



Evolution of Analytics in Prescription Monitoring in AB



Health Analytics Community, Canada Health Infoway

Dr. Salim Samanani, CEO and Medical Director

salim@okaki.com

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Who We Are

- OKAKI is Blackfoot for “Be Wise”
- Public Health Services and Informatics Social Enterprise, Founded in 2008
- Core Competencies
 - Public Health, Healthcare, Health Systems
 - Health Information Privacy and Security
 - Data Engineering, Analytics and Health Research
 - Software
 - IT Infrastructure and Operations
- Areas of Activity
 - Indigenous Health
 - Public Health and Immunizations
 - Prescription Monitoring
 - Population Precision Health



Tracked Prescription Program (TPP) Alberta (est. 1986)

- Provincial Rx drug monitoring program
- Administered by CPSA, TPP AB is a multi-stakeholder partnership
- Monitors medications prone to non-medical or inappropriate use
- Works with stakeholders to enable system level change to ensure appropriate use of monitored medications.

Physician Prescribing Practices (PPP) Program (est. 2009)

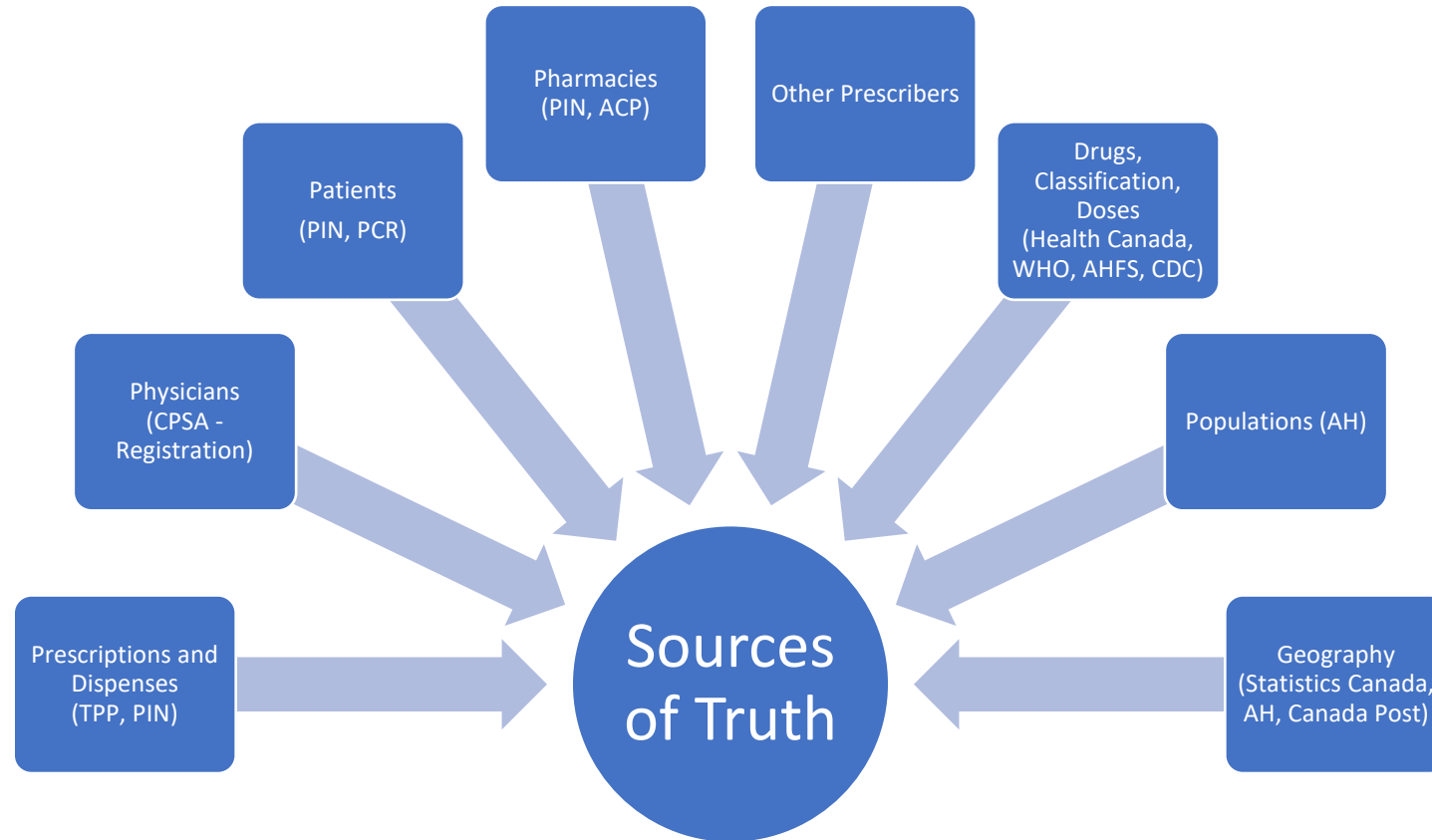
- Quality improvement program that provides educational support to all physicians generally to optimize quality of care provided to Alberta patients (introduced 2009)
- Direct MD and practice interventions for higher risk cases



TPP Member Organizations



Data Sources Used Today

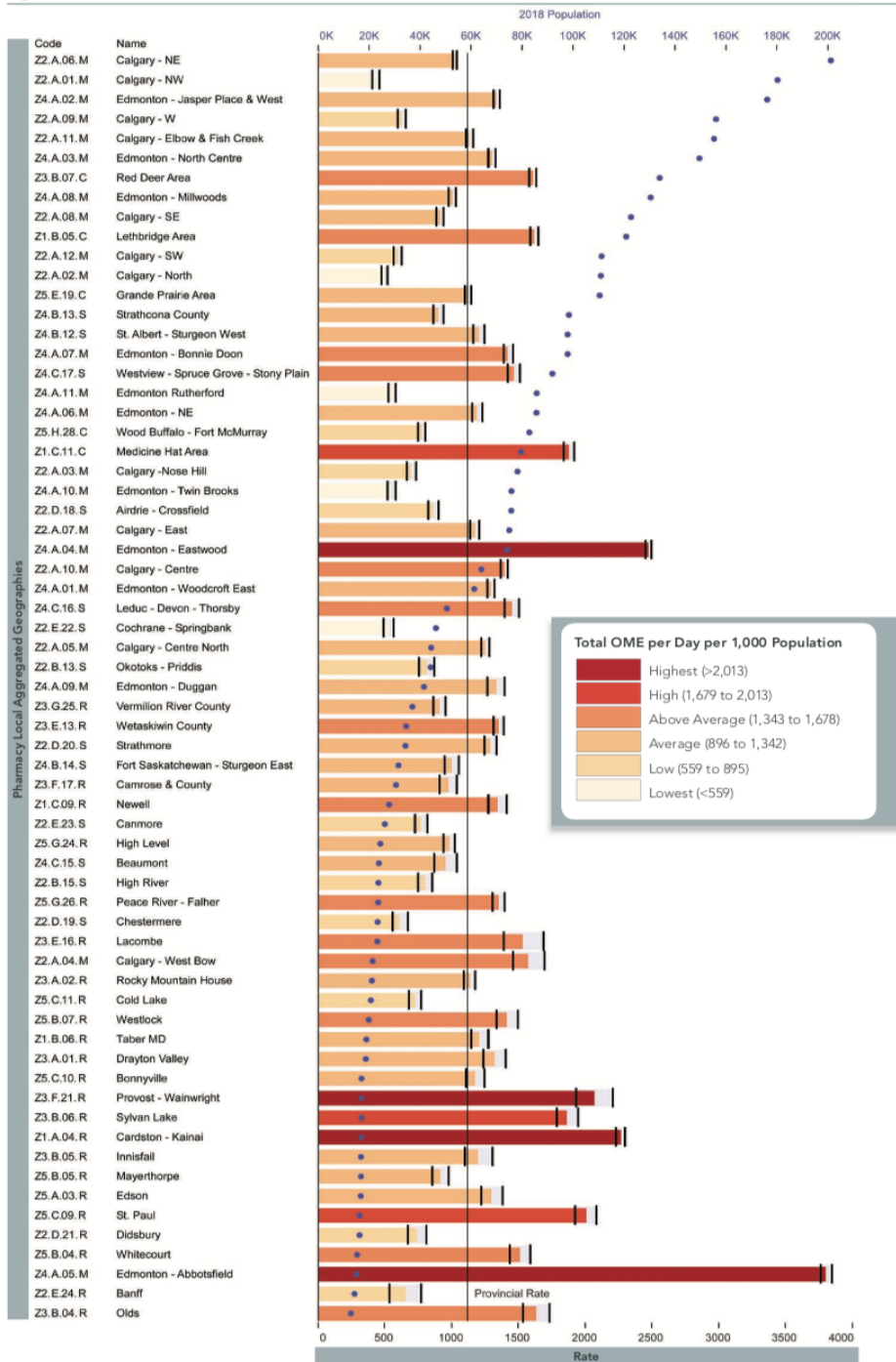




Aspects of our Work

Atlases – Population Trends and Profiling

Figure 6a. Age and Sex Standardized, Total OME per Day per 1,000 Population, by Pharmacy Local Aggregated Geographies, 2018*



Age and Sex Standardized Rate, Opioid Patients Who Received 90 OME+ per Day per 1000 Population by Pharmacy Local Aggregated Geography

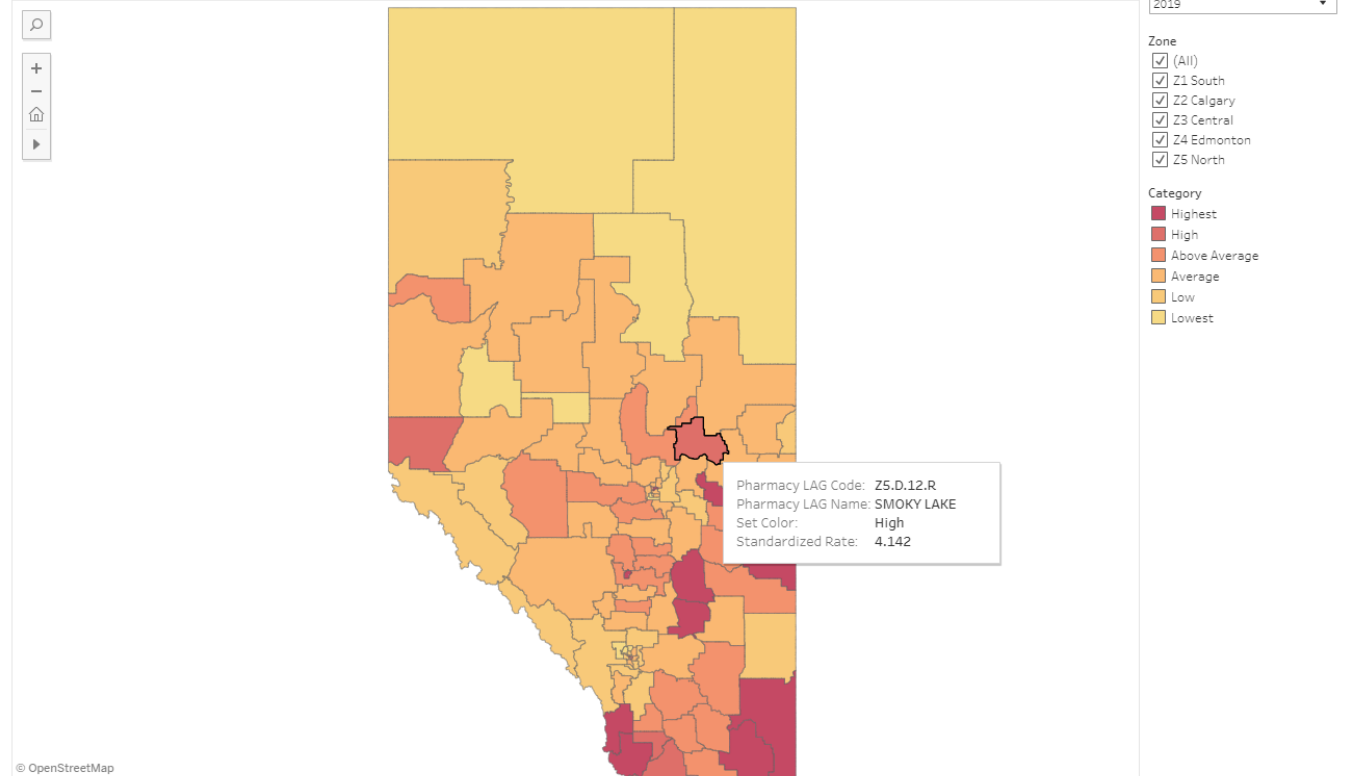
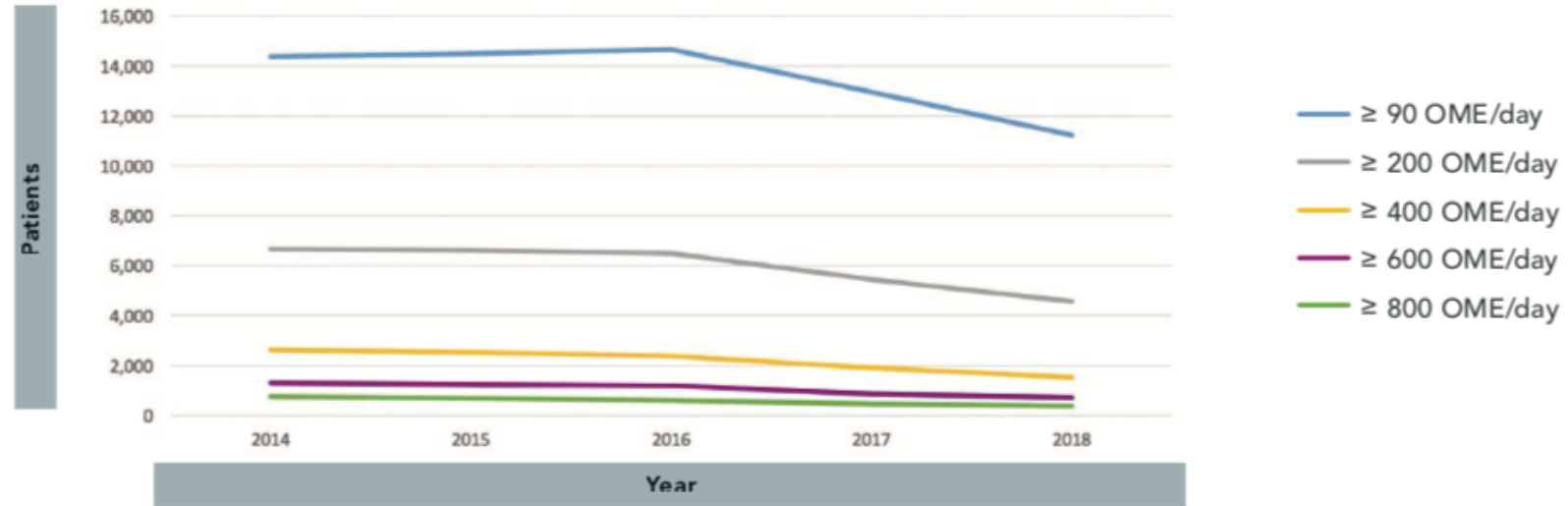
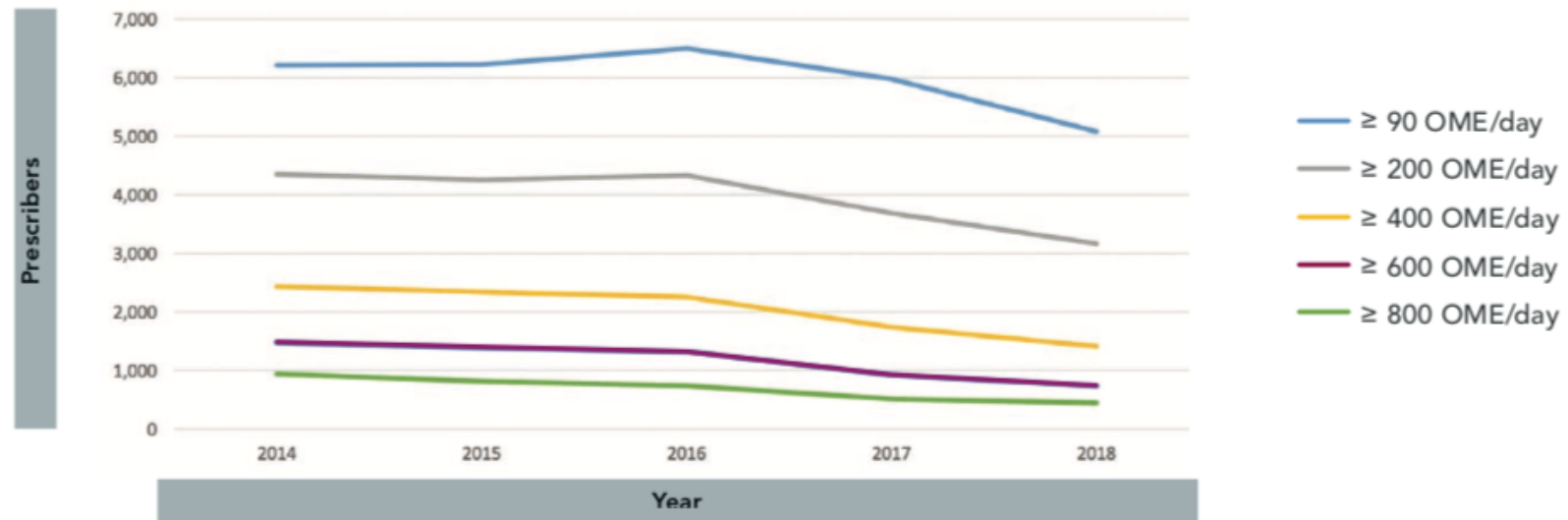


Figure 3. Opioid Patients by Dose, 2014-2018



There is a downward trend at all dosage levels since 2016. The pattern is even stronger with higher dosages.

Figure 4. Opioid Prescribers by Dose, 2014-2018





Aspects of our Work

Atlases – Population Trends and Profiling

TPP Application – Management of TPP Data and Ops



Aspects of our Work

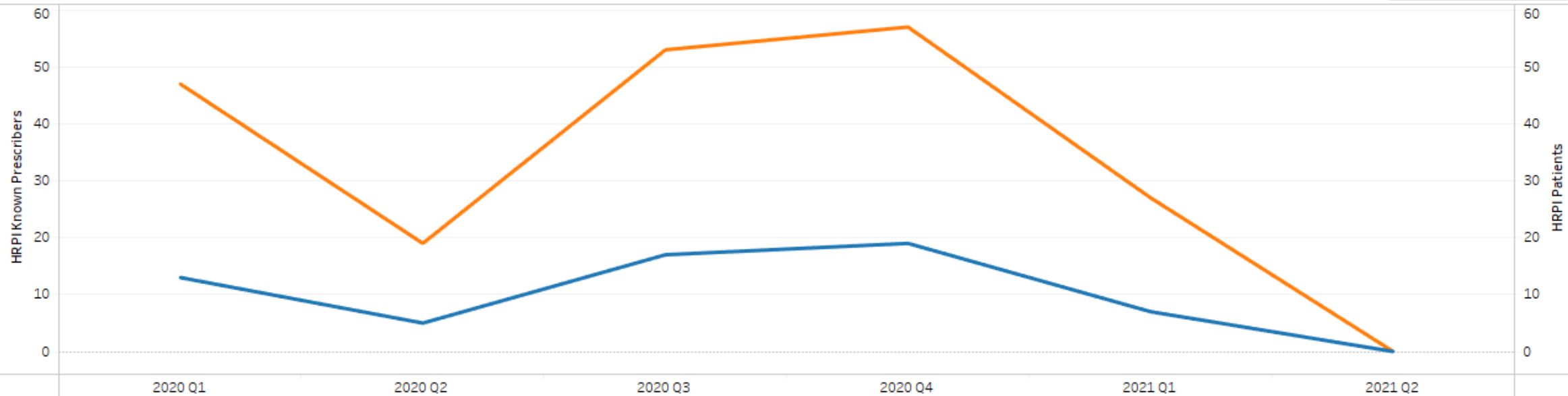
Atlases – Population Trends and Profiling

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SLICE – Patient and Prescriber Profiling

HRPI500 Patient and Prescriber Counts

Quarters to Display Last 17 quarters



Select a patient or prescriber to display profile

Data Source: PIN TPP

Year: 2020 Quarter: Q2

Select Number of Prescribers: 3

Select Number of Pharmacies: 3

Select OME/day cut off for HRPI: 500

Prescriber Type: (All) OTC Codeine: Not available.

Legend: ■ HRPI Known Prescribers ■ HRPI Patients

HRPI500 Patient-Prescriber Cases for Q2 2020

Cases: 23 (includes unknown prescribers)

PHN	Patient Name	Prescriber Name	Registration Number	Prescriber Type	Specialty	Total # Known Prescribers per Patient	Total # Pharmacies/Patient	Prescriber OME/day	Total Patient OME/day
UL110898600	Wilzak, Dunte	Rawicki-Gielniak, Agapita	021936	MD	Radiation Oncology	5	3	226	695
		Rejza, Flara	019940	MD	Family Medicine (Palli..	5	3	263	695
		Sudzinski-Parada, Michael	028896	MD	Family Medicine	5	3	16	695
		Wedkiewicz, Eura	4390	MD	Family Medicine	5	3	180	695
		Wiechec, Melinda	227	MD	Radiation Oncology	5	3	11	695
UL272848911	Markiefka-Zlotos, Auline	Prozorow, Margretta	027927	MD	Family Medicine	6	3	37	542
		Przegiend, Boyce	015736	MD	Family Medicine	6	3	237	542
		Remczynski, Bertram	008598	MD	Addiction Medicine	6	3	116	542
		Repa, Hildagarde	029039	MD	Family Medicine	6	3	5	542
		Staszecki, Belle	026304	NP	None	6	3	9	542
		Wodara, Ednah	100777	MD	Family Medicine	6	3	137	542

Prescriber Summary

Click the orange buttons below to view more information

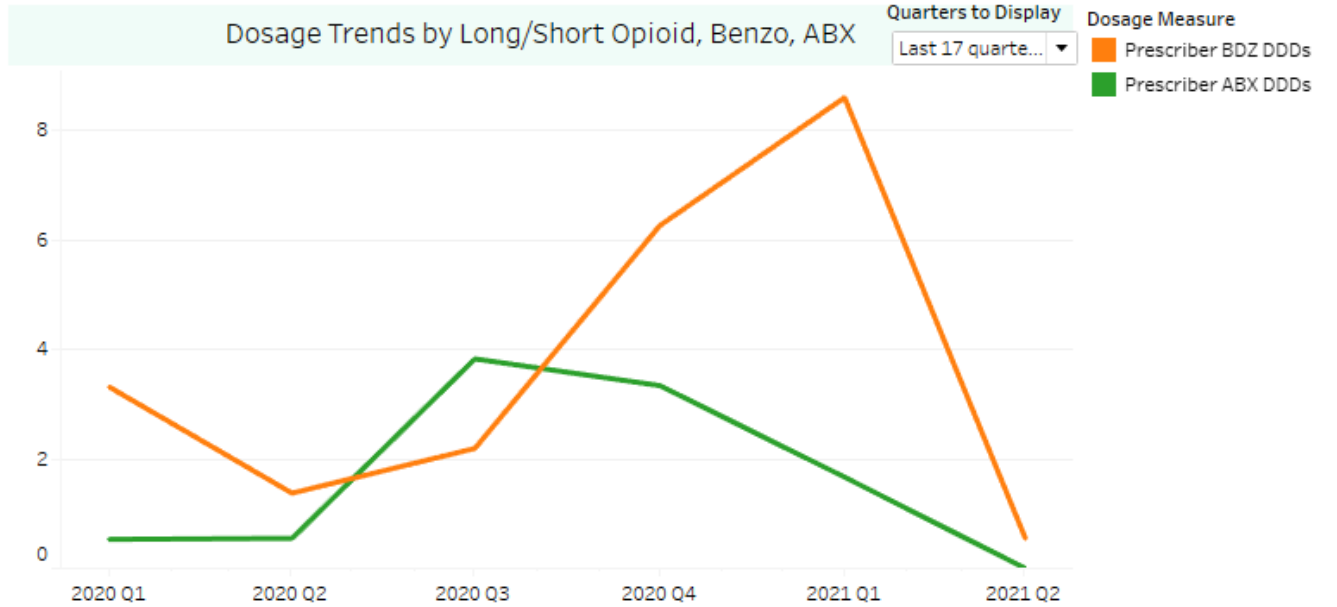
Year - Quarter	Registration Number	Prescriber Name	Registration Number	Practitioner Geography	Prescriber Age	Prescriber Type	Specialty
2020-Q3	017748	Wojczal, Lindsey	017748	CALGARY - WEST BOW	Null	MD	Family Medicine

Category	Value
Dispenses	80
Total Patients	64
ABX Patients	47
Active ABX Prescriptions	55
ABX DDDs	3.83
Benzo Patients	6
Active Benzo Prescriptions	7
BDZ DDDs	2.19
Opioid Patients	11
Active Opioid Prescriptions	13
OME/day	42
Stimulant Patients	0
Active Stimulant Prescriptions	0
Gabapentinoid Patients	0
Active Gabapentinoid Prescriptions	0
BDZ and Opioid Patients	0
Pharmacies	41
Max. Single Patient BDZ DDDs	0.99
Max. Single Patient OME/Day	19.8
Distinct count of OAT Patients	2.00
Highest Total Dispenses From One Pharmacy	8.00

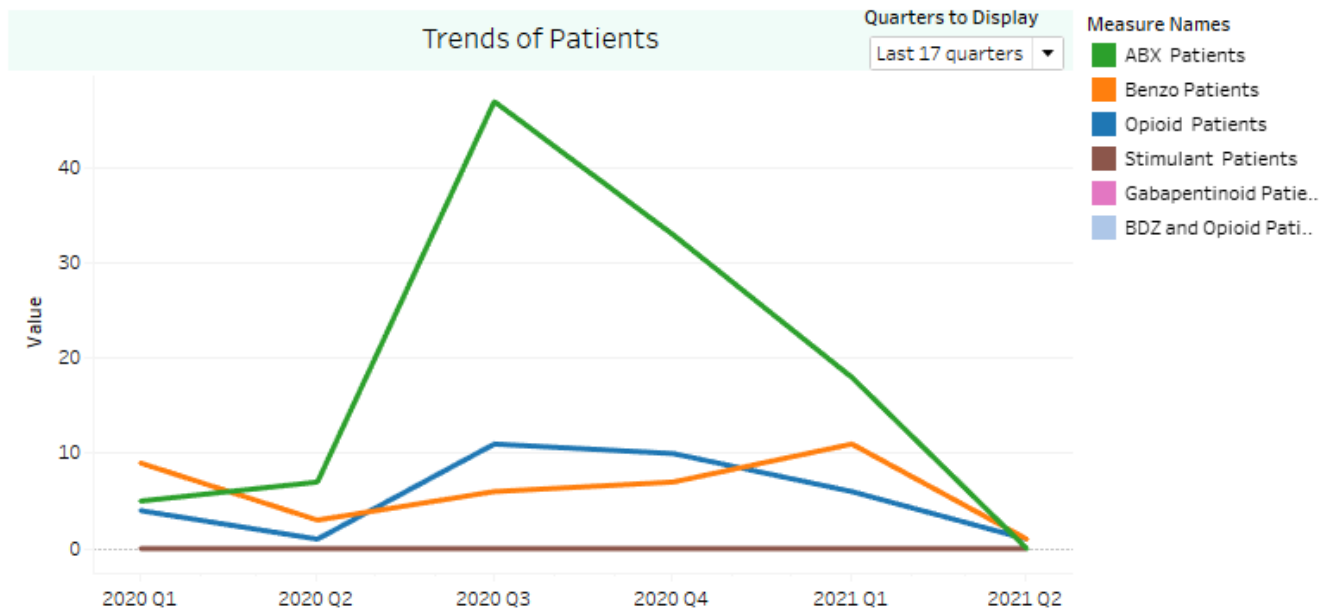
Dispense and Patient Proportions

View Pharmacies

Dosage Trends by Long/Short Opioid, Benzo, ABX



Trends of Patients





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SLICE – Patient and Prescriber Profiling

PERFORxM – High Risk Case Management

View Archives Move To Archives Change Status Add Note Assign Case Delete
Find To Excel To Word Generate Letter Reports Help



Groups

- New
- Edit
- Run
- 3+Opiods 3+BDZ
- 5 BDZ
- DOME 5000
- HRPI 600**
 - 2015-01-01 - 2015-03-31
 - 2014-10-01 - 2014-12-31
 - 2014-07-01 - 2014-09-30

HRPI 600

Select Items | View Case

Cases: 93 Selected Cases: 0

Drag a column header and drop it here to group by that column

	Case ID	Patient First Nam	Patient Last Nam	PHN	Practitioner First	Practitioner Last	Registration ID	Prac
<input type="checkbox"/>	94	Charles Harry	Evans	UL108557810	LEANNE	FAGAN	011702	Phy
<input type="checkbox"/>	95	Charles Harry	Evans	UL108557810	ALBERT R.	FUSHTEY, NP	012072	Phy
<input type="checkbox"/>	96	Charles Harry	Evans	UL108557810	MICHAEL	GINCHER		Phy
<input type="checkbox"/>	97	JEANNE	LIDDAR	AB592157521	DANIEL	DUFF	021587	Phy
<input type="checkbox"/>	98	JEANNE	LIDDAR	AB592157521	SABAHAT	GUO	016404	Phy
<input type="checkbox"/>	99	JEANNE	LIDDAR	AB592157521	STEVE	HENDSON	020609	Phy
<input type="checkbox"/>	100	JEANNE	LIDDAR	AB592157521	TODD	JOHNSON	019748	Phy
<input type="checkbox"/>	101	JEANNE	LIDDAR	AB592157521	IGNACIO III Y	KROETSCH, DDS	008985	Phy
<input type="checkbox"/>	102	JEANNE	LIDDAR	AB592157521	WILLIAM J.	MONZON	012024	Phy
<input type="checkbox"/>	103	James Joseph	MUESKE	UL209753220	ADAM KIEL	AHMED		Phy
<input type="checkbox"/>	104	James Joseph	MUESKE	UL209753220	WILLIAM	HELLQUIST, DVM	005932	Phy
<input type="checkbox"/>	105	James Joseph	MUESKE	UL209753220	MICHAEL H.C.	TODOSIJCZUK		Phy
<input type="checkbox"/>	106	Caitlin	MURRAY	UL133594300	DARREN A.	LOVE	011829	Phy
<input type="checkbox"/>	107	Caitlin	MURRAY	UL133594300	DAVID	LUC	009739	Phy
<input type="checkbox"/>	108	Caitlin	MURRAY	UL133594300	MOFOLUWASO	SAMUELSON	018146	Phy
<input type="checkbox"/>	109	Caitlin	MURRAY	UL133594300	EVAN	VAN DER MERWE	019961	Phy
<input type="checkbox"/>	110	Uvis Martin	Craig	BC012375000	RICHARD	CROCKFORD	007400	Phy

Save Delete Move View Archives Find Reports Help



- Groups
- New Edit Run
 - 3+Opiods 3+BDZ
 - 5 BDZ
 - DOME 5000
 - HRPI 600**
 - 2015-01-01 - 2015-03-31
 - 2014-10-01 - 2014-12-31
 - 2014-07-01 - 2014-09-30

HRPI 600 - Case ID: 83

Case | Status History | Occurences

Case ID: 83 **PHN:** BC9123750682 **Registration ID:** 021571
Measure: HRPI **Patient:** Huria Martin Saini **Practitioner :** JOHANN HEINRICH DRIE
Current Status: In Progress **Period Opened:** 2014-10-01 **Practitioner Type:** Physician
Assigned To: John Niruban **Practitioner Alias ID:** URID-000010516900

Practitioner Address:
430 MAYOR MAGRATH DR S, Lethbridge, AB T1J 3M1

Comments: + New

Date	Status	Detail	Created On	Created By	Modified On	Modified By
2015/05/20	Closed - Notification Sent	Comment: Repeat case.	2015/05/20	ssamanani	2015/05/20	E
2015/05/20	Closed - Notification Sent	Comment: Letters sent Feb 8, 2015	2015/05/20	ssamanani	2015/05/20	E



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PERFORxM – High Risk Case Management

MD Snapshot – Individualized MD Practice Information



Test, Prescriber

Registration Number:
tableau-admin



Quick links to summaries and resources:

Click on the menu in the top left to see other dashboard views and more

- [Opioid Naive Summary](#) →
- [Opioid Summary](#) →
- [Benzodiazepine Summary](#) →
- [Antibiotic Summary](#) →
- [Opioid Trends](#) →
- [Benzodiazepine Trends](#) →
- [Antibiotic Trends](#) →
- [Patients by Drug Ingredient](#) →
- [Supporting Material](#) →
- [View Full Prescribing Snapshot](#) →

This snapshot was generated with data current as of October 20, 2023.
Data for the most recent quarter is visible three weeks after the last day of the quarter.



Test, Prescriber

Date: 2022-Q3

Comparator Group: Specialty

Three Month Prescribing Snapshot: Antibiotics¹

	Your Practice	Comparator Group Median ²	Your Percentile
Patient(s) receiving antibiotics prescribed by you	322	322.00	322.0
Total antibiotic DDDs prescribed ³	322.00	322.00	322.0
Antibiotic DDDs/patient ³	322.00	322.00	322.0
Total antibiotic prescriptions from you	322	322.00	322.0
Average antibiotic prescriptions per day prescribed by you ⁴	322.00	322.00	322.0
Average number of prescriptions per patient receiving an antibiotic ⁴	322.00	322.00	322.0
Patient(s) receiving three or more different antibiotics ⁵	2		
Patient(s) receiving antibiotics from three or more prescribers ⁵	2		

(1) Analytic group includes drug dispenses for antibiotics, excluding topical formulations.
 (2) Information from your annual renewal (RIF) determines your assignment to a comparator group. In addition, if you belong to a Primary Care Network (PCN) and have selected the PCN comparator group, your prescribing practice is compared to physicians identified as members of your PCN. Comparator group members may have similar qualifications, clinical practices, or prescribing distribution, but variations at an individual level are possible. Please take your context into consideration as you review and interpret the information provided.
 (3) The DDD of a drug is the average daily maintenance dose for its primary use in adults. Total DDDs is the sum of the DDDs for all antibiotics dispensed under your name in the quarter, except topical formulations and where a DDD is not assigned, e.g. otic and ophthalmic preparations. DDDs serves as a proxy for the actual dose and days of treatment. Doses are measured in multiples of Defined Daily Doses (DDD). The measure must be interpreted in the context of individual practice type (e.g. episodic vs. chronic), typical number of patients seen per day and practice scope (e.g. Infectious Diseases).
 (4) Includes dispenses only. Refills or Part-fills on a given prescription are individually counted. Prescriptions that the patient does not fill are not included.
 (5) Multiple prescriber types, including other physicians, pharmacists, nurse practitioners and dentists, may contribute to this antibiotic measure.

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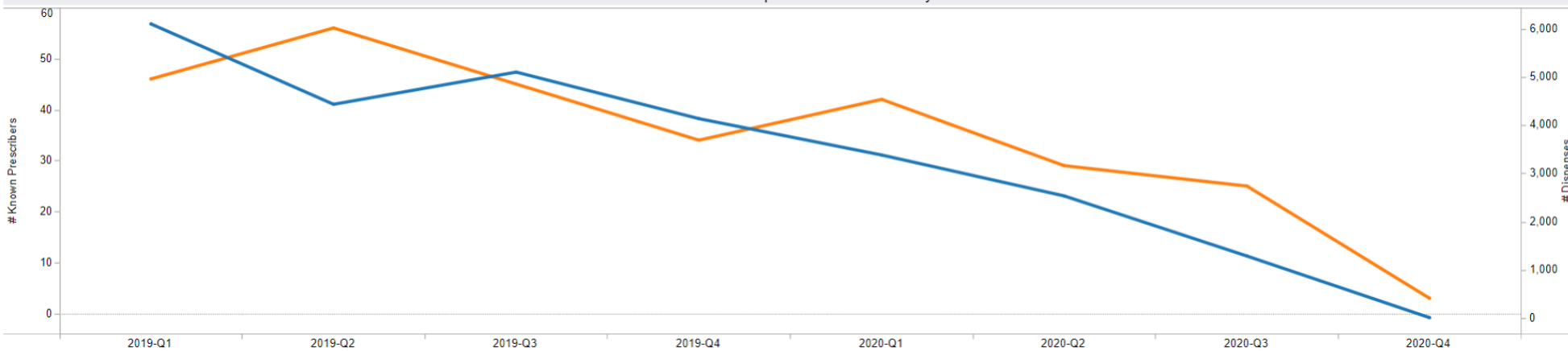
SLICE – Patient and Prescriber Profiling

PERFORxM – High Risk Case Management

MD Snapshot – Individualized MD Practice Information

SLICE – Data Quality Assurance

Prescribers and Prescriptions of Deceased Physicians



■ # Disposes
■ # Known Prescribers

Quarter **Year**

Dispenses of Deceased Physicians in Q4 2020

Disposes: 9

Prac Death Date	Prescriber Name	Registration Number	Dispense Date	Brand Name	Patient Name	PHN	Pharmacy Licence Number	Pharmacy Name	Pharmacy Phone Number	Days Between Dispense Date and Prescriber Death Date
-----------------	-----------------	---------------------	---------------	------------	--------------	-----	-------------------------	---------------	-----------------------	--

Number of Reversals After 14 Days by Year and Quarter





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Ad Hoc Requests – Anything Not Captured Above

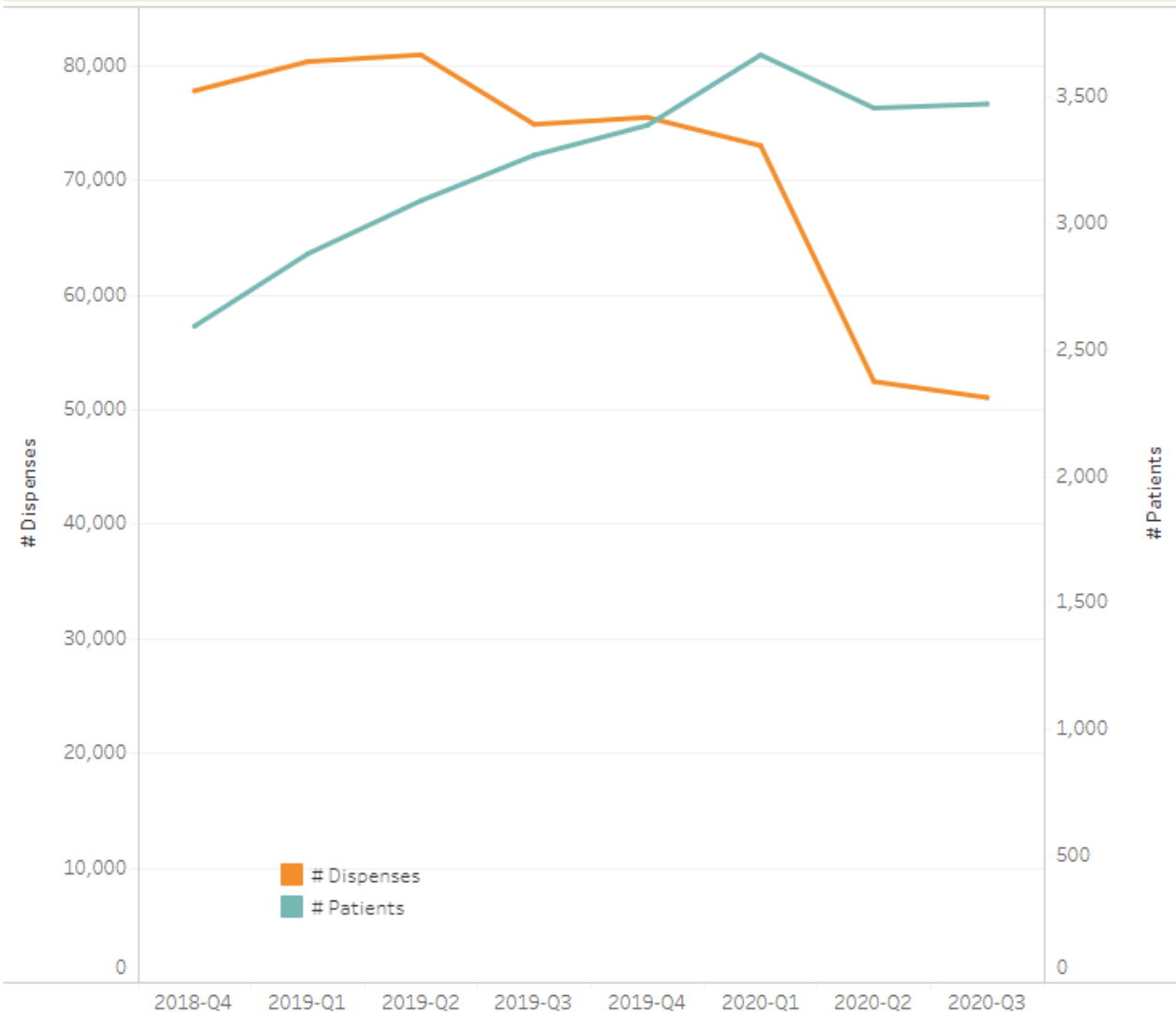
Use the filters to change the line graphs below independently

OAT Modalities Injectables
(Multiple values) Non-Injectable

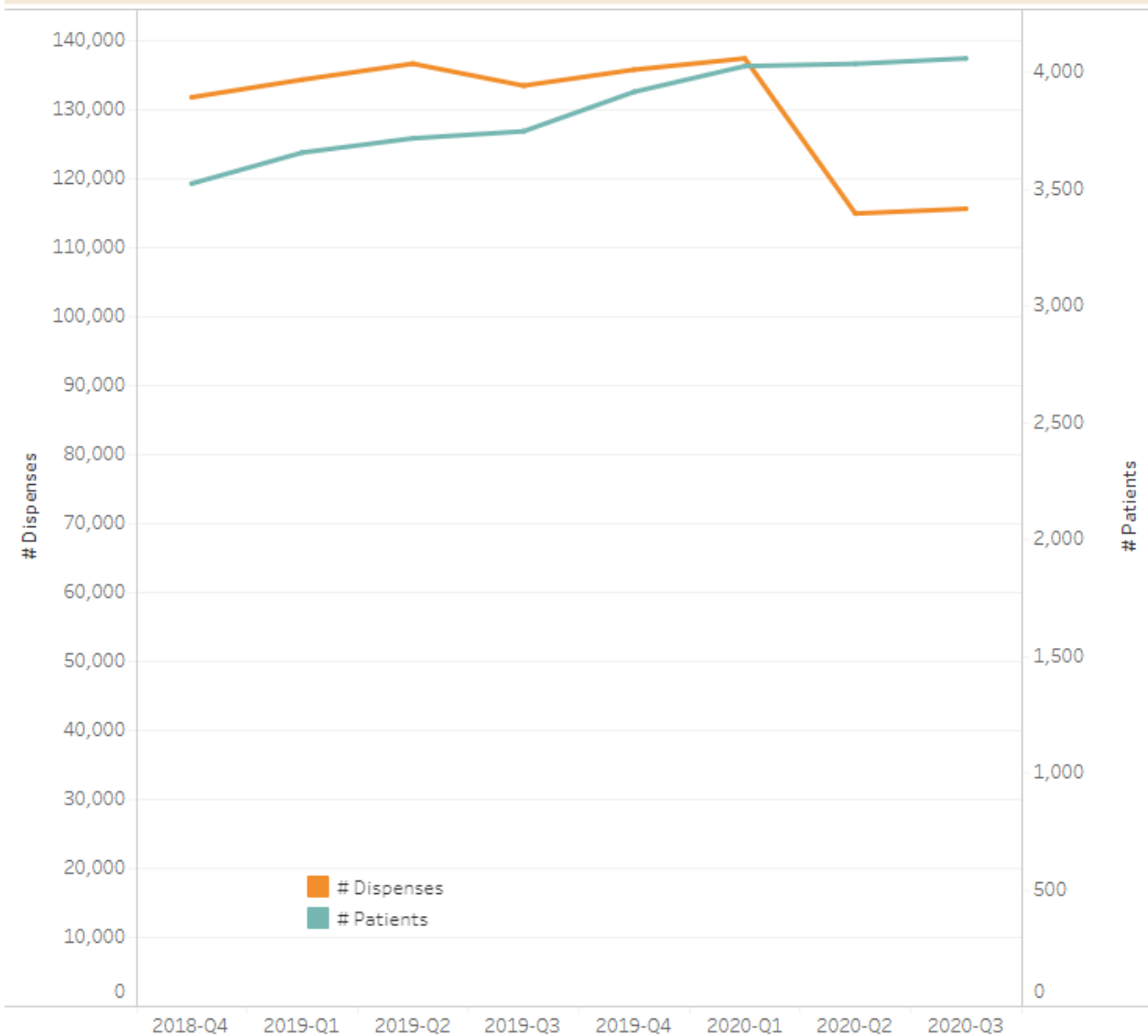
Number of Quarters to Display
Last 8 quarters

OAT Modalities Injectables
Methadone Non-Injectable

Patients and Dispenses Over Time for Buprenorphine - Implant, Buprenorphine - Subcutaneous, Buprenorphine/Naloxone



Patients and Dispenses Over Time for Methadone

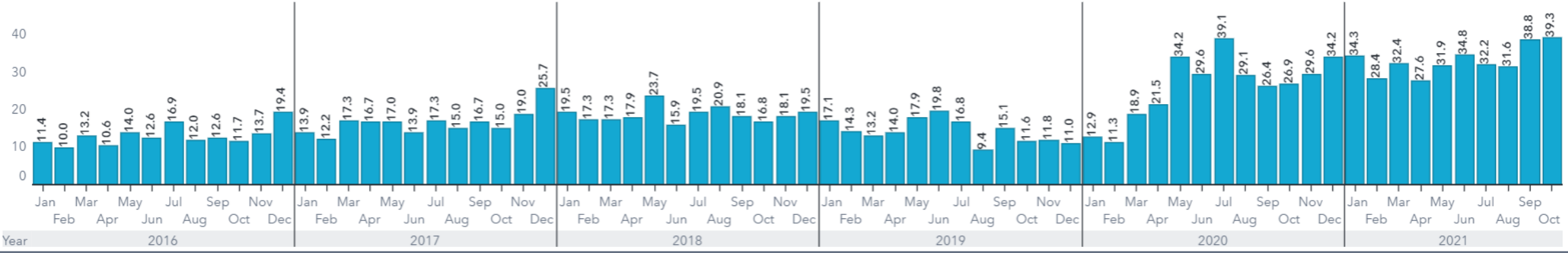


Alberta

Any opioid

Rate of drug poisoning deaths per 100,000 person years by month

Rate per 100,000 person years

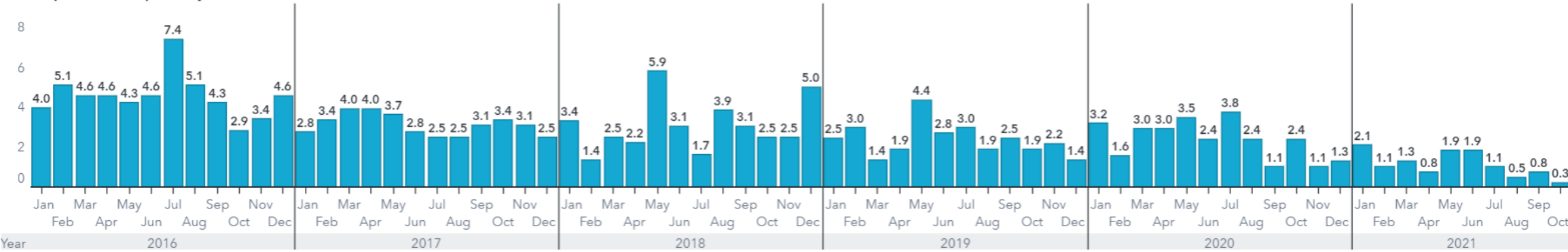


Alberta

Pharmaceutical opioids

Rate of drug poisoning deaths per 100,000 person years by month

Rate per 100,000 person years



Alberta's Prescription Monitoring Programs



- The goal is to identify patients at risk of adverse outcomes and their associated prescribers, and to intervene to improve safe/appropriate practice and reduce patient risk.
- Since 2008, TPP Alberta and the CPSA have continuously enhanced their capacity to use data to support safe prescribing and professional regulation.

Guideline-Based Risk Assessment

- 2019 Centers for Medicare & Medicaid Services (CMS) uses these safety measures to identify high risk opioid use from administrative health data sets:
 - >120 OME/day for >90 days, AND
 - 4 or more opioid prescribers and pharmacies, AND
 - Concurrent use of opioids and benzodiazepines for >30 day
- 2017 Canadian opioid prescribing guidelines for non-cancer pain:
 - Avoid use if history of substance use disorder or mental illness
 - >90 OME/day is high risk

Outcome-Specific Risk Prediction

- Framingham Heart Study (1976) – 8491 participants between 30 and 74 years of age, from the town of Framingham MA, who were free of cardiovascular disease (CVD) at the time of entering the study, were followed for development of a first CVD event.
- Framingham Risk Score estimates the 10 year risk of CVD events (e.g., heart attack, stroke, peripheral vascular disease, heart failure). Patients at higher risk warrant more aggressive preventative therapy.
- The original model achieved good discrimination (C statistic 0.763 men and 0.793 women) and calibration.
- **Widely adopted around the world**

The Absolute CVD Risk/Benefit Calculator

Framingham US Data, 10 Year Risk

Heart attacks + angina/coronary insufficiency + heart failure + strokes + intermittent claudication

QRISK[®]2-2014 UK Data, 10 Year Risk

Heart attacks + strokes

ACC/AHA ASCVD US Data, 10 Year Risk

CHD death + nonfatal heart attacks + fatal/nonfatal strokes

PREDICT New Zealand Data, 5 Year Risk

Heart attacks + angina + heart failure + strokes/TIAs + peripheral vascular disease

Age

51 years

Gender

Male Female

Smoker

Yes No

CVD risk is reversed after 5-10 years of no smoking

Diabetes

Yes No

Systolic Blood Pressure

120 mmHg

Enter present blood pressure regardless of treatment

120 mmHg is used for baseline risk

On treatment for BP

Yes No

Click YES if taking blood pressure medication

Only applies if SBP is greater than 120 mmHg

Total Cholesterol

3 mmol/L

Cholesterol should be prior to drug treatment

3 mmol/L is used for baseline risk.

Relative Benefit: 0%

Benefit often has *nothing* to do with the effect on the surrogate marker. At present, you can only select one intervention at a time.

Physical Activity

Mediterranean Diet vs Low fat

Vitamin/Omega-3 supplements

BP meds (not atenolol/doxazosin)

Low-mod intensity statins

High intensity statins

Fibrates

Niacin

Ezetimibe

Metformin

Sulfonylureas

Insulins

Glitazones

GLPs

DPP-4s

Meglitinides

SGLT2

Smoking Cessation

ASA

[Benefit Estimate Details](#)

Risk Time Period

10 years



94.7%

No event



5.3%

Total with an event



0.0%

Number who benefit from treatment

NNT

∞

Number needed to treat

<https://cvdcalculator.com/>

Some Issues With Prediction Models

- Study population is not similar to yours (middle class, urban, Caucasians)
- Omit important predictive factors (e.g., physical activity, family history, SES)
- Does not take into account risk or treatment over time (e.g., BP medications)
- Crude and imprecise thresholds (few risk categories for entire population)
- The risk applies to the group, and does not provide a specific individual's risk
- Expensive to update or recreate for your specific population
- **Population paradox – a preventive measure brings large benefits to the community but offers little to each participating individual.**

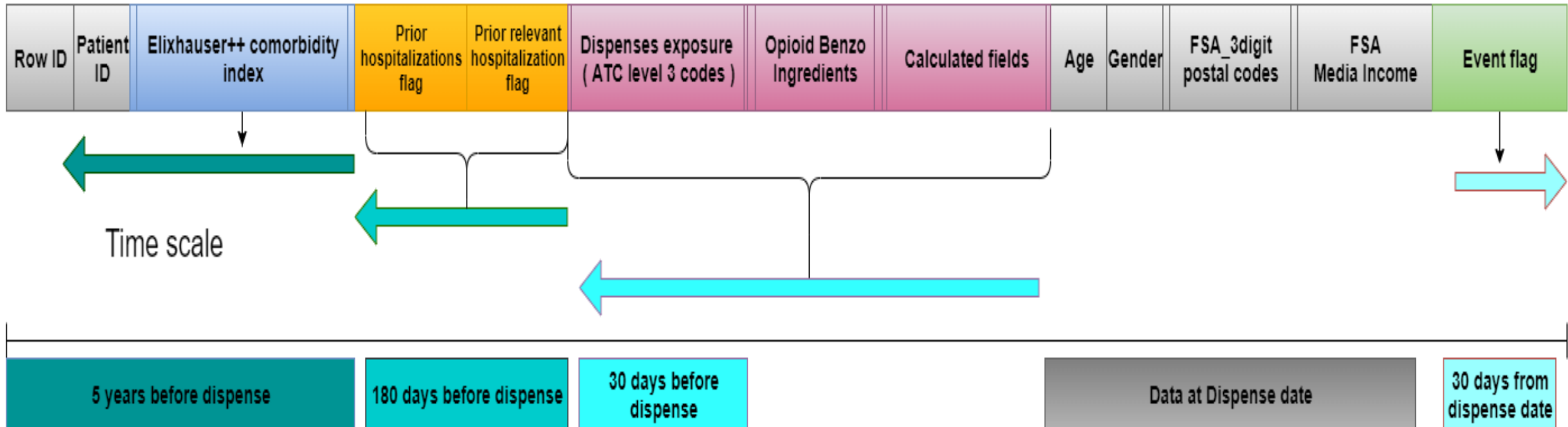
“Prophecy is a good line of business but it is full of risks.” – Mark Twain

Project Objective

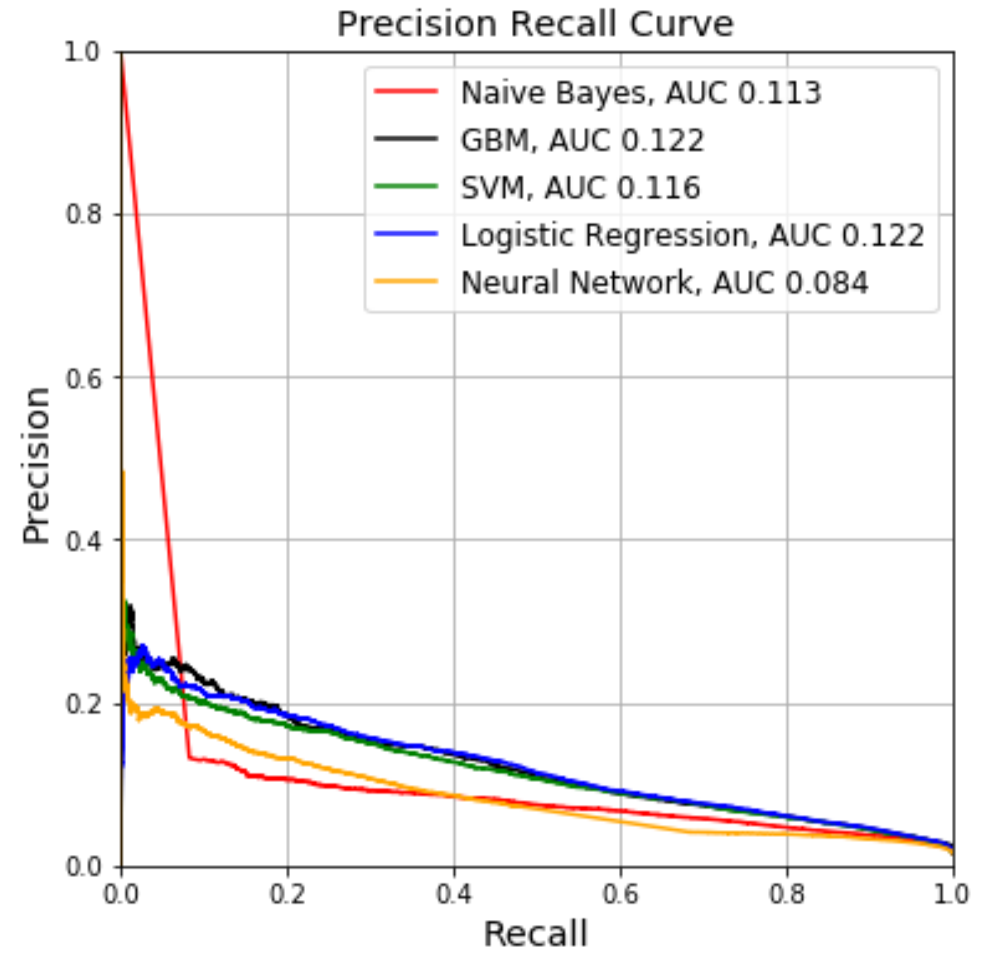
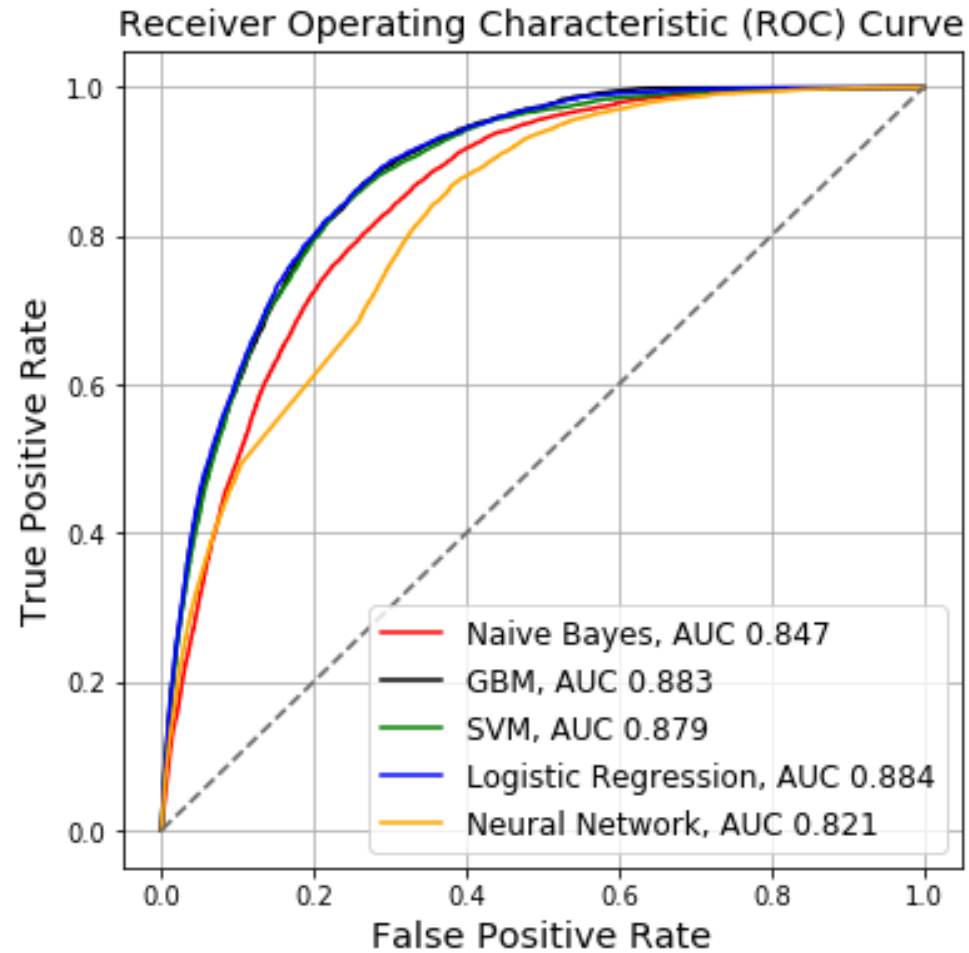
- Apply ML methods to AB's healthcare administrative databases to develop a population-based model to predict 30-day risk of drug poisoning related emergency visit, hospitalization, or death at the time of an opioid dispense.
- Explore potential predictive factors ('features') from MD claims, Rx dispensations, hospitalizations, emergency department visits and death data.

Database	Records
Patients who received at least 1 opioid 2012-2018	1,794,606
Pharmaceutical Information Network (PIN)	255,446,107
Practitioner Claims (Claims)	164,221,577
National Ambulatory Care Reporting System (NACRS)	34,360,160
Discharge Abstract Database (DAD)	1,657,230
Vital statistics (VS)	93,098

Predictive Features in Relation to Outcomes



ROC vs PR curves



2018 Model Performance Using Centers for Medicare & Medicaid Services (CMS) Guidelines

Condition	Sensitivity	Specificity	AUC
High dose and Multiple doctors and multiple pharmacies and concurrency	0.001	0.999	0.500
High dose or Multiple doctors and multiple pharmacies or concurrency	0.472	0.681	0.583
High dose or Multiple doctor or multiple pharmacies or concurrency	0.607	0.636	0.622
>120 OME/day for >90 days - Highdosage	0.057	0.950	0.504
(>4 doctors) and (>4 pharmacies)	0.085	0.972	0.529
Concurrency (Opioid with BDZ over 30 days)	0.407	0.733	0.570
Multiple pharmacies	0.120	0.956	0.538
Multiple doctors	0.291	0.889	0.590
(>4 doctors) or (>4 pharmacies)	0.326	0.872	0.599

2018 Model Performance using Canadian Practice Guidelines

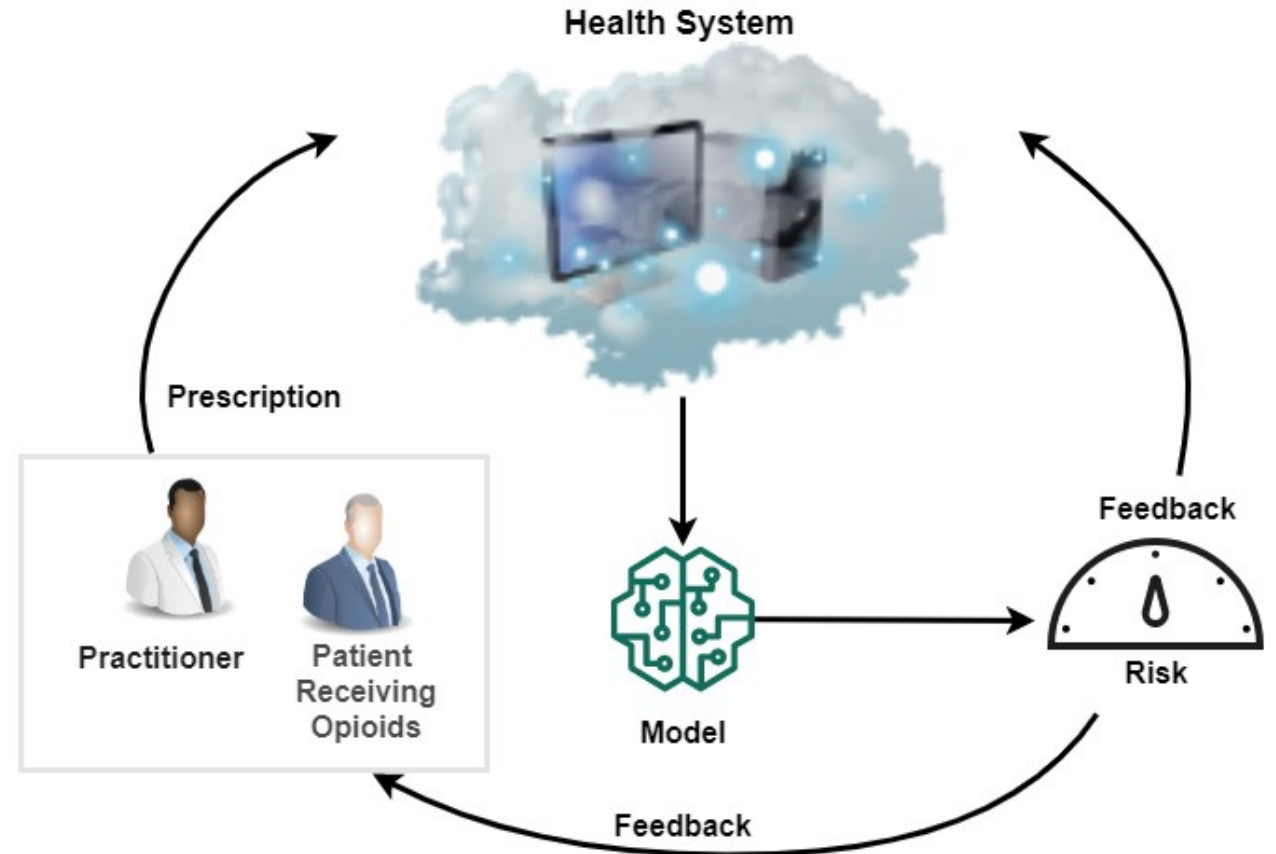
Guideline	Sensitivity	Specificity	AUC
Hx of mental disorder only	0.92	0.32	0.626
Substance abuse only	0.99	0.38	0.688
OME/day >90 only	0.14	0.88	0.517
(Mental disorder and substance abuse) OR OME/day >90	0.92	0.48	0.702
Mental disorder and substance abuse AND OME/day >90	0.14	0.93	0.534
Mental disorder OR substance abuse OR OME/day >90	0.99	0.18	0.589

The Results

- "Machine learning classifiers, especially incorporating hospitalization and physician claims data, have better predictive performance compared with guideline or prescription history only approaches when predicting 30-day risk of adverse outcomes. Prescription monitoring programs and health departments with access to administrative data can use machine learning classifiers to effectively identify those at higher risk compared with current guideline-based approaches."
- Sharma V, Kulkarni V, Eurich DT, Kumar L, Samanani S. Safe opioid prescribing: a prognostic machine learning approach to predicting 30-day risk after an opioid dispensation in Alberta, Canada. *BMJ open*. 2021 May 1;11(5):e043964.

How would we apply this model in practice?

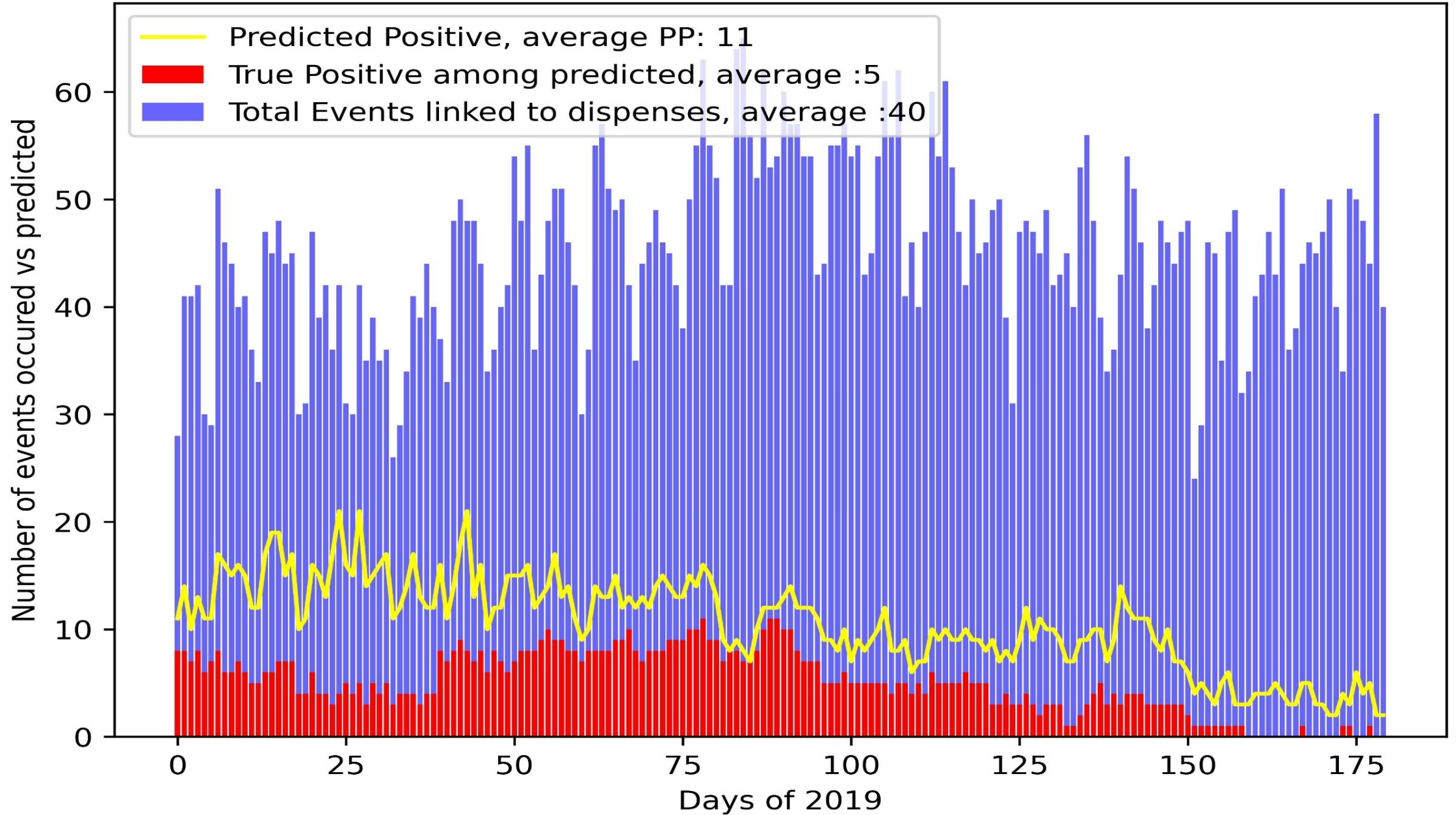
- As part of practice feedback?
- As part of population-based high risk case management?
- As part of clinical decision support?



Daily identification of patients with new starts based on high-risk threshold

Threshold	PP	TP	FN	FP	TN	PPV	NPV	Sensitivity	Specificity	LR+
0	2098.81	39.93	0	2058.88	0	0.02		1	0	1
0.1	1508.97	39.37	0.56	1469.60	589.28	0.03	1.00	0.99	0.28	1.38
0.2	1015.76	38.30	1.63	977.46	1081.43	0.04	1.00	0.96	0.51	2.00
0.3	761.54	37.09	2.83	724.45	1334.43	0.05	1.00	0.93	0.63	2.60
0.4	604.34	35.45	4.48	568.89	1489.99	0.06	1.00	0.89	0.70	3.16
0.5	460.43	32.00	7.93	428.43	1630.46	0.07	0.99	0.81	0.78	3.82
0.6	300.12	26.64	13.29	273.48	1785.41	0.09	0.99	0.68	0.86	5.10
0.7	154.31	19.84	20.08	134.46	1924.42	0.14	0.99	0.51	0.93	7.74
0.8	56.89	11.94	27.99	44.95	2013.93	0.22	0.99	0.31	0.98	14.28
0.85	28.56	8.34	31.58	20.22	2038.67	0.29	0.98	0.22	0.99	22.69
0.9	10.60	5.05	34.88	5.55	2053.33	0.44	0.98	0.13	1.00	
0.95	1.88	1.28	38.64	0.60	2058.28	0.70	0.98	0.03	1.00	
1	0	0	39.93	0	2058.88		0.98	0	1	

0.9 predicted probability threshold



Explaining Individual Predictions

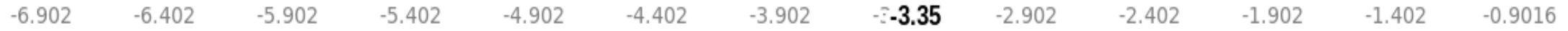
higher ⇌ lower
model output value



DrugAbuse = 0 Alcohol = 0 PriorHosp = 0 Poison = 0 Depression = 0 Injury = 0 Fluiddo = 0 Psychoses = 0

higher ⇌ lower

base value model output value



Gender_M = 1 Liver = 1 Fluiddo = 1 Poison = 1 Alcohol = 1 DrugAbuse = 0 PriorHosp = 0 Psychoses = 0

higher ⇌ lower

base value model output value



RCPT_AGE = 28 PriorHosp = 1 DrugAbuse = 1 Alcohol = 1 Poison = 1 Fluiddo = 0 Psychoses = 0 Depression = 0

higher ⇌ lower

base value model output value



macy_Risk_30 = 2 DrugAbuse = 1 Fluiddo = 1 PriorHosp = 1 PriorHospRelevant = 1 RCPT_AGE = 24 FSA_T3H = 1 Alcohol = 1 Poison = 1



Questions and Discussion



Health Analytics Community, Canada Health Infoway

Dr. Salim Samanani, CEO and Medical Director

salim@okaki.com

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