Table 1 Raw Variable Description

Column	Description
X_ id	Unique row identifier for open
	Data database
EVENT_UNIQUE_ID	Offence Number
REPORT_DATE	Date Offence was Reported.
OCC_DATE	Date of Offence
•	•
•	•
	•
	•
•	•

Table 2: Sample of Clean Data: Part 1

X_id	DCCDA	47E	DCC_	YEAR	OCC_/	MOUNTH	OCC-D	Ar	occ. [Pow
1	2014-0		2014		_	19-4	1		Wedne	sday
3	2014-01	1-01	2014		Janu	10-4	1		Nedne	
15	2014-0	01-01	2014		Jan	uary	1		Nedne	esday
•	•		•		•		•			
•			•						/	
•	•						•		•	

Table 3: Sample of Clean Data: Part 2

OCC_HOUR	DIVISION	PREMISES_TYPE	UCR, _CODE
8	D51	House	2142
11	542	House	2142
0	DIY	outside	2142
•	•	•	•
•	•	•	•
	•		•

Table 4: Sample of Clean Data: Part 3

OFFENCE	LONG_WGS84	LAT_WGS84
Thett From Motor Vehicle under	-79.37453	43.65707
Thet & From Motor Vehicle unde	-79. 27716	43.81731
That I from Motor Vehicle unde	-79.40111	43. 65 227
	•	

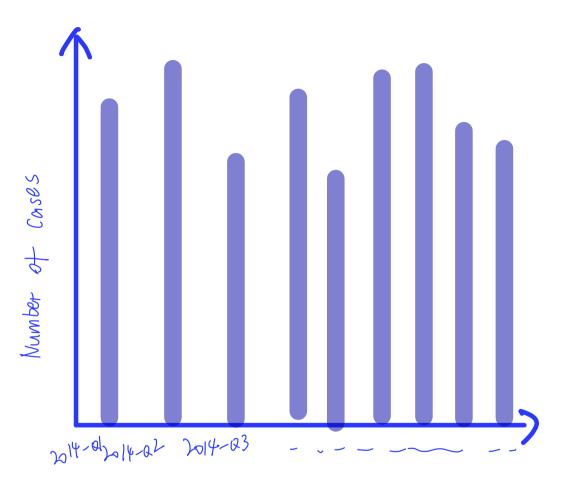


Figure 1: Histogram of quarterly count of Car thete in Toronto

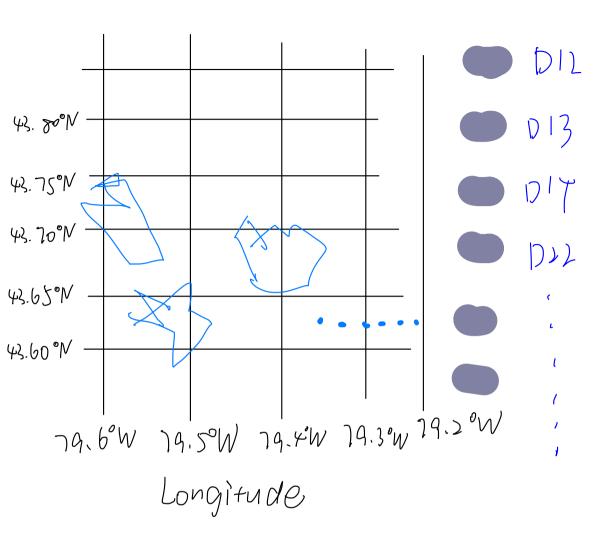


Figure 2: Spatial view of Ve hicle theft cases by division for September 2024

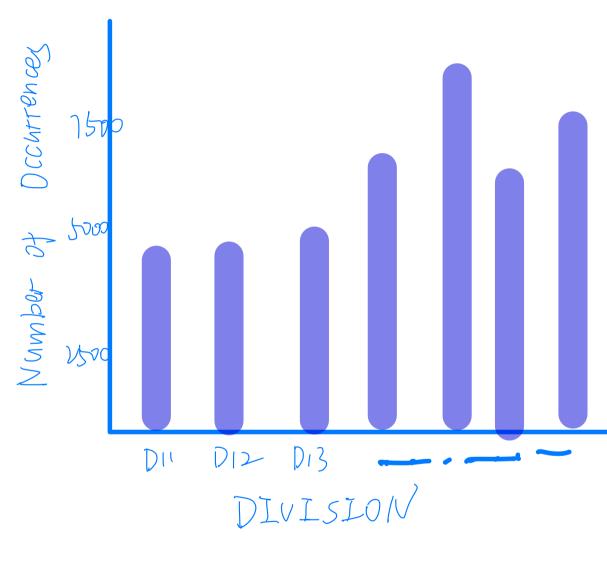


Figure 3: Barchart of count of occurrences by divisions

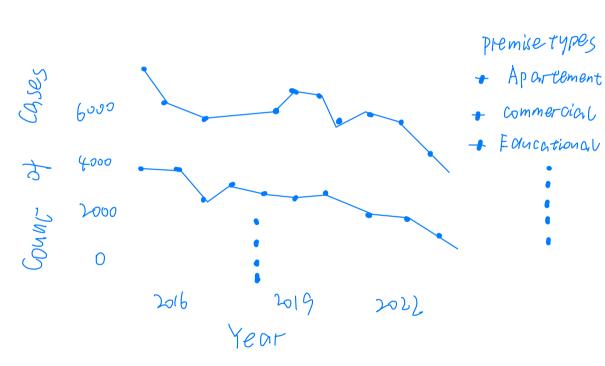


Figure 4: Linechart of count of occurrences by premise types and year

+ Thett From Motor Vehicle Over Thett From Motor vehicle Under 7500 2016 2019 7022 Year

ottence Type

Figure 5: Linechart of count of occurrences by offence types and year

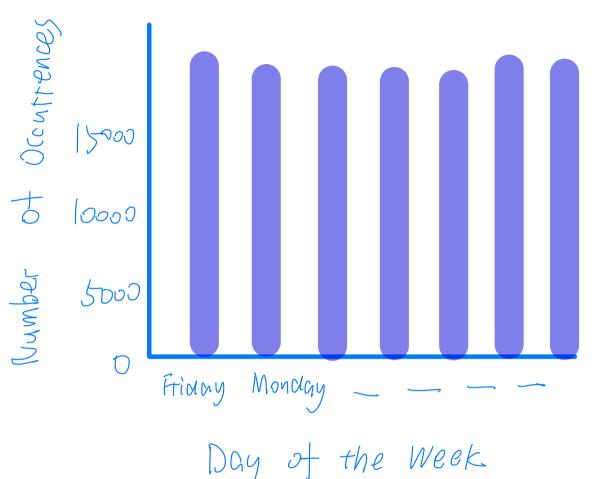


Figure 6. Barchart of count of occurrences by day of the week

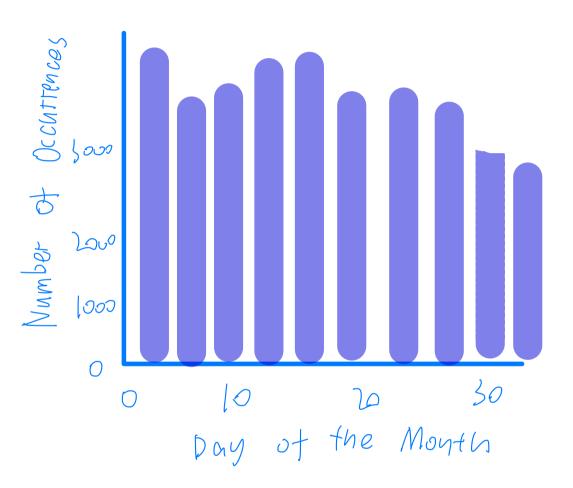


Figure 7: Barchart of count of occurrences by day of the Month

Table 5: Poisson teglession results

	Estimate	Std. Error	Zvalue	Pr (7/2/)
(Intercept)	6.4422634	0.0568984	113.231977	0.000000
time action			•	•

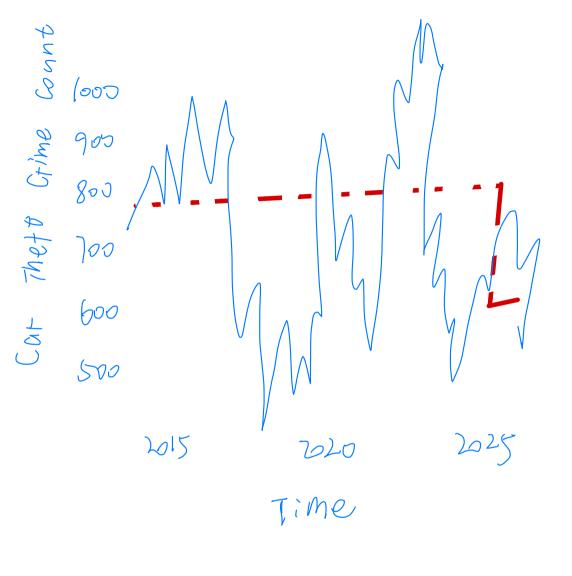


Figure 8: Poisson regression prediction

Table 6: Poisson regression

dispersion tactor

Stats Value

Dispersion Factor 14.73955

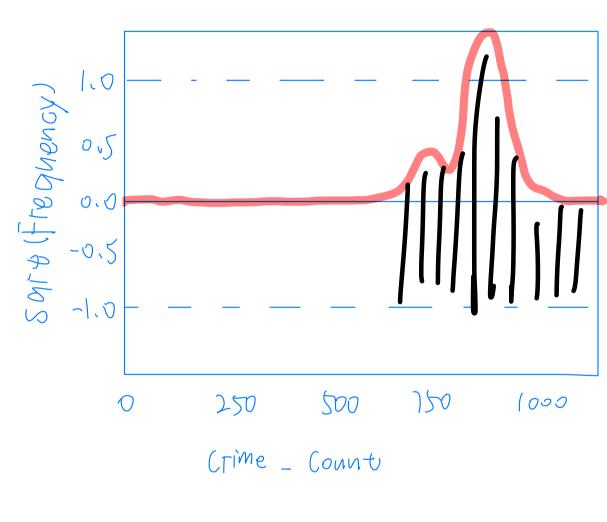


Figure 9: Poisson regression nootogram

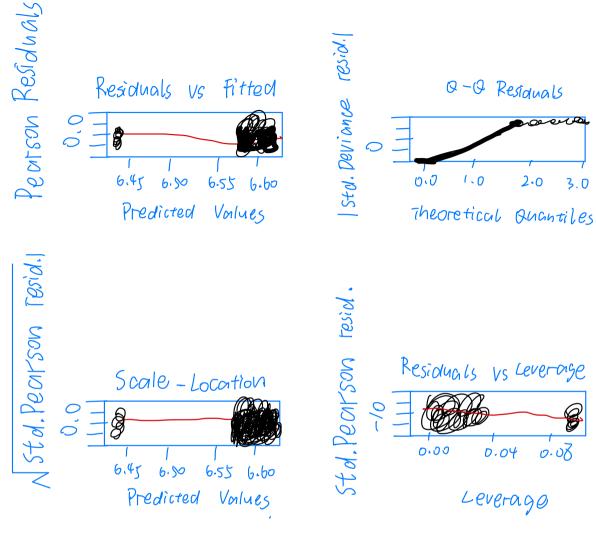


Figure 10: Poisson regression models diagnosis

(able 7: (Quasi-likelihood	Poisson
re gression	results	

	Estimate	Std. Error	tvalue	Pr() t
Intercept)	6.4422634	0.2184296	29.4935438	0.0000

17 time

action

Table 8: Comparison between Quasi-likelihood Poisson and Poisson estimated contidence intervals

	2.5%	97.5%	model -type
[Intercept]	6.3707293	6.5537576	Poisson Mode C
time	0.0000032	0.0007	Poisson Mode (
action	•	•	ſ
(Intercept)	(•	1
time 1		•	•
action 1	•	•	•

Table 9: Quasi-likelihood Poisson regression with avariate of premise types

	Estimate	Std Firor	tvalue	Pr(> t)
(Intercepe)	3.8692034	0.1853514	20.8749597	0,000000
time	0.00000 89	g-000010)	0.8824230	0-3777979
PREMISES_TYP.	E Commercial.	•	1	
PREMISES_TYP	EE au cational •			
PREMISES_TYP	EHouse	•		
PREMISEJ_TYPE	E Other	•	1	
PREMISEJ_TYPE	E Outside .	•	•	
PREMISES_TYPE Action	Transit	1	1	ı

	2,20,161	_		
Intecept)	3.0982947	०.।७४४५११	17.67848	b 0.00000
	Estimate	Std Firor	tralue	Pr() (+ ()
With	covariate o	t Offence	types	
Table 10	: Quasi-lik	elihood Po	t Noziic	regression

0.0000095 tlme OFFENCEThete From Motor vehicle Under

action

Table 11: Quasi-likelihood Poisson regression with covariate of divisions

	Estimate	Std Fnor	tvalue	Pr() [t])
Intecept	2.8242025	0.1359010	20.7808661	6,0000000
time	०. ०००००१५९	0.000073	3.5471310	0.0003832
DIVISIONDI	•	•	•	ſ
DIVISIOND	13	,	(1
DIVISIOND		(•	1
DIVISIOND	27	,	1	
•	•	•	1	
			•	•
•				
(