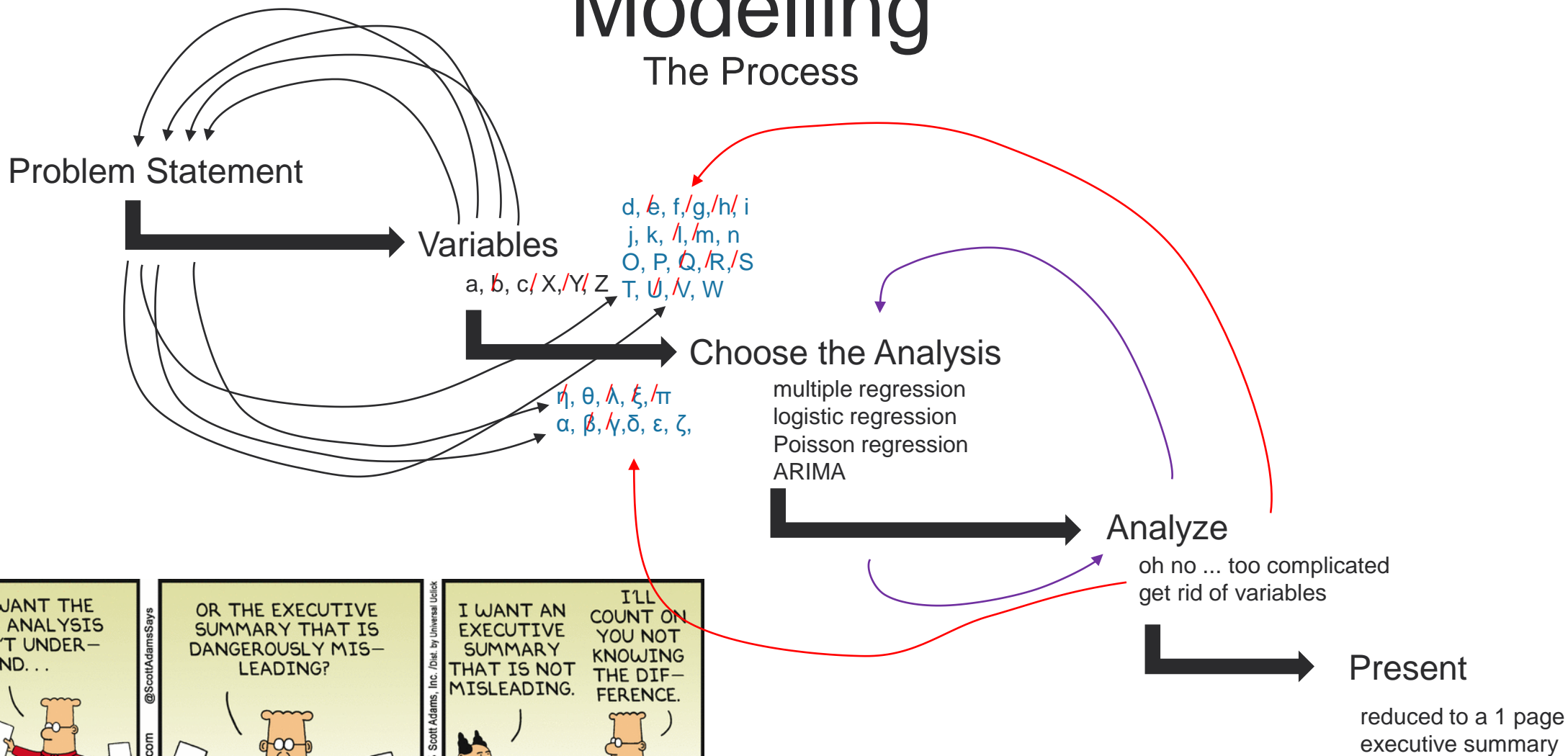
The quality of the model depends on the choice of variables, and the quality of measurement

- What is modeled?
  - Dependent Variable (visits, occupancy, admissions, etc.)
- How is it modeled?
  - Independent Variable or Predictors
  - prior pattern of the DV -- the future will look like the past
  - this is “time”
  - It works reasonably well



# Modelling

## The Process



# Modelling

## The Actual Analysis

The analysis is all done in Stata – it is quite straightforward currently

This little bit of code does all the work of modelling (and forecasting).

```

regress ED_Board                                     ///
    DayNum_1 DayNum_2 DayNum_3 DayNum_4 DayNum_5   ///
    FM_1 FM_2 FM_3 FM_4                             ///
    WkDay_1 WkDay_2 WkDay_3 WkDay_4 WkDay_5 WkDay_6 ///
    , cformat(%9.3f) pformat(%5.3f) sformat(%6.3f)
local Fit : display %6.3f e(r2)
local FValue : display %6.3f e(F)
scalar PValue = 1- F(e(df_m),e(df_r),e(F))
local pValue : display %6.3f PValue
local Sample : display e(N)
display "Boarding: "
display "R2 = `Fit' (F=`FValue', p<`pValue'), based on `Sample' points. "
predict Expected
predict StdErr, stdp
predict ForErr, stdf
gen Error = temp - Expected
gen LCI95 = Expected + invnormal(.025)*(ForErr )
gen UCI95 = Expected - invnormal(.025)*(ForErr )

```



Source	SS	df	MS	Number of obs	=	1,249
Model	3214.09803	15	214.273202	F(15, 1233)	=	65.54
Residual	4031.01326	1,233	3.26927272	Prob > F	=	0.0000
				R-squared	=	0.4436
				Adj R-squared	=	0.4369
Total	7245.11129	1,248	5.80537764	Root MSE	=	1.8081

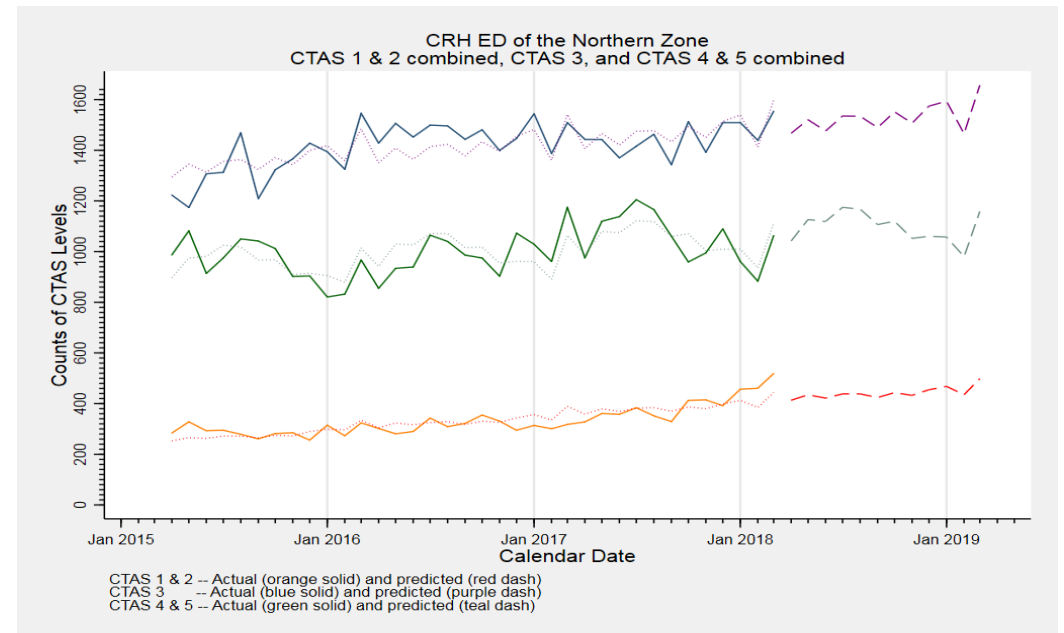
ED_Boarded	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
DayNum_1	0.003	0.001	6.111	0.000	0.002 0.005
DayNum_2	0.000	0.000	8.706	0.000	0.000 0.000
DayNum_3	0.000	0.000	4.451	0.000	0.000 0.000
DayNum_4	-0.000	0.000	-6.328	0.000	-0.000 -0.000
DayNum_5	-0.000	0.000	-5.492	0.000	-0.000 -0.000
FM_1	0.060	0.052	1.162	0.246	-0.041 0.162
FM_2	0.003	0.016	0.161	0.872	-0.028 0.033
FM_3	-0.002	0.003	-0.695	0.487	-0.008 0.004
FM_4	0.000	0.001	0.083	0.934	-0.001 0.001
WkDay_1	-0.139	0.057	-2.457	0.014	-0.250 -0.028
WkDay_2	-0.198	0.096	-2.065	0.039	-0.386 -0.010
WkDay_3	-0.060	0.043	-1.411	0.159	-0.144 0.024
WkDay_4	-0.034	0.036	-0.938	0.348	-0.105 0.037
WkDay_5	0.007	0.043	0.155	0.877	-0.077 0.091
WkDay_6	-0.032	0.016	-1.997	0.046	-0.064 -0.001
_cons	1.010	0.130	7.752	0.000	0.755 1.266

But ....

## This much code is required to make a single graph

```
summarize CTAS01_02
local YScale1 = r(max) * 1.25
summarize CTAS03
local YScale2 = r(max) * 1.25
summarize CTAS04_05
local YScale3 = r(max) * 1.25
display `YScale1' " " `YScale2' " " `YScale3'
local YScale = max(`YScale1', `YScale2', `YScale3')
* Create the Title(s)
local Title1 "`NameZone' Zone"
local Title2 "CTAS 1 & 2 combined, CTAS 3, and CTAS 4 & 5 combined"
* Create the legend
local Line1Text "CTAS 1 & 2 -- Actual (orange solid) and predicted (red dash)"
local Line2Text "CTAS 3 -- Actual (blue solid) and predicted (purple dash)"
local Line3Text "CTAS 4 & 5 -- Actual (green solid) and predicted (teal dash)"
twoway (line CTAS01_02 Date if (Date >= FirstPredict & Date <= LastCheck)
, lcolor(orange) lwidth(thin) lpattern(solid)
connect(direct) cmiss(n)
(line CTAS01_02_Exp Date if (Date >= FirstPredict &
Date <= LastPredict)
, lcolor(red) lwidth(thin) lpattern(tight_dot)
connect(direct) cmiss(n)
(line CTAS01_02_Exp Date if (Date >= LastPredict &
Date <= LastForecast)
, lcolor(red) lwidth(thin) lpattern(dash)
connect(direct) cmiss(n)
(line CTAS03 Date if (Date >= FirstPredict &
Date <= LastCheck)
, lcolor(navy) lwidth(thin) lpattern(solid)
connect(direct) cmiss(n)
(line CTAS03_Exp Date if (Date >= FirstPredict &
Date <= LastPredict)
, lcolor(purple) lwidth(thin) lpattern(tight_dot )
connect(direct) cmiss(n)
(line CTAS03_Exp Date if (Date >= LastPredict &
Date <= LastForecast)
, lcolor(purple) lwidth(thin) lpattern(dash)
connect(direct) cmiss(n)
(line CTAS04_05 Date if (Date >= FirstPredict &
Date <= LastCheck)
, lcolor(dkgreen) lwidth(thin) lpattern(solid)
connect(direct) cmiss(n)
(line CTAS04_05_Exp Date if (Date >= FirstPredict &
Date <= LastPredict)
, lcolor(teal) lwidth(thin) lpattern(tight_dot)
connect(direct) cmiss(n))
```

```
(line CTAS04_05_Exp Date if (Date >= LastPredict &
Date <= LastForecast)
, lcolor(teal) lwidth(thin) lpattern(dash)
connect(direct) cmiss(n)
, ytitle (Counts of CTAS Levels) ytitle(, size(small) )
yscale(range(0 `YScale') )
ylabel(#8, labsize(vsmall) format(%5.0f) ticks tcolor(black)
tposition(outside) nogrid)
ymtick(##10, ticks tcolor(black) tposition(outside) nogrid)
xtitle(Calendar Date) xtitle(, size(small))
xlabel(#5, labsize(vsmall) labcolor(black) ticks tcolor(black)
tposition(outside) grid)
xmtick(#60, ticks tcolor(black) tposition(outside) nogrid)
title("`Title1'" "`Title2'")
, size(small))
note( "`Line1Text'" "`Line2Text'" "`Line3Text'"
, size(vsmall))
legend(off)
```



and ....

# This chunk code is required for data cleaning

```

gen str FacilityCode = substr(Facility, 1, 3), after (Facility)
replace FacilityCode = FacilityCode + "*" + SiteCB if (FacilityCode == "CBM")
label variable FacilityCode "Facility Code (Mnemonic)"
tabulate FacilityCode

* The encode command is convenient
* It will create numericals in file order.
* encode FacilityCode, gen (FacilityNum)
* IT ADDS VALUE LABELS AUTOMATICALLY
* tabulate FacilityNum, nolabel
* order FacilityNum, after (FacilityCode)
* However, the following will provide alphabetical or custom orders
* Herein the regional facilities are listed first
* In alphabetical order (by full name), with 4 sites for CBR
* Numerical codes are required because some procedures (eg collapse)
* are restricted to numeric values.

gen FacilityNum = 0, after (FacilityCode)
label variable FacilityNum "Facility Coded by Number (see labels)"
replace FacilityNum = 1 if ( FacilityCode == "ARM")
replace FacilityNum = 2 if ( FacilityCode == "CBH-CBC")
replace FacilityNum = 3 if ( FacilityCode == "CBH-CBN")
replace FacilityNum = 4 if ( FacilityCode == "CBH-CBR")
replace FacilityNum = 5 if ( FacilityCode == "CBH-CBW")
replace FacilityNum = 6 if ( FacilityCode == "CRB")
replace FacilityNum = 7 if ( FacilityCode == "CCU")
replace FacilityNum = 8 if ( FacilityCode == "SMR")
replace FacilityNum = 9 if ( FacilityCode == "SSR")
replace FacilityNum = 10 if ( FacilityCode == "VRM")
replace FacilityNum = 11 if ( FacilityCode == "YRM")
replace FacilityNum = 12 if ( FacilityCode == "ACM")
replace FacilityNum = 13 if ( FacilityCode == "ASH")
replace FacilityNum = 14 if ( FacilityCode == "BMH")
replace FacilityNum = 15 if ( FacilityCode == "DGH")
replace FacilityNum = 16 if ( FacilityCode == "EMH")
replace FacilityNum = 17 if ( FacilityCode == "FMH")
replace FacilityNum = 18 if ( FacilityCode == "GMH")
replace FacilityNum = 19 if ( FacilityCode == "ICM")
replace FacilityNum = 20 if ( FacilityCode == "LFM")
replace FacilityNum = 21 if ( FacilityCode == "NCR")
replace FacilityNum = 22 if ( FacilityCode == "OGH")
replace FacilityNum = 23 if ( FacilityCode == "RWH")
replace FacilityNum = 24 if ( FacilityCode == "SAC")
replace FacilityNum = 25 if ( FacilityCode == "SCC")
replace FacilityNum = 26 if ( FacilityCode == "SHH")
replace FacilityNum = 27 if ( FacilityCode == "SMH")
replace FacilityNum = 28 if ( FacilityCode == "SSM")
replace FacilityNum = 29 if ( FacilityCode == "SRH")
replace FacilityNum = 30 if ( FacilityCode == "VCM")

tabulate FacilityNum
label define FacilityNum 1 "ARM: Pictou County Health Authority", modify
label define FacilityNum 2 "CBH-CBG: Cape Breton Healthcare Complex, Glace Bay", modify
label define FacilityNum 3 "CBH-CBN: Cape Breton Healthcare Complex, Northside", modify
label define FacilityNum 4 "CBH-CBR: Cape Breton Healthcare Complex, CBR", modify
label define FacilityNum 5 "CBH-CBW: Cape Breton Healthcare Complex, New Waterford", modify
label define FacilityNum 6 "CCM: Colch East Hants Health Centers", modify
label define FacilityNum 7 "CCM: Cumberland Reg Healthcare Ctr", modify
*/
tabulate Triage if ( Triage_Date == . & DepartureStatus == .)
Total | 1 100.00

count if ( Triage < . & Triage_Date < . & DepartureStatus == .)
" NCR NT - PT LT BEFORE BEING TRIAGED(OP)"
* 19

replace CommentsTime = "Triage: Flag Only - left without being triaged, " +
"but has triage time and value"
if ( Triage < . & Triage_Date < . & DepartureStatus == .)
" NCR NT - PT LT BEFORE BEING TRIAGED(OP)"

count if ( Triage == . & Triage_Date < . & DepartureStatus == .)
" NCR NT - PT LT BEFORE BEING TRIAGED(OP)"

replace CommentsTime = "Triage: Flag Only - left without being triaged, " +
"but has triage time but no value"
if ( Triage == . & Triage_Date < . & DepartureStatus == .)
" NCR NT - PT LT BEFORE BEING TRIAGED(OP)"

count if ( Triage < . & Triage_Date == . & DepartureStatus == .)
" NCR NT - PT LT BEFORE BEING TRIAGED(OP)"

replace CommentsTime = "Triage: Flag Only - left without being triaged, " +
"but has triage value but no time"
if ( Triage < . & Triage_Date == . & DepartureStatus == .)
" NCR NT - PT LT BEFORE BEING TRIAGED(OP)"

save "MT_FY2015-18_Cleaned", replace

save "MT_FY2015-18_Cleaned", replace

label define FacilityNum 8 "SMR: St. Martha's Regional Hospital", modify
label define FacilityNum 9 "SSR: South Shore Regional", modify
label define FacilityNum 10 "VCM: Valley Regional Hospital", modify
label define FacilityNum 11 "YRM: Yarmouth Regional Hospital", modify
label define FacilityNum 12 "ACM: Annapolis Community Health Ctr", modify
label define FacilityNum 13 "ASH: All Saints' Springhill Hosp", modify
label define FacilityNum 14 "BMH: Buchanan Memorial Hospital", modify
label define FacilityNum 15 "DGH: Digby General Hospital", modify
label define FacilityNum 16 "EMH: Eastern Memorial Hospital", modify
label define FacilityNum 17 "FMH: Fishermen's Memorial Hospital", modify
label define FacilityNum 18 "GMH: Guysborough Memorial Hospital", modify
label define FacilityNum 19 "OGH: Inverness Consolidated Mem", modify
label define FacilityNum 20 "LFM: Lillian Fraser Memorial Hosp", modify
label define FacilityNum 21 "NCR: North Cumberland Memorial Hosp", modify
label define FacilityNum 22 "OGH: Queens General Hospital", modify
label define FacilityNum 23 "RWH: Roseway Hospital", modify
label define FacilityNum 24 "SAC: St Anne's Community Centre", modify
label define FacilityNum 25 "SCC: South Cumberland Comm Care Ctr", modify
label define FacilityNum 26 "SHH: Sacred Heart Hospital", modify
label define FacilityNum 27 "SMH: Soldiers' Memorial Hospital", modify
label define FacilityNum 28 "SSM: St. Mary's Memorial Hospital", modify
label define FacilityNum 29 "SRH: Strait-Richmond Hospital", modify
label define FacilityNum 30 "VCM: Victoria County Memorial Hosp", modify

label values FacilityNum FacilityCode
tabulate FacilityNum
label define FacilityNum, nolabel

gen SiteCBNum = ., after (SiteCB)
replace SiteCBNum = 1 if (SiteCB == "CBR")
replace SiteCBNum = 2 if (SiteCB == "CBG")
replace SiteCBNum = 3 if (SiteCB == "CBN")
replace SiteCBNum = 4 if (SiteCB == "CBW")

label variable SiteCBNum "CB Site: CBR (1), CBG-CBN (2), NNCB-CBW (4)"
label define SiteCB 1 "CBR: Cape Breton Regional ED"
label define SiteCB 2 "CBG: Glace Bay ED", add
label define SiteCB 3 "CBN: Northside ED", add
label define SiteCB 4 "CBW: New Waterford ED", add
label values SiteCBNum SiteCB
tabulate SiteCB SiteCBNum

count if (Zone == .)
* 3,374
display _N - r(N)
* 1170206

tabulate Zone
* missing 3374
tabulate FacilityNum if ( Zone == . ), nolabel
tabulate Zone if ( FacilityNum == 24 )
* Missing on all
replace Zone = 3 if ( FacilityNum == 24 )
* Done

count if (HCN == .)
display _N - r(N)
* missing on 196

count if (Sex == "")
tabulate Sex
* No missing values -- some
gen SexNum = ., after (Sex)
replace SexNum = 1 if (Sex == "F" | Sex == "f")
replace SexNum = 2 if (Sex == "M" | Sex == "m")
label variable SexNum "Sex: 1 = Male, 0 = Female"
tabulate SexNum Sex

count if (Age == .)
* No missing values

count if ( Residence == "" )
* Missing on 4,995
gen int ResidenceCode = real(substr(Residence, 1, 2)), after (Residence)
label variable ResidenceCode "Place of Residence coded by number (see labels)"
gen int ResidenceSimple = ., after (ResidenceCode)
replace ResidenceSimple = 1 if ( ResidenceCode < 80 )
replace ResidenceSimple = 2 if ( ResidenceCode > 79 & ResidenceCode < 97 )
replace ResidenceSimple = 3 if ( ResidenceCode == 97 )
replace ResidenceSimple = 4 if ( ResidenceCode == 98 )
replace ResidenceSimple = 5 if ( ResidenceCode == 99 )
replace ResidenceSimple = 5 if ( ResidenceCode == . )
label value ResidenceSimple "Simplified Place of Residence: NS, CAN, US, Other, Unknown/Missing"
label define Residence 1 "Nova Scotia"
label define Residence 2 "Canada other than Nova Scotia", modify
label define Residence 3 "USA", modify
label define Residence 4 "Other Countries", modify
label define Residence 5 "Unknown or Missing", modify
tabulate Residence ResidenceSimple
count if ( ResidenceSimple == . )
* missing on 4,995

* Not done here down... more like a "template" depending on what is needed.

count if ( FamilyDoctor == "" )
* Missing on 29776
count if ( FamilyDoctorCode == "" )
* Missing on only 23788
* Fix later when needed.

display _N
tabulate Insurance
* missing 1
tabulate InsuranceCode
* missing 2
tabulate Insurance if ( InsuranceCode == "" )
tabulate InsuranceCode
replace InsuranceCode = "DOH" if ( InsuranceCode == "" & Insurance != "" )

list ResidenceSimple if ( InsuranceCode == "" | Insurance == "" )
replace InsuranceCode = "DOH" if ( InsuranceCode == "" & Insurance == "" )

count if ( Scheduled == "" )
tabulate Scheduled
gen ScheduledNum = 0, after (Scheduled)
label variable ScheduledNum "Scheduled: Yes (1) or No (0), No assumed for missing"
replace ScheduledNum = 1 if ( Scheduled == "Y" | Scheduled == "y" )
tabulate ScheduledNum
tabulate ScheduledNum Scheduled

tabulate PresentingComplaint
* All have values, but 308,870 are "NULL"
tabulate FacilityNum if ( PresentingComplaint == "NULL" )
* Spread across all facilities.

tabulate AdmissionSource
* All have values, but 998,956 are "NULL"
* Another 46,811 are "UNKNOWN"

save "MT_FY2015-18_V2", replace
* use "MT_FY2015-18_V2", clear

* Basic fixes for times
* Times are the core to visits ... correct the obvious errors, and flag the rest

count if ( Arrival_Date == . )
* Missing on 127,186
count if ( Triage_Date == . )
* Missing on 2,565
count if ( Regsit_Date == . )
* Missing on 0
count if ( AssRoom_Date == . )
* Missing on 709,492
count if ( PhysAssess_Date == . )
* Missing on 409,053
count if ( OBS_In_Date == . )
* "Missing" on 1,156,862
count if ( OBS_Out_Date == . )
* "Missing" on 1,156,894
count if ( OBS_DisChg_Date == . )
* "Missing" on 1,161,004
count if ( CDU_In_Date == . )
* "Missing" on 1,156,862
count if ( CDU_Out_Date == . )
* "Missing" on 1,168,186
count if ( ADT_Decision_Date == . )
* "Missing" on 1,136,783
count if ( ADT_In_Date == . )
* "Missing" on 1,109,334
count if ( ADT_Out_Date == . )
* "Missing" on 1,035,125
count if ( ADT_OutBed_Date == . )
* "Missing" on 1,096,556
count if ( ADT_DisChg_Date == . )
* "Missing" on 1,168,079
count if ( Depart_Date == . )
* Missing on 5,765
count if ( LeftER_Date == . )
* Missing on 4,921

gen str CommentsTime = "", after (LeftER_Date)
* Comments have the form:
* TimeUnit: Flag Only - error
* TimeUnit: Fixed as ... - error

* Registration is the reference point.
* Registration time should equal the triage and arrival dates -- more or less.
count if ( Triage_Date == . )
* 2565
tabulate DepartureStatus if ( Triage_Date < . )
tabulate Triage if ( Triage_Date < . & DepartureStatus == .)

tabulate Triage if ( Triage_Date == . )
clearly triaged: Set date to Regsit_Date
count if ( Triage_Date == . & Triage < . )
replace CommentsTime = "Triage value but no date: " +
"Replaced Triage date with Registration date"
if ( Triage_Date == . & Triage < . )
replace Triage_Date = Regsit_Date if ( Triage_Date == . & Triage < . )
save "MT_FY2015-18_Cleaned", replace

* Can trust admission dates -- they are "important"
count if ( Triage_Date == . & Triage == . & ADT_Decision_Date < . )
* 19

* These have been admitted -- hence triaged (implicitly, at least)
replace CommentsTime = "Triage: Fixed Admitted but no triage date or value: " +
"Replaced Triage date with Registration date"
if ( Triage_Date == . & Triage < . & ADT_Decision_Date < . )
replace Triage_Date = Regsit_Date if ( Triage_Date == . & Triage == . &
ADT_Decision_Date < . )

count if ( Triage_Date == . & Triage == . & Depart_Date < . )
* 65

tabulate DepartureStatus if ( Triage_Date == . & Triage == . & Depart_Date < . )
* same as ...
tabulate DepartureStatus if ( Triage_Date == . & Triage == . )
count if ( Triage_Date == . & Triage == . & DepartureStatus != "" &
DepartureStatus != "NCR NT - PT LT BEFORE BEING TRIAGED(OP)" )
replace CommentsTime = "Discharged (not LWBS) but no triage date or " +
"value: Replaced Triage date with Registration date"
if ( Triage_Date == . & Triage == . )
& DepartureStatus != "" &
DepartureStatus != "NCR NT - PT LT BEFORE BEING TRIAGED(OP)" )
replace Triage_Date = Regsit_Date
if ( Triage_Date == . & Triage == . )
& DepartureStatus != "" &
DepartureStatus != "NCR NT - PT LT BEFORE BEING TRIAGED(OP)" )

* Patients who "should have been" triaged (at least implicitly)
count if ( Triage_Date == . & Triage == . & AssRoom_Date < . )
* 315
count if ( Triage_Date == . & Triage == . & PhysAssess_Date < . )
* 0
count if ( Triage_Date == . & Triage == . & CDU_In_Date < . )
* 6
count if ( Triage_Date == . & Triage == . & OBS_In_Date < . )
* 123
replace CommentsTime = "On treatment path but no triage date or value: " +
"Replaced Triage date with Registration date"
if ( Triage_Date == . & Triage == . &
AssRoom_Date < . | PhysAssess_Date < . |
CDU_In_Date < . | OBS_In_Date < . )
replace Triage_Date = Regsit_Date
if ( Triage_Date == . & Triage == . &
AssRoom_Date < . | PhysAssess_Date < . |
CDU_In_Date < . | OBS_In_Date < . )

tabulate DepartureStatus if ( Triage < . )

tabulate PresentingComplaint if ( Triage_Date == . & Triage == . )
tabulate DepartureStatus if ( Triage_Date == . & Triage == . &
DepartureStatus != "NULL" )
replace Triage_Date = Regsit_Date if ( Triage_Date == . & Triage == . )

* Numeric versions of variables that are "generic" and of likely use in any
* analysis.
* These include Sex, Admitted (yes=1), CDU patient (yes=1) and
* Observation patient (yes=1)
* Age is already numeric. Age groups depend on the analysis so are not included
* here
* Residence is coded as NS, vs Canada, vs USA, vs Other. Analyses below the
* level of NS will depend on the institution(s) used.


```

... it goes on and on and on ...

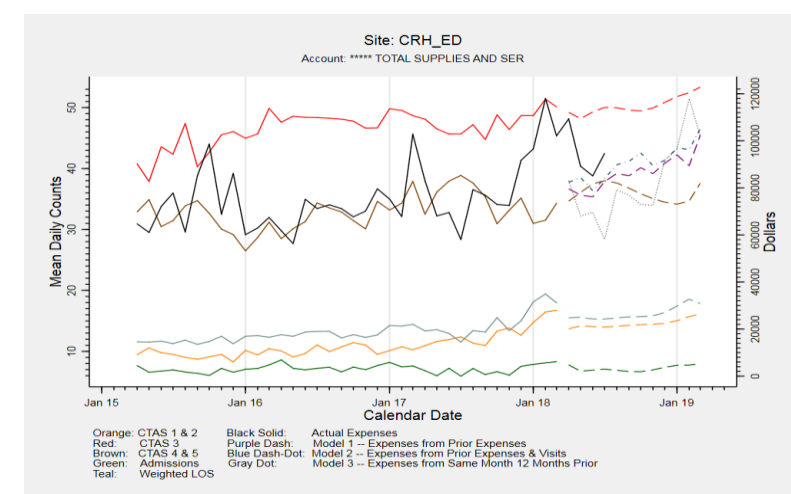
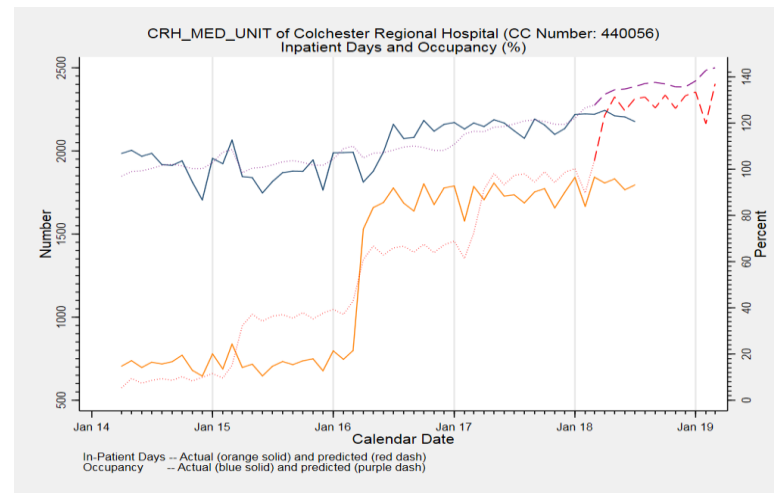
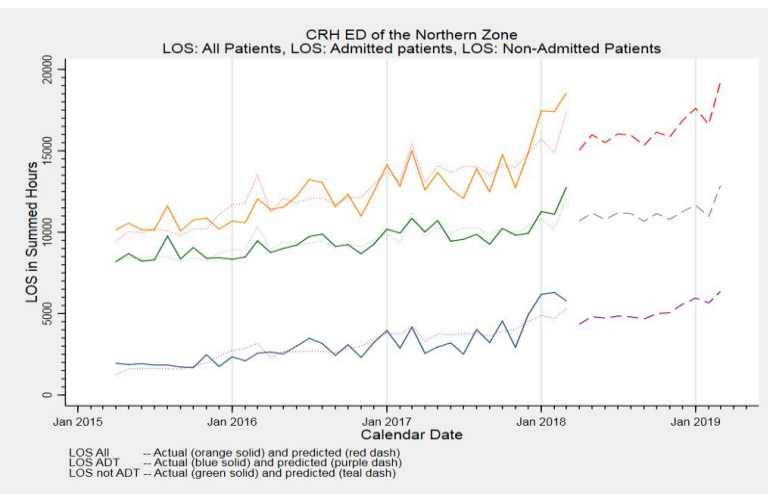
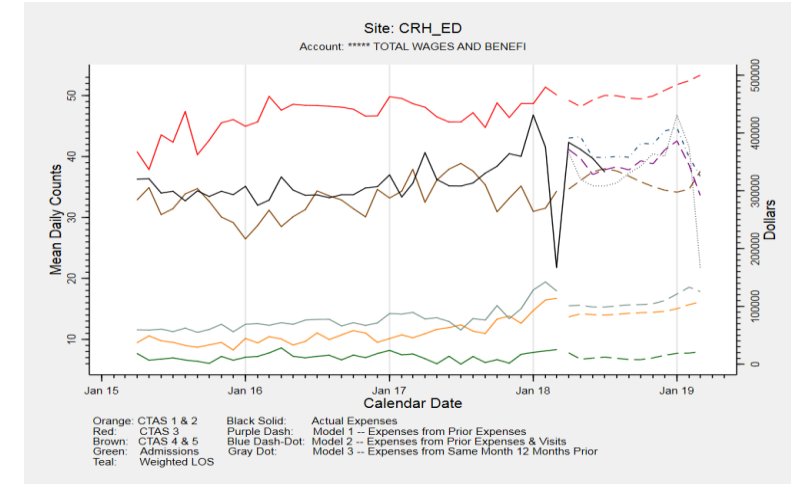
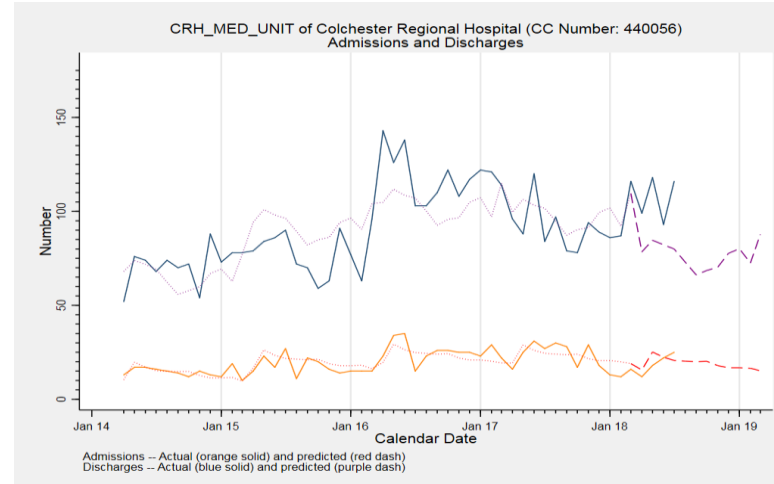
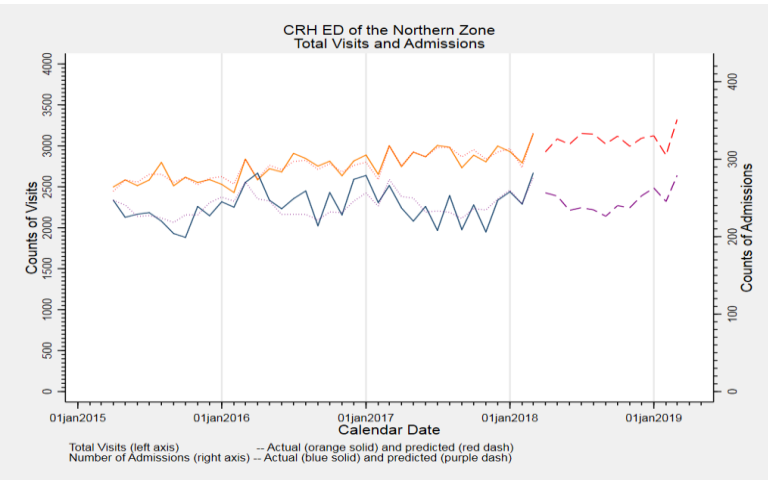


# What Have We Done?

## ED Visits, Admissions, LOS

## In-Patient, Admissions, Transfers, Discharges, LOS

## Expenses by Unit and Account



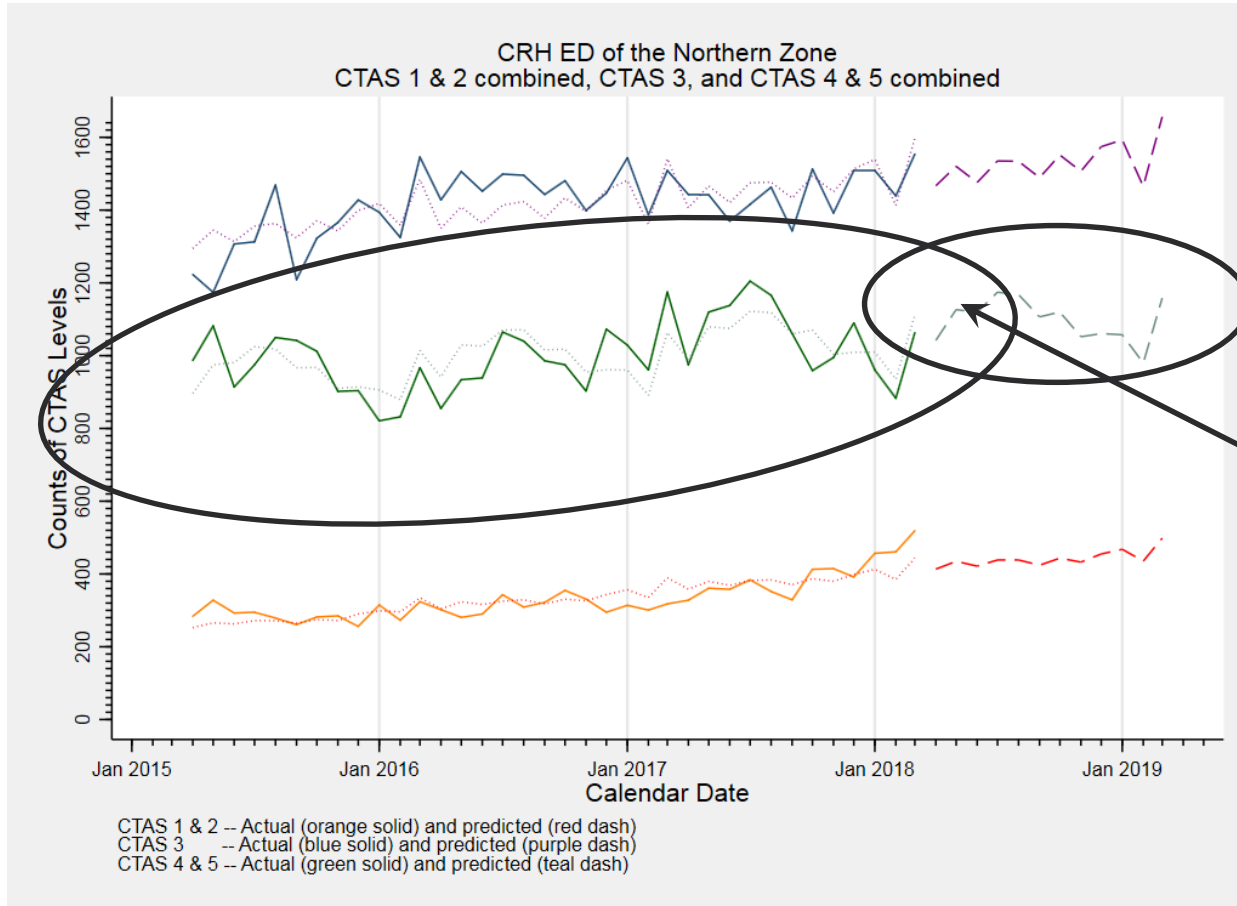
“all”

“all”

“pilot”

# Forecasting

Using an Understanding of the Past to Predict the Future



## Forecasting

- A basic extension of the model to future dates
- All IV (predictors) of the model must also exist in the future.

## Testing

- Models should always be tested on data that was NOT used to build the model.
- Fortunately, data is always coming in.

# Forecasting

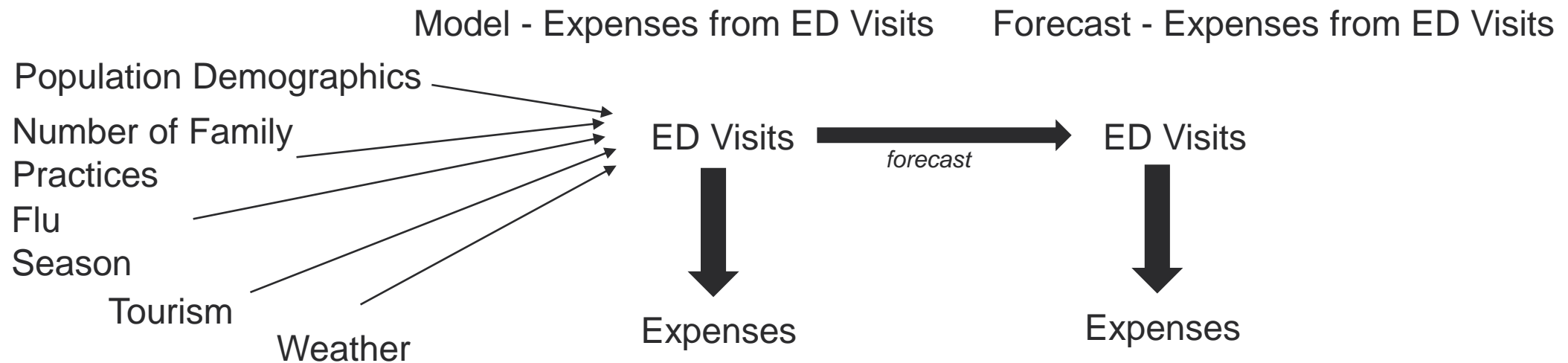
The quality (utility) of the forecast depends on the quality of the model,

**But**

Forecasting is not the same as modelling.

All the variables used in the model must exist in the future.

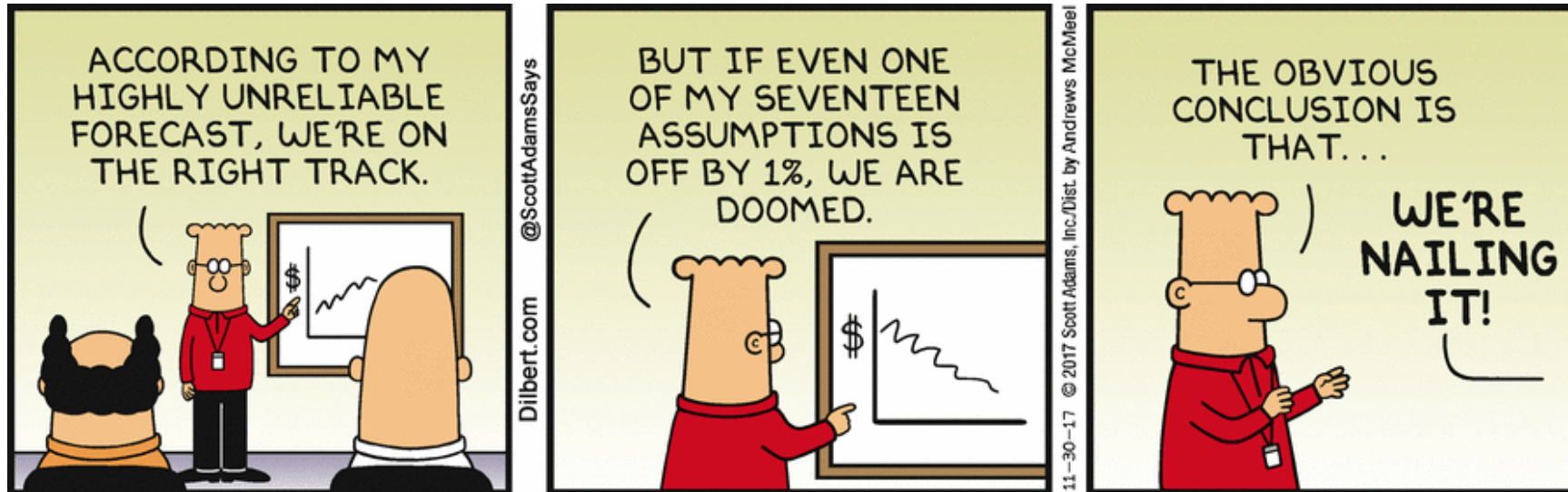
- Must be **KNOWN**.
- Must be *predictable*.





# Modelling and Forecasting

## Assessing the Analysis



- Can you describe the sources of data in the analysis?
- Are you sure the data (sample data) is representative of the population?
- Are there outliers in the data distribution? How did they affect the results?
- What assumptions are behind your analysis?
- Are there conditions that would make your assumptions invalid?

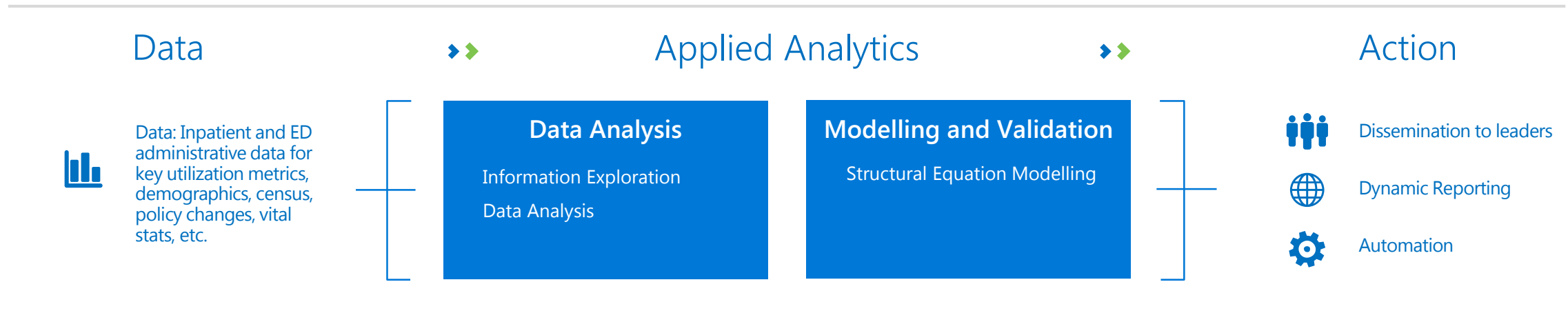
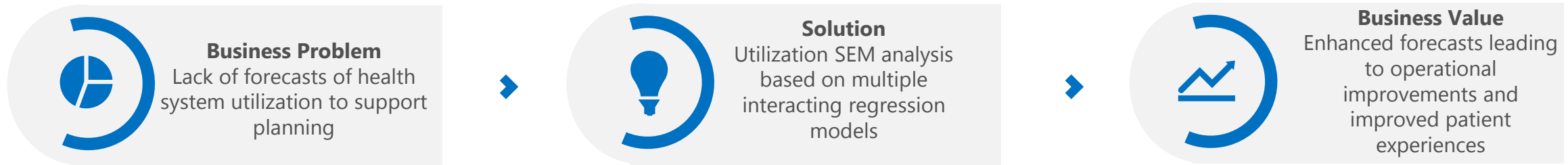
# Modelling and Forecasting Next Steps

Lots to be Done

- New variables / better variables
- New models / better models
- Application to other services
- Partnership with other teams



# Applied Analytics | Utilization Forecasting - Structural Equation Modelling



## Data Requirement

Record level and aggregate data from multiple systems and sources, complex ELT processes to clean and normalize data

## Modeling

### Examination of multiple models to produce analysis

- Path Analysis
- Confirmatory Factor Analysis
- Structural Equation Modeling
- Fuzzy Analytics Hierarchy Process (abstracting a system goal for optimization purpose)

## Improved Forecasting

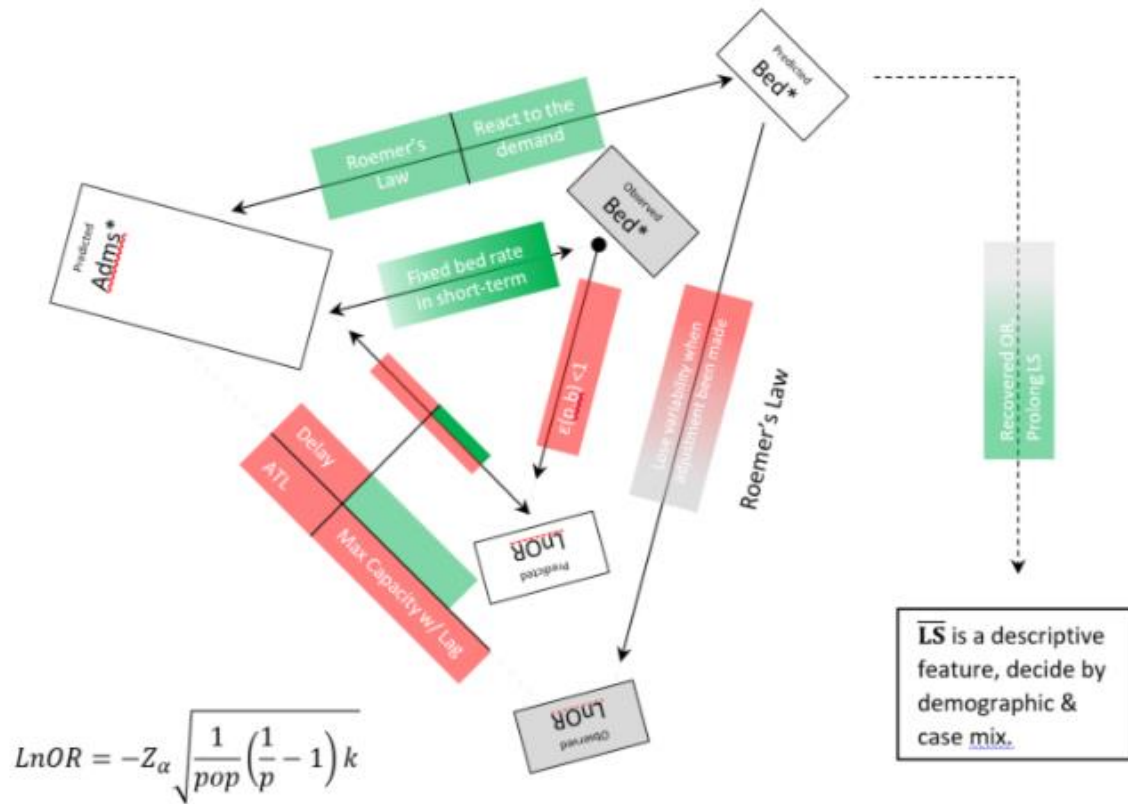
### Potential Short-Term Forecasting

- Explaining differences between areas
- Cross area spill-over effects
- Demand fluctuation
- Population emergency sensitivity
- Boarding process efficiency
- Bed substitutability

### Potential Long Term Forecasting

- Identify effects both temporal and spatial wise
- Bed pooling strategy
- System-wide optimization

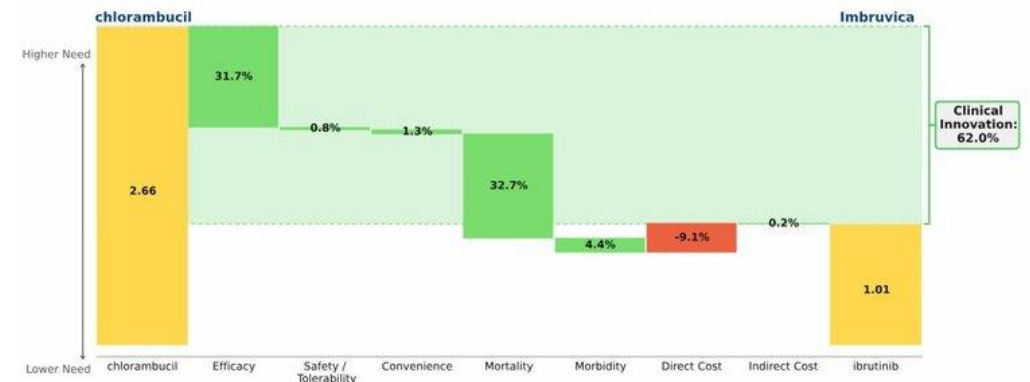
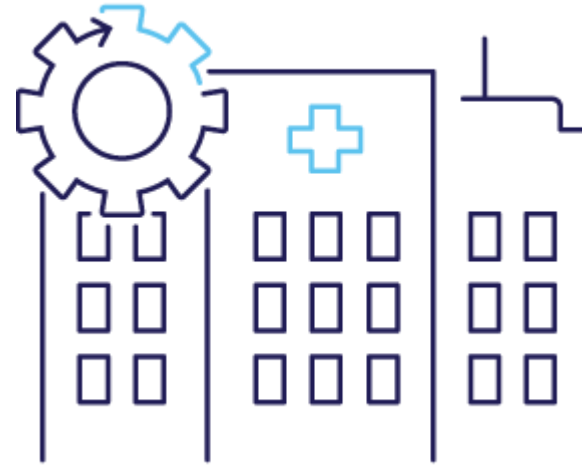
# System Utilization Forecasting using Structural Equation Modeling



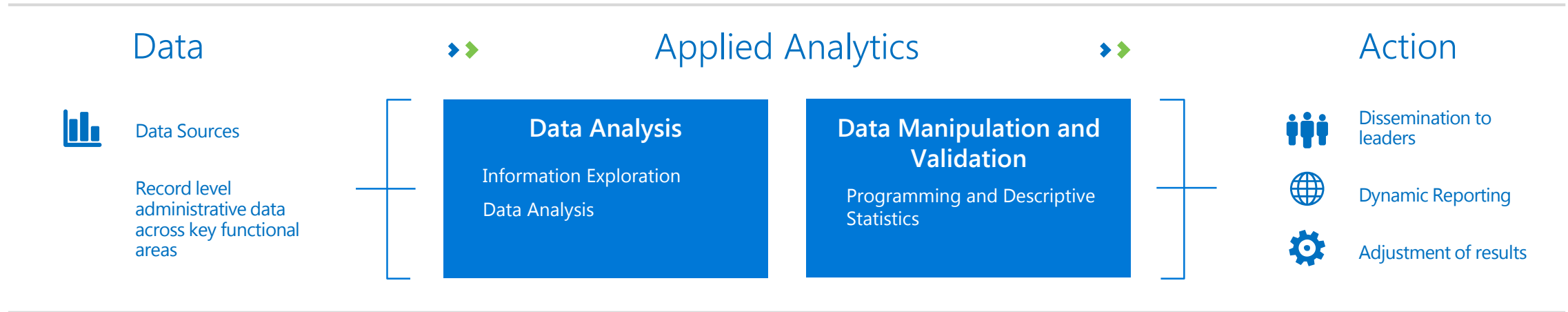
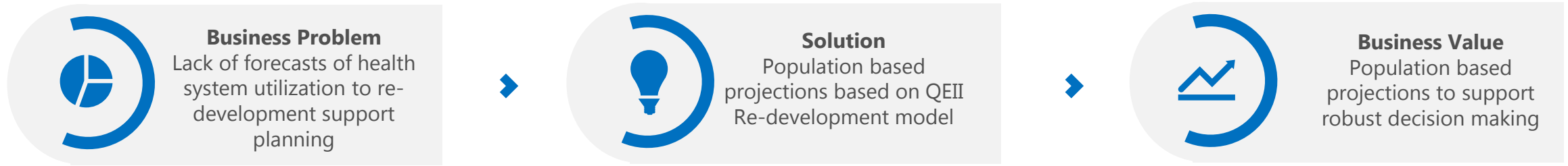
- The purpose of this project is to develop a predictive model to investigate and quantify the **interactions** among key utilization indicators both in short (inner) and long (outer) term.
- Focused on exploring the demand and supply dynamics behind the everyday individual activities using structural equation modeling

# System Utilization Forecasting using Structural Equation Modeling

- This project will introduce a different methodology with additional explanatory factors.
- The model will be used to identify key attributes for explaining historical congestion events, identifying sub-optimal areas and conducting optimization and “what-if” scenario simulation.



# Applied Analytics | Population Based Utilization Projections



## Data Requirement

Record level data identifying clinical details, demographics and geography

## Modeling

Population models to produce forecasts

Data management and programming problem with some modeling of population

## Improved Forecasting

**Potential Short-Term Planning**

Data to support functional and master planning for re-development

**Potential Long Term Planning**

Greater understanding of what we can expect if we deliver services in current models of care

# Our Approach

## Multi-step analysis of historical and projected data as inputs

This approach is focused on providing as valid as possible a sense of the future need within the catchment area, through an iterative (but, consistent algorithmic) process to meet the project and other planning needs.



### POPULATION PROJECTIONS

Utilized NS Dept. of Finance projections to 2028 by sex and 5 year age groupings for NS at the county level, projections extended to 2035.



### UTILIZATION PROJECTIONS

To project (forecast) the utilization, there were two main processes. “Baseline projections” and “adjusted projections”. The main problem with both is not the process, rather the number of variables to be considered.



### BASELINE PROJECTIONS

The first step computed the percentage of the population that used each service. These computations were conducted within each age group, sex and catchment area separately. The count was then projected into the future within each grouping.



### ADJUSTED PROJECTIONS

The adjusted projections are not yet started. The adjusted projections take the baseline projections and then adjust based on in year and historical trends as well as adjustment factors around clinical or patient flow efficiency.



### ITERATIVE ANALYSIS

This analysis is designed to be iterative, we will deliver the baseline projections and adjust the analysis to meet the project needs as necessary.

# Population Projections

## Statistics Canada:

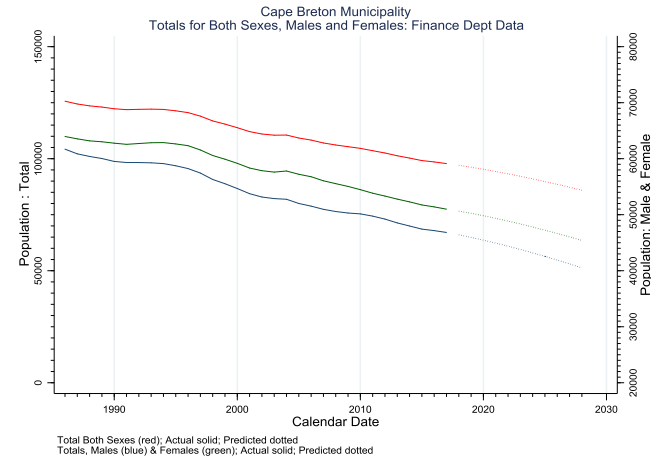
- NS historical population data from 1986 to 2017
- by sex and age group

## NS Department of Finance and Treasury Board:

- converted to the county level
- projections from 2018 to 2028

## NSHA Performance and Analytics

- converted to catchment areas
  - CBRM
  - Rest of Cape Breton
  - Rest of Nova Scotia
  - Canada
  - Other
- extended to 2035 within each Sex, Age Group, and Catchment area





# From Population to Utilization Projections

Age groups were collapsed

- adjacent levels combined
- retained the Statistics Canada boundaries
- clinically relevant age groups

Final Population Categories

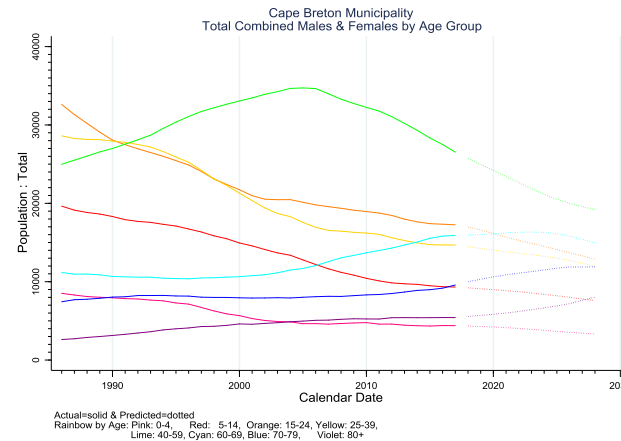
- 5 Catchment Areas by 8 Age Groups by 2 Sexes
- 80 combinations

Converted Population Level to Change in Population Relative to 2017

- percentage of the population in 2017

Forecasted utilization was computed as:

- $\% \Delta \text{Utilization} = \% \Delta \text{population} * \text{utilization in 2017}$ 
  - $\% \Delta$  is percentage change
  - utilization is "visits" or "admissions" or "discharges" etc.
  - population is each Sex by Age Group by Catchment Area



# Example: Forecasting Utilization in the CB ED

## Number of Patient Visits

### Population in 2017, and Population Forecasts

- grouped by:
  - catchment areas
  - age groups
  - sex

### Emergency Patient Data from Meditech

- data cleaning
- grouped by:
  - catchment areas
  - age groups
  - sex

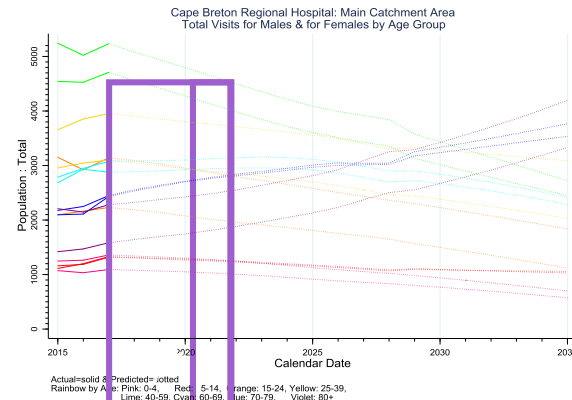
### Visits in 2017

- convert to proportion of population
- create forecast

### Forecasted Visits in FY2018 – 2035:

$$\text{Visits}_{\text{Future}} = \text{Population}_{\text{Future}} * \text{Visits}_{2017} / \text{Population}_{2017}$$

- With each age and sex group



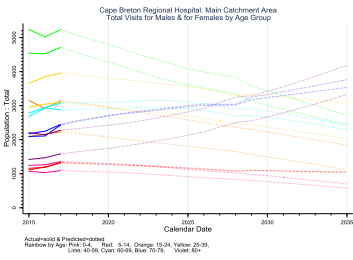
Historical Data

Base Year for Prediction (2017)

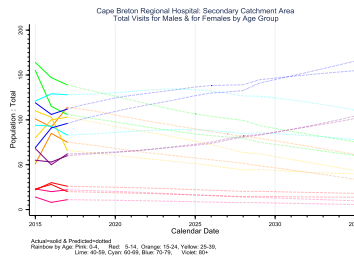
# Example: Forecasting Utilization in the CB ED

## Repeat for Catchment Areas 2 & 3

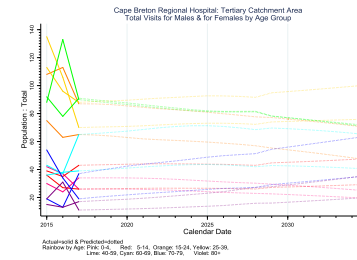
Main Catchment Area  
Cape Breton Regional Municipality



Secondary Catchment Area  
Victoria, Inverness, Richmond Counties



Tertiary Catchment Area  
Rest of NS



Collapse

over catchment area, age group, and sex

# Example: Forecasting Utilization in the CB ED

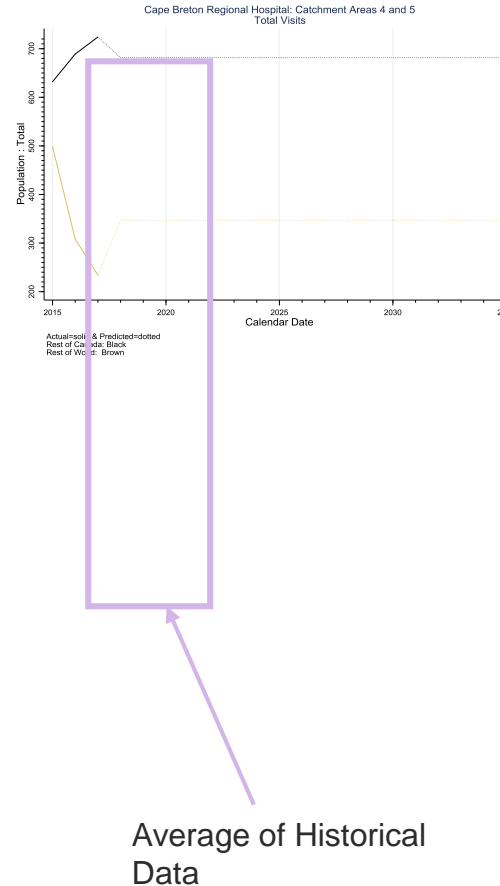
## Catchment Areas 4 (Canada) and 5 (World)

### Catchment Areas 4 & 5

1. Cape Breton Regional Municipality
2. Victoria, Inverness and Richmond Counties
3. Nova Scotia excluding Cape Breton
4. Canada excluding NS
5. the world excluding Canada

### Get Utilization

1. collapse over sex and age groups
2. assess average number of visits prior to 2018
3. use that average as the forecast

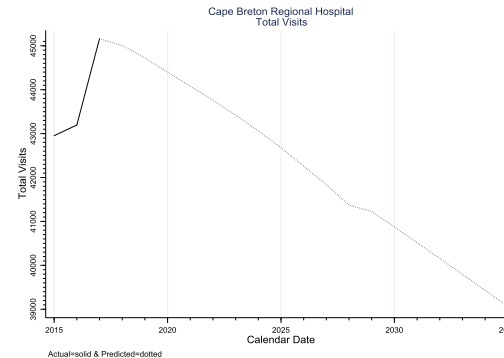


# Example: Forecasting Utilization in the CB ED

## Final Forecast for Visits

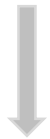
Have Forecasts for Catchment Areas 1, 2, 3, 4 & 5

1. Cape Breton Regional Municipality
2. Victoria, Inverness and Richmond Counties
3. Nova Scotia excluding Cape Breton
4. Canada excluding NS
5. the world excluding Canada



Collapse

collapse over all catchment areas



Repeat for other Facilities

- New Waterford Consolidated Hospital,
- Northside General Hospital,
- Glace Bay Hospital



Repeat for other Utilization Metrics

- Visits by CTAS, Admissions, LOS, Required Beds

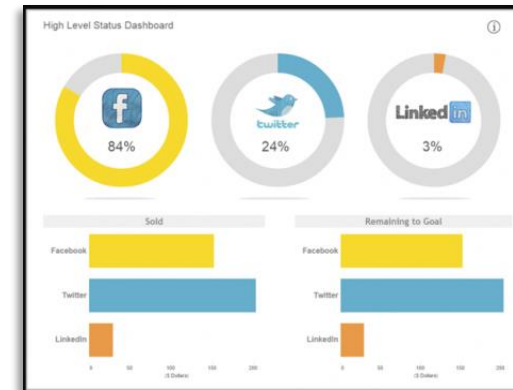
# Adjustments

Forecasts were packaged as Excel and Tableau files

- Presented to relevant redevelopment teams

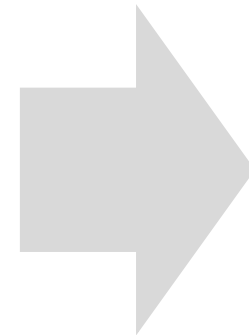
Various meetings (& emails) to:

- explain the process
  - particularly the “roll-up” of services
- assess the “validity” of the results
  - “face validity”
  - various checks
    - most teams could provide their own analysis of visits
    - current bed requirements vs forecasted



Adjustments

- resolving discrepancies
- combining or dropping particular services
  - often related to observations of data prior to 2017
- splitting services into finer categories
- use of a different base year



Re-analysis

- not usually the entire process
  - there were built-in “break points”



More meetings

# General Comments on the Population Based Projections Process

A lot of data manipulation “programming”, not a lot of statistics

Could not be done in Excel, was done in Stata

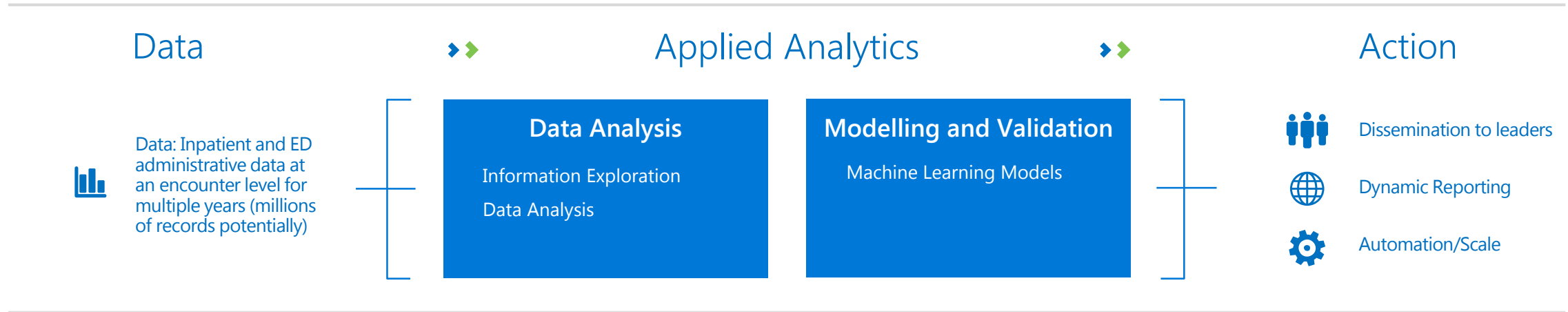
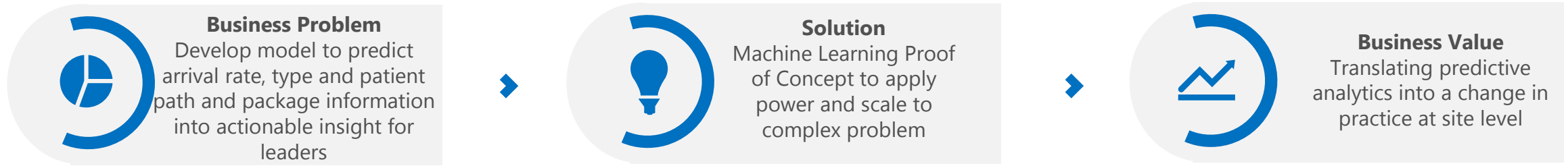
- could be done in SAS, SPSS, R, C/C++, Python, etc.

Requires automation for efficiency and scale in the interim relies on re-usable code

- automation is required
  - the volume of work
  - generalizations across units/services
- process is iterative at all stages
  - data cleaning
  - roll-ups
  - need to re-run analyses several times



# Applied Analytics | Predicting Patient Flow



## Data Requirement

Record level ED and Inpatient data for pilot sites, complex ELT processes to prepare and anonymize data

## Modeling

Examination of multiple models to produce analysis

Classification	Understanding flow of patients and patient level bottleneck in a dynamic and scalable model
Prediction	
Clustering	

## Improved Forecasting

<b>Potential Short-Term Forecasting</b> Predicting LOS and Outcomes at patient level	<b>Potential Long Term Forecasting</b> Identify clusters of patients with common characteristics and planning appropriate interventions
---	--



# Predicting Patient Flow

## Proof of Concept and EXTRA Project

- Predictive statistical models have been shown to predict patient flow with fairly high accuracy
- There have also been specific examples, including in the Canadian context, of how applied analytics can improve understanding of an impact of patient flow and decision-making.
  - Humber River Hospital Command Center – this is describes as a ‘data-driven mission control’: <https://www.hrhfoundation.ca/commandcentre/>
  - Institute for Clinical Evaluative Sciences: <https://www.ices.on.ca/Publications/Journal-Articles/2012/January/Evaluating-the-effect-of-clinical-decision-units-on-patient-flow-in-seven-Canadian-emergency>, <https://www.ices.on.ca/Publications/Journal-Articles/2017/July/Emergency-department-flow-measures-for-adult-and-pediatric-patients-in-British-Columbia-and-Ontario>
  - Presentation at the most recent HIMSS international conference titled ‘Deriving value from patient flow analytics’: <https://www.himssconference.org/session/deriving-value-patient-flow-analytics>



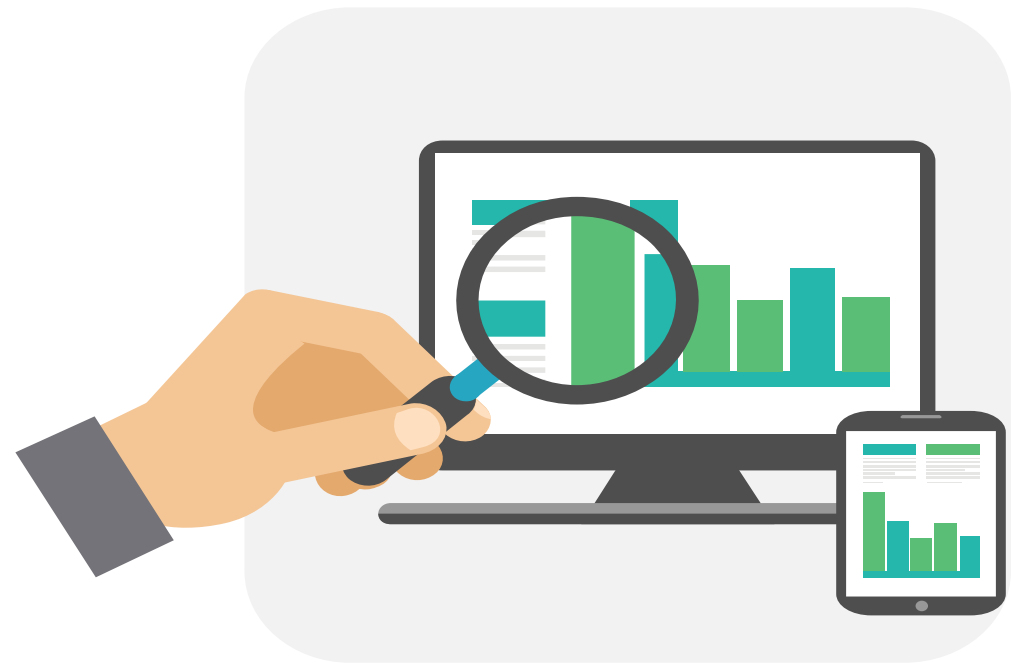
# Applicability and Feasibility



- This improvement project is one of the first at NSHA.
- Prior to now, we have had limited analytical capacity to take on this time of project.
- The skill sets of our team members in terms of data integration and analytics have significantly increased.
- Furthermore, core data elements from disparate information systems have been harmonized and are well understood.

# Data Is Worthless If You Don't Communicate It

- To often we compile vast troves of data and analytical reports that never see the light of day
- We need to communicate findings well, rather than assuming the results speak for themselves
- Simple communication model
  - Understanding of problem
  - Measurement of impact of problem
  - What data is available
  - Initial solution hypothesis
  - The Solution
  - The impact of the Solution



# Decisions Don't Start with Data



- Data can provide new insights and evidence to inform our toughest decisions, but numbers alone won't convince others.
- Good stories – with a few key facts woven in – are what attach emotions to your argument, prompt people into unconscious decision making, and ultimately moves them to action.

# Continuing Education

Let's Talk Informatics has been certified for continuing education credits by;

- College of Family Physicians of Canada and the Nova Scotia Chapter for 1 Mainpro+ credit.
- Digital Health Canada for 1CE hour for each presentation attended. Attendees can track their continuing education hours through the HIMSS online tracking certification application, which is linked to their HIMSS account.



**THANK YOU**

Question ?