

# PRÉDICTION DES ÉMISSIONS DE CO2 DES BÂTIMENTS

Manu, David

#### **CO2 DES BÂTIMENTS**

Le projet

**Preprocessing** 

**Analyses** 



**Prospection de modèles** 

**Application** 

**Conclusions** 

Recommandations Retour d'expérience

# Le projet

#### Distribution des tâches

Installation Azure (David)

Imputations (Manu)

DevOps (David)

Analyses (Manu)

Tests (David)

Conception de models

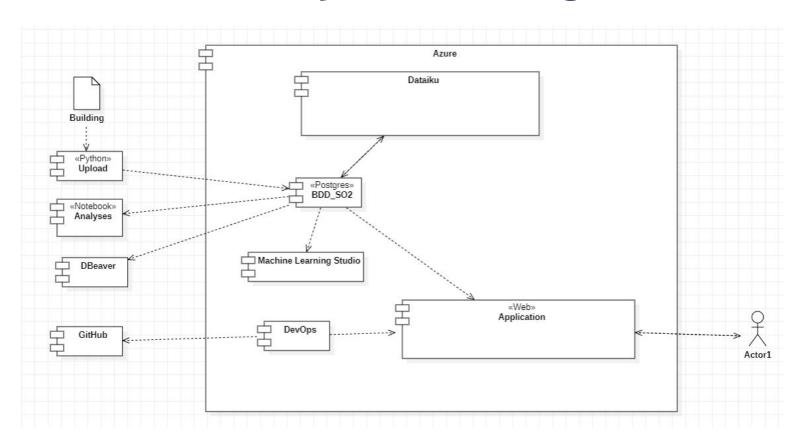
(Manu, David)

Chargement base (David)

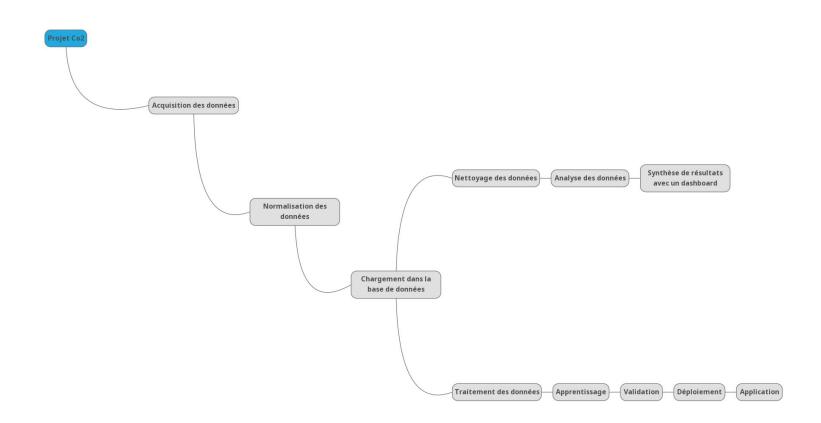
Présentation (Manu, David)

Flow Dataiku (Manu)

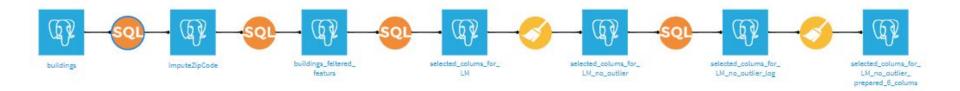
## Installation, DevOps, Tests, Chargement Base.



#### **Schema fonctionnel**



## Flow principal Dataiku



# Preprocessing

#### **Suppression de colonnes**

SUMMARY			Top 1 out of 1 values in sample	Count	%	Cum. %
Valid •	3,376	100.0 %	2016	3376	100.0	100.0
Нарах 🐧	0	0.0 %				
Invalid •	0	0.0 %				
Empty •	0	0.0 %				
0 HAPAXES		0.0 %				
0 INVALIDS		0.0 %				
SUMMARY			Top 1 out of 1 values in sample	Count	%	Cum. %
Valid •	3,376	100.0 %	Seattle	3376	100.0	100.0
Нарах 📵	0	0.0 %		00-09937002		
Invalid •	0	0.0 %				
-	0					
Empty =	0	0.0 %				
Empty •  0 HAPAXES	0	0.0 %				

#### **Suppression de colonnes**

SUMMARY			Top 50 out of 3362 values in sample	Count	%	Cum. %
Valid •	3,376	100.0 %	Northgate Plaza	3	0.1	0.1
Нарах 🕕	3,349	99.2 %				
Invalid •	0	0.0 %	Airport Way	2	0.1	0.1
Empty •	0	0.0 %	Bayview Building	2	0.1	0.2
<ul> <li>#4706 Bitt</li> </ul>		99.2 %	Canal Building	2	0.1	0.3
<ul> <li>#8944 Wes</li> <li>(71367A) S</li> </ul>	st Seattle SEATTLE Ma	acy's	Central Park	2	0.1	0.3
• (71371A) f	NORTHGAT	E Macy's	Crestview Apartments	2	0.1	0.4
0 INVALIDS		0.0 %	Fairview	2	0.1	0.4

Idem pour adresse

#### **Suppression de colonnes**

Comments	Text	100.00%	<b>*</b>
☐ YearsENERGYSTARCertified	Integer	96.48%	0
☐ ThirdLargestPropertyUseType	Text	82.35%	
☐ ThirdLargestPropertyUseType	Decimal	82.35%	0
☐ SecondLargestPropertyUseType	Text	50.27%	0
☐ SecondLargestPropertyUseTyp	Decimal	50.27%	0

#### **Suppression outliers**

Outlier Text 99.05%

SUMMARY		
Valid •	3,376	100.0 %
Нарах 🐧	0	0.0 %
Invalid •	0	0.0 %
Empty =	3,344	99.1 %
0 HAPAXES		0.0 %
0 INVALIDS		0.0 %

Top 3 out of 3 values in sample	Count	%	Cum. %
No value	3344	99.1	99.1
Low outlier	23	0.7	99.7
High outlier	9	0.3	100.0

Suppression des features quantitatives liées à la target

☐ SiteEUIWN_kBtu_sf_	Decimal	99.82%	3-
☐ SourceEUI_kBtu_sf_	Decimal	99.73%	j.
☐ SourceEUIWN_kBtu_sf_	Decimal	99.73%	Ž.
☐ SiteEnergyUse_kBtu_	Decimal	99.85%	ž.
☐ SiteEnergyUseWN_kBtu_	Decimal	99.82%	ž.
☐ SteamUse_kBtu_	Decimal	99.73%	ħ.
☐ Electricity_kBtu_	Decimal	99.73%	Å
☐ NaturalGas_kBtu_	Decimal	99.73%	No.

**Imputation NumberofBuildings** 

```
D. "Neignbornood",

CASE WHEN "NumberofBuildings" IS NULL THEN 1 else "NumberofBuildings" END AS "NumberofBuildings",

b. "NumberofEloors"
```

SUMMARY			SUMMARY		
Valid •	3,376	100.0 %	Valid •	3,367	100.0 %
Нарах 🐧	6	0.2 %	Hapax 🛈	6	0.2 %
Invalid •	0	0.0 %	Invalid •	0	0.0 %
Empty •	8	0.2 %	Empty •	0	0.0 %

Nb\_building = f(Nb\_Floors)

	NumberofFloors	*	123 round	*
1		4		1
2		3		1
3		2		1

#### Imputation LargestPropertyUseType

D. PropertyGFABUTIOTING\_S\_ ,

CASE WHEN "LargestPropertyUseTypeGFA" IS NULL THEN "PropertyGFABuilding\_s\_" else "LargestPropertyUseTypeGFA" end as "LargestPropertyUseTypeGFA",

b. "SecondLargestPropertyUseTypeGFA"



#### **Transformation de features**

#### **Utilise un type d'énergie**

SteamUse_kBtu_	Decimal	☐ Have_Stream_Energy	Boolean
☐ Electricity_kBtu_	Decimal	☐ Have_Electricity_Energy	Boolean
☐ NaturalGas_kBtu_	Decimal	☐ Have_NaturalGas_Energy	Boolean

```
b."SteamUse_kBtu_" > 0.0 as "Have_Stream_Energy",
b."Electricity_kBtu_" > 0.0 as "Have_Electricity_Energy",
b."NaturalGas_kBtu_" > 0.0 as "Have_NaturalGas_Energy",
```



SUMMARY		
Valid •	3,314	100.0 %
Нарах 🕕	0	0.0 96
Invalid •	0	0.0 96
Empty =	0	0.0 98

Top 2 out of 2 values in sample	Count	96	Cum. %
true	2090	63.1	63.1
false	1224	36.9	100.0

#### **Transformation de features**

#### Log de features

```
log ("LargestPropertyUseTypeGFA") as "LargestPropertyUseTypeGFA_log",
log ("TotalGHGEmissions") as "TotalGHGEmissions_log",
log ("SiteEnergyUse_kBtu_") as "SiteEnergyUse_kBtu_log"
```



#### Sélection de features

☐ YearBuilt	Integer	100.00%
☐ BuildingType	Text	100.00%
☐ Neighborhood	Text	100.00%
☐ Have_Stream_Energy	Boolean	100.00%
☐ Have_Electricity_Energy	Boolean	100.00%
☐ Have_NaturalGas_Energy	Boolean	100.00%
☐ PrimaryPropertyType	Text	100.00%
NumberofBuildings	Integer	100.00%
☐ LargestPropertyUseTypeGFA	Decimal	100.00%
☐ TotalGHGEmissions	Decimal	100.00%
☐ SiteEnergyUse_kBtu_	Decimal	100.00%
☐ LargestPropertyUseTypeGF	Decimal	100.00%
☐ TotalGHGEmissions_log	Decimal	100.00%
☐ SiteEnergyUse_kBtu_log	Decimal	100.00%

On conserve les targets, TotalGHGEmission et SiteEnergyUse.

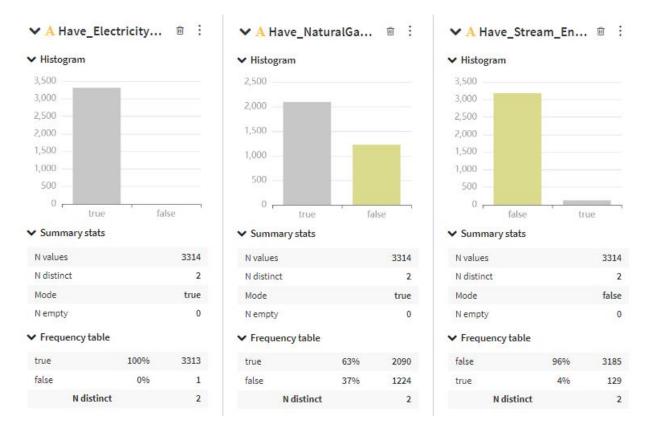
On cherche à prédire les deux targets en fonction du projet.

#### Un projet intègre:

- Des types d'énergie.
- Des usages (résidentiels).
- Une zone géographique (quartier).
- Une surface.
- Un nombre de bâtiments.

La date de construction de bâtiments est un élément temporel.

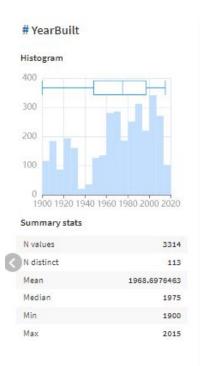
## Analyses data énergie

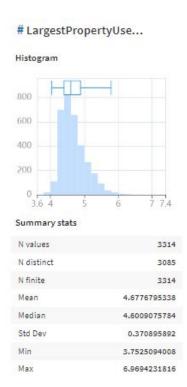


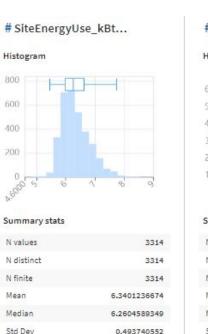
## **Analyses data numériques**

Min

Max

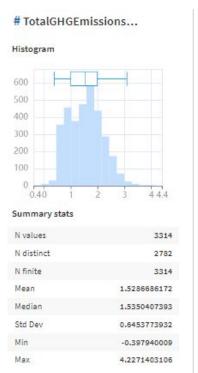




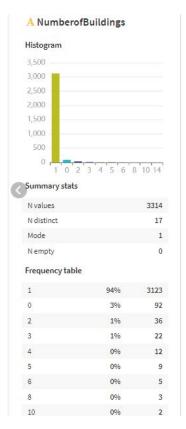


4,7568885434

8.9414735231

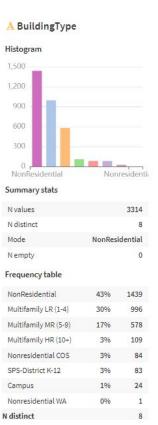


### Analyses data catégorielles

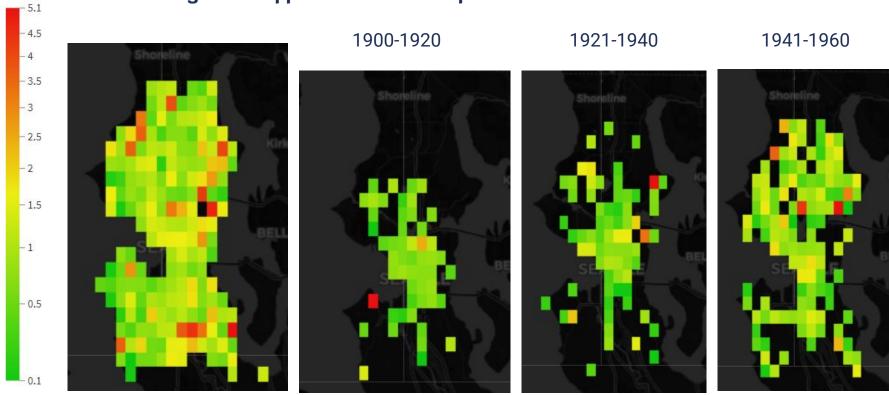




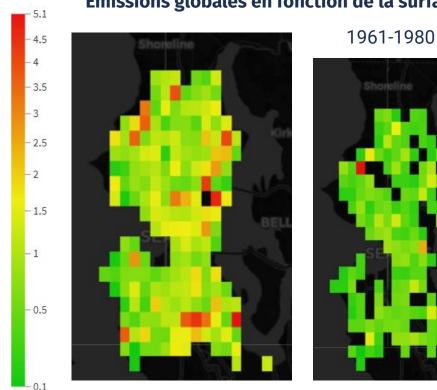




#### Émissions globales apportées à la surface par tranche d'années de construction

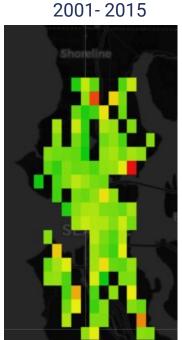


#### Émissions globales en fonction de la surface par tranche d'années de construction



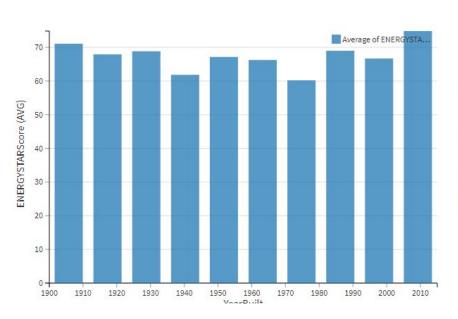


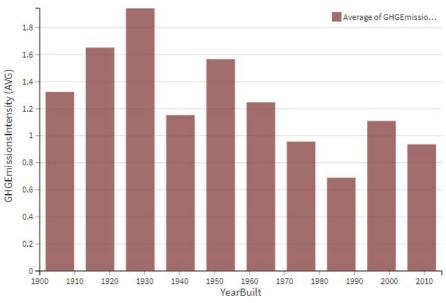




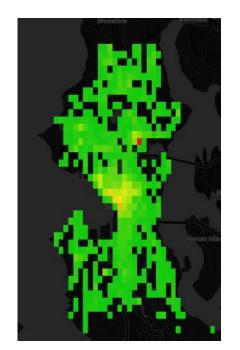
**Analyses** 

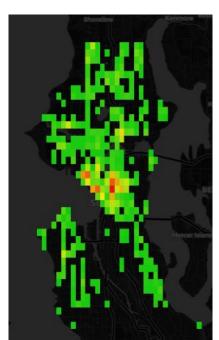
#### Émissions globales à comparer avec les évaluation de EnergyStarScore

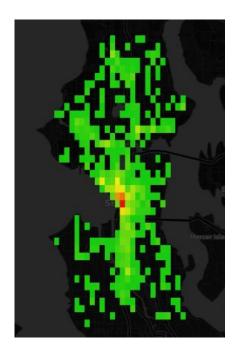


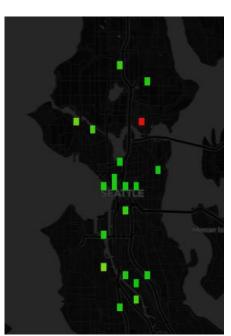


# **Analyses**Répartition des bâtiments









Tout

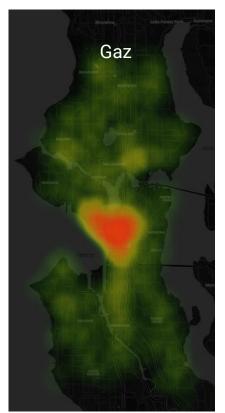
Multifamily

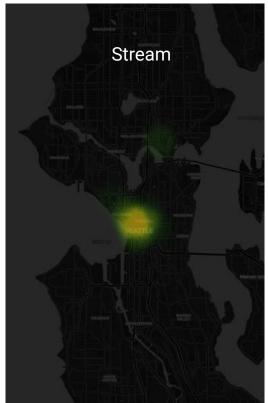
Non Residentiel

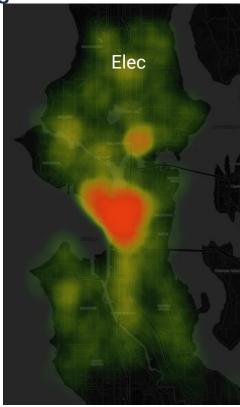
Campus

Analyses

Densité d'utilisation des énergies

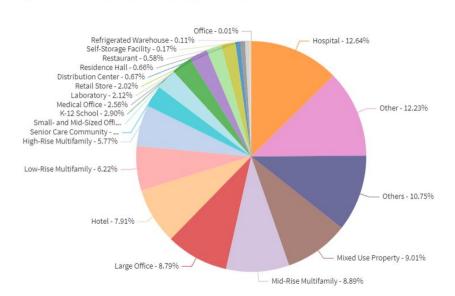




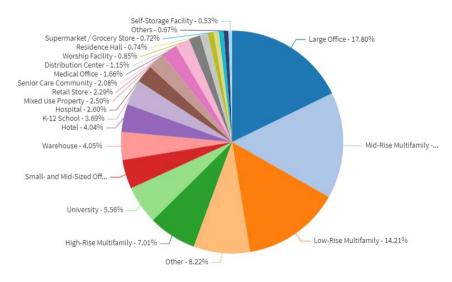


#### Répartition énergétique et surface au sol par usages

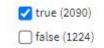
#### Sum of TotalGHGEmissions by PrimaryPropertyType



#### Sum of LargestPropertyUseTypeGFA by PrimaryPropertyType



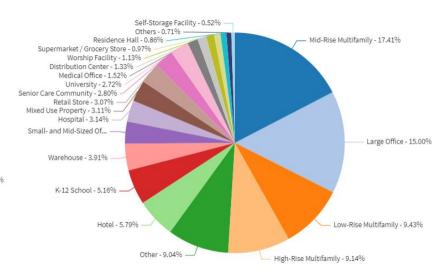
#### Répartition énergétique et surface au sol par usages en fonction de l'utilisation de gaz



Sum of TotalGHGEmissions by PrimaryPropertyType

Restaurant - 0.67% Distribution Center - 0.75% Hospital - 12.83% Worship Facility - 0.86% Others - 0.97% Laboratory - 1.71% Warehouse - 2,10% Medical Office - 2.17% Retail Store - 2.30% Supermarket / Grocery Stor... -Other - 12.24% Small- and Mid-Sized Offi... K-12 School - 3.35% University - 3.41% Senior Care Community -... ---Low-Rise Multifamily - 6.06% -Mixed Use Property - 10.09% High-Rise Multifamily - 6.26% Mid-Rise Multifamily - 9.80% Large Office - 6.54% - Hotel - 9.18%

Sum of LargestPropertyUseTypeGFA by PrimaryPropertyType

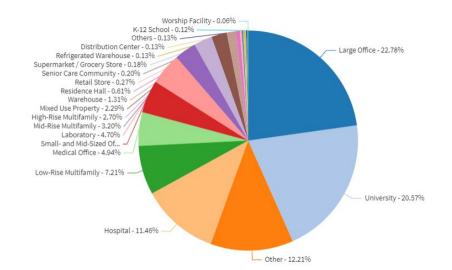


#### Répartition énergétique et surface au sol par usages en fonction de l'utilisation de gaz

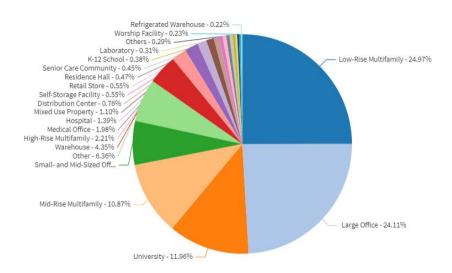


✓ false (1224)

Sum of TotalGHGEmissions by PrimaryPropertyType

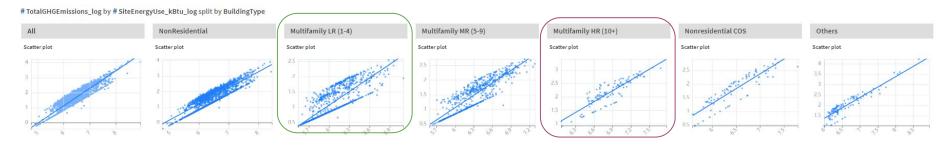


Sum of LargestPropertyUseTypeGFA by PrimaryPropertyType

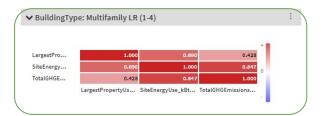


Le score d'évaluation est inadapté à informer sur le degrés d'émissions le fait d'avoir ou non du gaz semble plus fiable à informer sur le degré d'émission



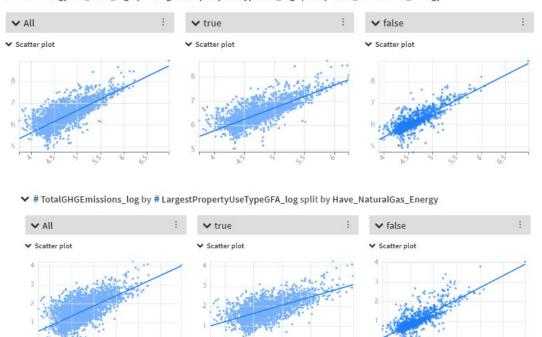


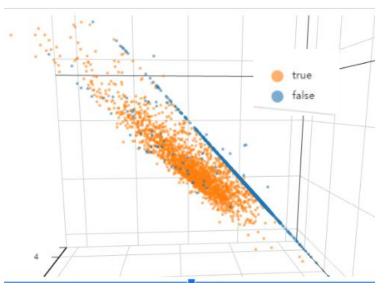




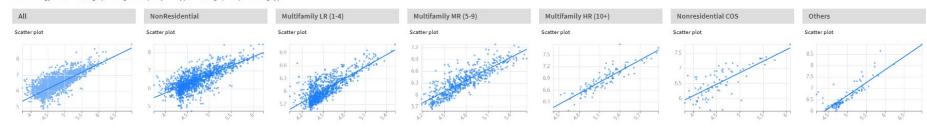
LargestPro	1.000	0.907	0.704	
SiteEnergy	0.907	1.000	0.854	
TotalGHGE	0.704	0.854	1.000	
Lar	gestPropertyUs SiteEne	revilse kRt TotalGH	GEmissions	

▼ # SiteEnergyUse\_kBtu\_log by # LargestPropertyUseTypeGFA\_log split by Have\_NaturalGas\_Energy

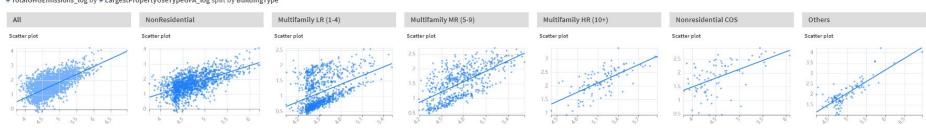




#### #SiteEnergyUse\_kBtu\_log by #LargestPropertyUseTypeGFA\_log split by BuildingType



#### # TotalGHGEmissions\_log by # LargestPropertyUseTypeGFA\_log split by BuildingType



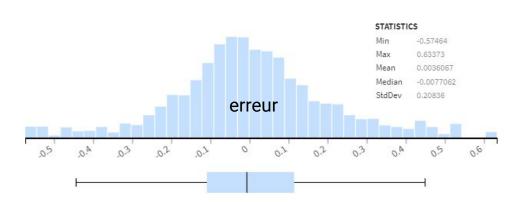
# Prospection de modèles

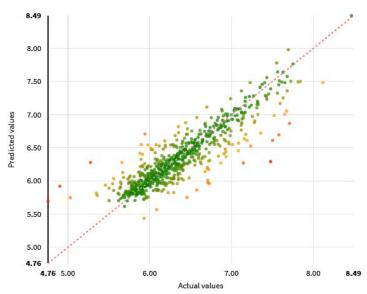
## Prospection de modèles

Avec Dataiku (SiteEnergyUse)

#### **R2 Score**

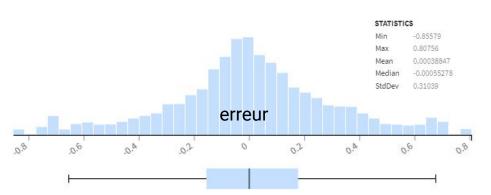
☐ ● Rand	om forest (s19)		0.748 (± 0.027)	☆
☐ • Ridge	e (L2) regression (s19)		0.780 (± 0.024)	☆
SVM	(s19)	7	0.785 (± 0.024)	☆
☐ ● Singl	e Layer Perceptron (s19)		0.740 (± 0.168)	₩.



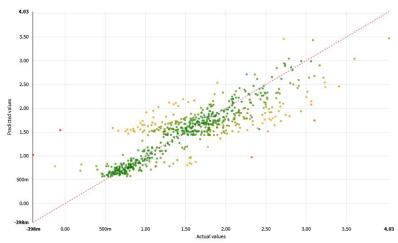


**Avec Dataiku (TotalGHGEmission)** 

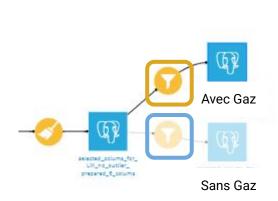
# □ Random forest (s22) 0.724 (± 0.025) ☆ □ Ridge (L2) regression (s22) 0.721 (± 0.017) ☆ □ SVM (s22) 0.735 (± 0.025) ☆ ✓ Single Layer Perceptron (s22) ♀ 0.736 (± 0.022) ☆

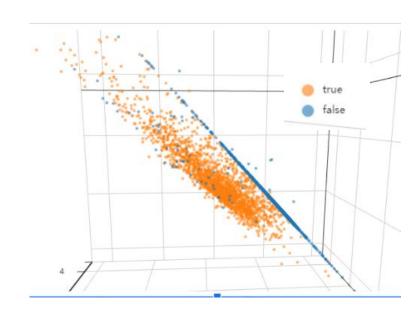


R2 Score

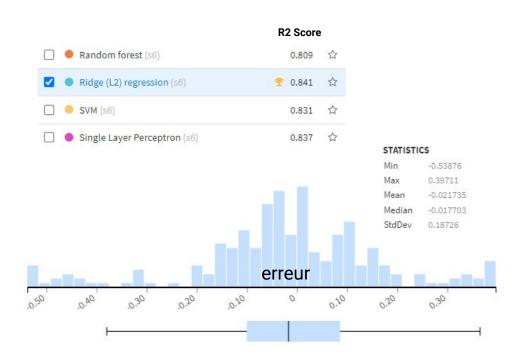


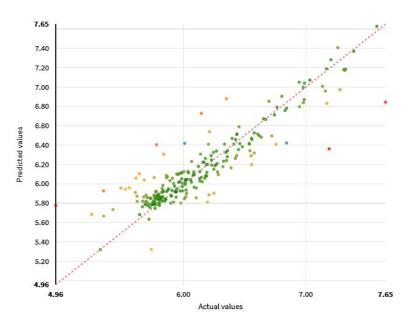
Avec Dataiku (TotalGHGEmission si avec GazEnergy ou sans GazEnergy)



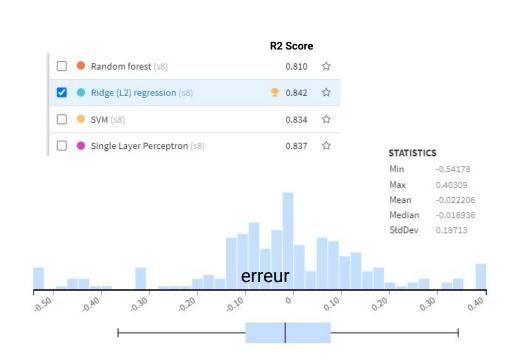


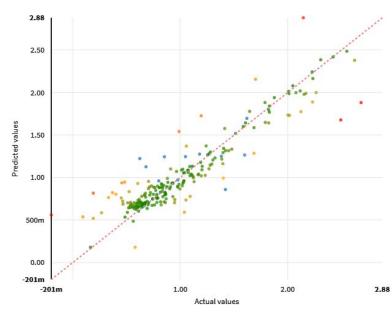
Avec Dataiku (TotalGHGEmission sans GazEnergy)





Avec Dataiku (TotalGHGEmission avec GazEnergy)



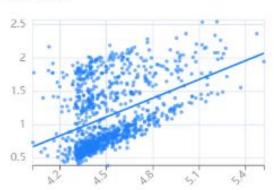


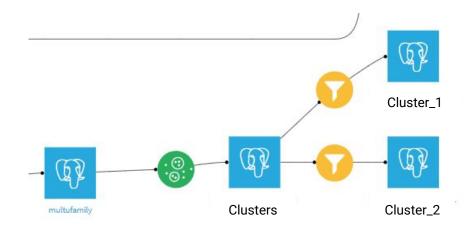
# Clustering

#### Multifamily (1-4) issue de BuildingType

### Multifamily LR (1-4)

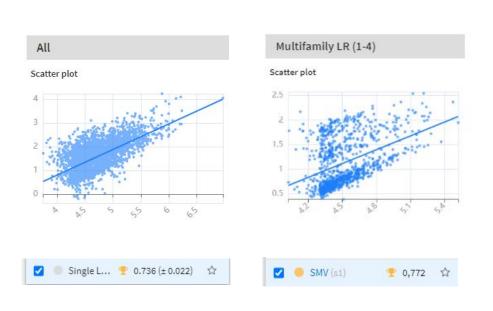
#### Scatter plot

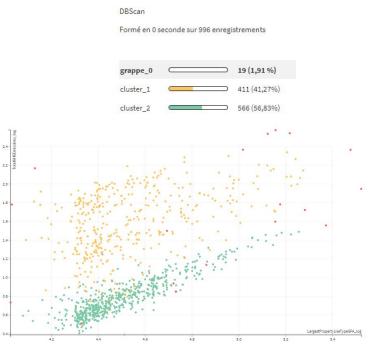




# Clustering

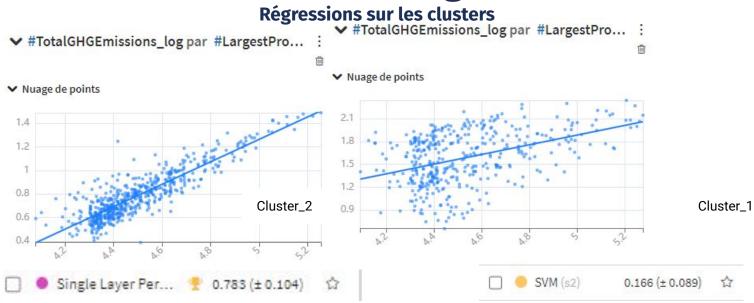
#### Multifamily (1-4) issue de BuildingType





DBScan (s9)

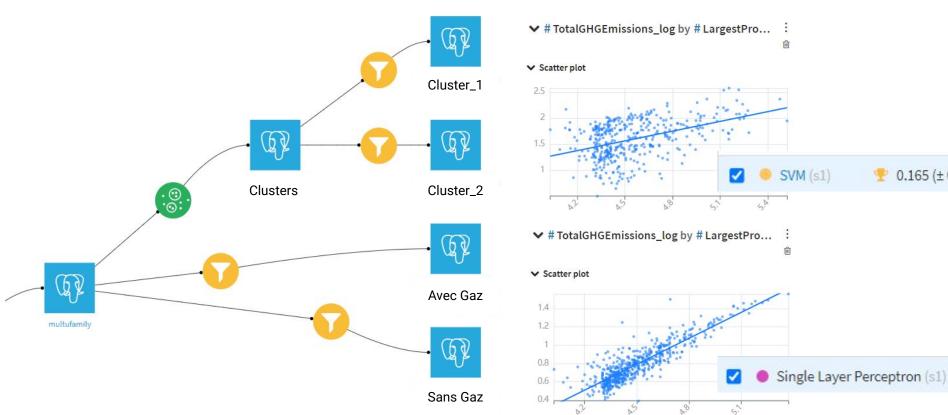
# Clustering



- 100% of the cluster has false for Have\_NaturalGas\_Energy (against 57.13 % globally)
- TotalGHGEmissions\_log is in average 30.39% smaller: mean of 0.79 against 1.13 globally
- LargestPropertyUseTypeGFA log is in average 0.43% smaller: mean of 4.50 against 4.52 globally

- 100% of the cluster has true for Have\_NaturalGas\_Energy (against 42.87 % globally)
- . TotalGHGEmissions\_log is in average 39.98% greater: mean of 1.59 against 1.13 globally
- LargestPropertyUseTypeGFA\_log is in average 0.45% greater: mean of 4.54 against 4.52 globally

Multifamily (1-4) issue de BuildingType, split avec ou sans gaz



# **Applications**

# Recommandations et retour d'expérience

### Recommendations

- Limiter la superficie des nouveaux bâtiments
- Encourager l'utilisation des énergies alternatives au gaz
- Surveiller les bâtiments énergivores (campus & hôpitaux)
- Revoir le mode de notation EnergyStarScore qui est peu représentatif des émissions de CO2.

# Retour d'expérience sur azure

- Facile à prendre en main
- Problèmes de rôles qui empéchent la gestion de certaines ressources
- Intégration à azure pose des problèmes de contrôle des données
- Azure ml : efficace mais le modèle ne pouvait pas être déployé
- Découverte d'outils intéressants : mlflow et interpretml

# Des questions?

