CommonEUROPE LUXEMBOURD

Modernisation des
Applications IBM i
& Open Source
(webinaire)

Jeudi 25 mars – 14h00 – 17h00



Smarter IBM i Applications Made Easy with Al

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IBM Garage for Systems / Red Hat CoC , Montpellier, France
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Agenda

Introduction to AI & Machine Learning

Solutions on IBM i Free Open Source, AutoML, API's, POWER9 and POWER10

How to get started?
Online Demos, code & materials





Machine learning is everywhere – influencing nearly everything we do





Waze provides a personalized driving experience for its users

The future is now ... on IBM i

Images



Transactional

Machine Learning and AI are everywhere

Transactional Data: Store Level

Transaction ID Date

AMOUNT Card_number

		-	p3913045367	11/24/2010	3:20:32 AM	14.47525	**********2323
			q6781991970	12/4/2010	10:30:10 AM	39.46618	**********6451
			r9178278202	11/2/2010	1:23:21 AM	24.99964	**********2179
			s7329082178	10/8/2010	12:29:40 AM	19.21324	*****5826
	-		t6835491952	10/17/2010	10:36:12 PM	39.56086	********9498
duct_Code			u8333871154	12/5/2010	5:22:59 PM	33.35401	********9379
4516281 0257187	2.32		v1240343519	10/27/2010	4:12:38 AM	49.64481	******5466
2773511	1.89		w5440123613	11/26/2010	10:36:36 PM	24.23247	*********1816
5338161	2.2		×8906115216	11/3/2010	1:57:33 PM	32.45101	**********2662
3194534	3.99		v1400944560	11/25/2010	7:26:45 PM	32.12293	*********4330
7229145	1.6		z3665080889	10/9/2010	11:09:58 AM	21.93351	**********6533
AL	14.47		p6738256686	11/23/2010	10:14:36 AM	21.71996	*********4615
			o8964299443	11/24/2010		15.46741	*******7694
			r9012945206	11/22/2010	2:02:15 PM	31.14201	********7140
			s2116305133	11/23/2010		43.16047	********9208
			t7478724264		12:08:46 PM	40.14018	**********2695
			u9844400319	10/11/2010		23.70907	******2980
			v2388298974	11/23/2010		32.73302	**********6777
amn	le: Flat	File	w5672364584		11:54:37 PM	25.11828	******9934

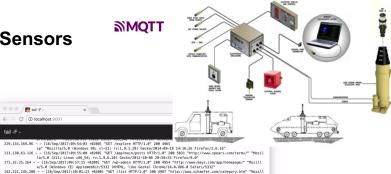
Example: Flat File

a/5.0 (Windows NT 5.1) AppleWebKit/5322 (KHTML, like Gecko) Chrome/14.0.888.0 Safari/5322"
- - [16/Sec/2017:10:06:23 +0200] "DELETE /wp-content HTTP/1.0" 404 5022 "http://www.nielsen-king.com/register.html ts" "Mozilla/5.0 (Macintoshs); PPC Mac OS X 10_6 9; rv:1.9.2.20) Gecko/2015-10-25 10:58:01 Firefou/7.0" "235.99.209.140 - [18/5ep/2017:10:109:11 +0200] "DOS7 /wp-content HTP/1.0" 200 5029 "http://www.niler-foster.org/ "Mozill a/5.0 (Macintosh); pPC Mac OS X 10_5.7 vr.2; it-tTl/Apd/ebekht/1531.0 (GHDM_, like Gecko) Version/5.1 5 fari/531.9.6"

231.252.210.231 - [i6/Sep/2017:10:12:24 +0200] "PUT /app/main/posts HTTP/1.0" 200 5032 "http://www.cisneros.biz/" "Mozilla/ 5.0 (Windows NT 5.2) AppleWebKit/5342 (KHTML, like Gecko) Chrome/15.0.826.0 Safari/5342"

134.55.85.239 - [16/Sep/2017:10:15:50 +0200] "GET /app/main/posts HTTP/1.0" 200 4945 "http://elliott.com/fag/" (X11; Linux x86_64; rv:1.9.5.20) Gecko/2011-12-18 15:51:32 Firefox/6.0" K11; Linux x80_e4; rv:1.9.3.20 Gecko/2011-12-18 15:31:32 Firefox/6.0"
38.226.28.255 - [[6/Sep/2017:10:19:53 40:200] "DELETE /apps/cart.jsp?appID=1494 HTTP/1.0" 301 5058 "http://www.perkins-mendo za.org/categories/sexplore/index.html" "Opera/8.27 (Kindows NT 6.0: it-IT) Presto/2.9.171 Version/12.00" 4.169.212.86 -- [16/Sep/2017:10:21:06 40200] "GET /wp-admin HTTP/1.0" 200 4969 "http://www.graham.org/faq/" "Mozilla/5.0 (Ma cintosh; Intel Mac SX 10_07; rv:1.95.20) Gecko/2011-06-03 22:09:59 Firefox/3.0" 128.242.65.27 -- [16/Sep/2017:10:21:140-0200] "GET /wp-onten HTTP/1.0" 200 5909 "http://www.burgess.com/explore/explore/ex plore/register.html" "Mozilla/5.0 (Macintosh; U; Intel Mac OS X 10_6_8) AppleWebKit/5331 (KHTML, like Gecko) C 168.203.133.24 -- [16/Sep/2017:10:26:18 +0200] "GET /apps/cart.jsp?appID=1968 HTTP/1.0" 200 4972 "http://www.terry.org/" "Mozilla/5.0 (Macintosh; PPC Mac OS X 10_7_8 rv:6.0; it=IT) AppleNebKit/534.26.3 (KHTML, like Gecko) Version/5.0

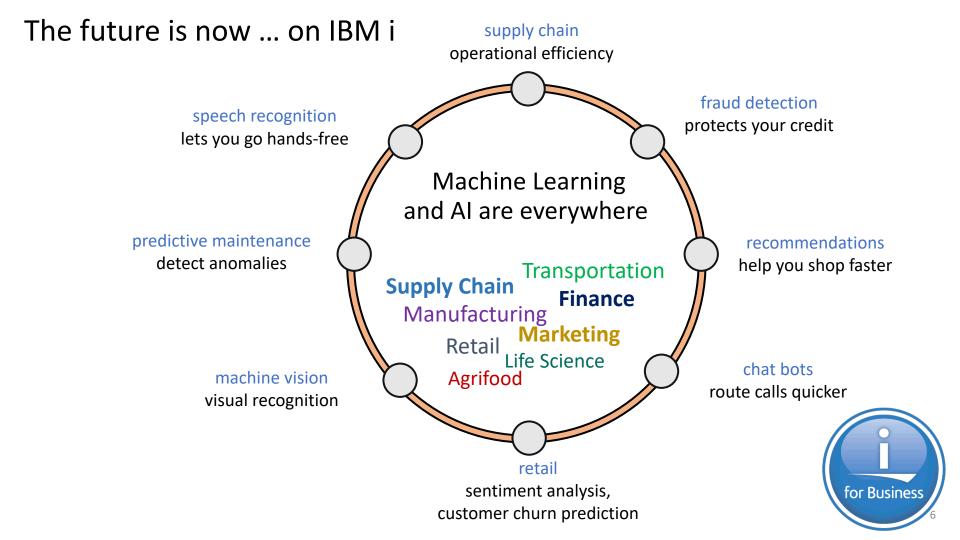
Sensors



Conversation



Log



Why augmenting my apps with AI?

Decision Optimization

Event / Pattern detection (images, structured data), Classification, customer segmentation. Decision Optimization: do more with less, make better decisions, better address my customers, etc. From a simple (but smart) dashboard to complex fraud detection

Augmentation/ Automation

- Repetitive tasks and time-consuming analysis based on images, cases with constrained access... Ex: A doctor could spend hundreds of hours per week to read new studies & publications
- Enhance user experience by exposing new user interfaces image, voice, text, Ex: Chabot (Natural Language Processing)

All is necessary when dealing with complex situations, where rules can't easily be coded with if-then-else.





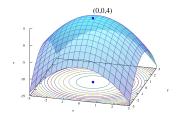


Why augmenting my apps with AI?

UX / Personalization



Business Processes Optimization Do more with less



Gain competitive edge & Grow your Revenue



Why augmenting **IBM** i apps with Al?

 Close relationship between AI, and "core business" data and applications. AI requires Data. Data is the new Oil

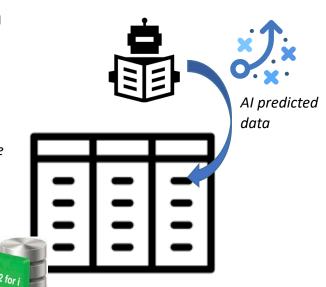


• IBM i: Platform of choice for hosting critical apps and data

→locate AI models and mission-critical applications on a reliable & secure system.

IBM i & AI cases and customer stories

- → Turnover Optimization The good product at the good store
- → Fraud Detection transactional data at work
- → Predictive Maintenance based on IoT data
- → Harvest Prediction per land parcel in Tons
- → Customer Churn prediction augmented CRM
- → Chatbot helpdesk, customer support



Example: Customer Churn CRM & Customer Churn scenario

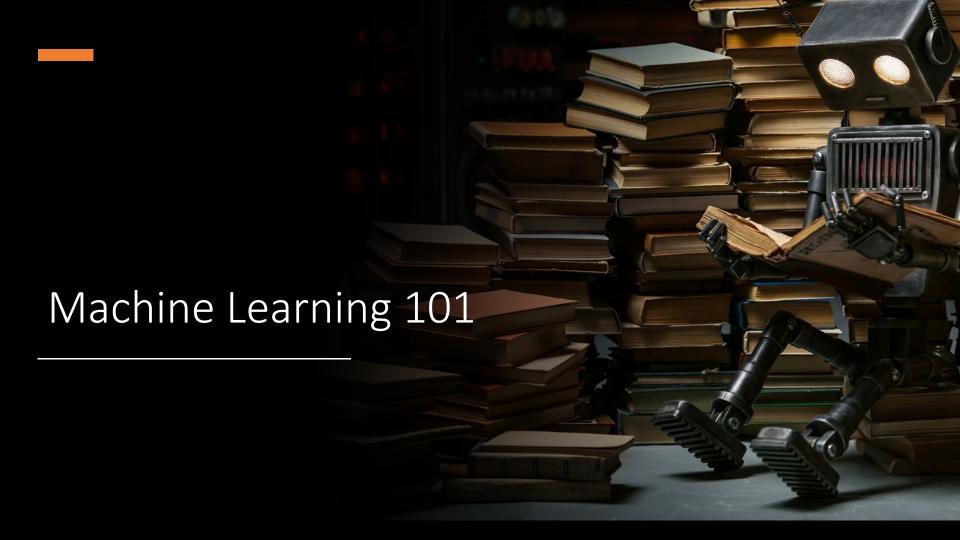
	hboard							^	RE
CUSTOMERID -	GENDER -	STREAMI	CONTRACT A	PAPERLE	PAYMENT	MONTHL	TOTALCH A	CHURN?	sco
7590-VHVEG	Female	No	Month-to-mon	Yes	Electronic check	29.85	29.85		500
5575-GNVDE	Male	No	One year	No	Mailed check	56.95	1889.50		
3668-QPYBK	Male	No	Month-to-mon	Yes	Mailed check	53.85	108.15		REF
7795-CFOCW	Male	No	One year	No	Bank transfer (42.30	1840.75		
9237-HQITU	Female	No	Month-to-mon	Yes	Electronic check	70.70	151.65		
9305-CDSKC	Female	Yes	Month-to-mon	Yes	Electronic check	99.65	820.50		
1452-KIOVK	Male	Yes	Month-to-mon	Yes	Credit card (au	89.10	1949.40		
6713-OKOMC	Female	No	Month-to-mon	No	Mailed check	29.75	301.90		
7892-POOKP	Female	Yes	Month-to-mon	Yes	Electronic check	104.80	3046.05		
6388-TABGU	Male	No	One year	No	Bank transfer (56.15	3487.95		
9763-GRSKD	Male	No	Month-to-mon	Yes	Mailed check	49.95	587.45		
7469-LKBCI	Male	No internet ser	Two year	No	Credit card (au	18.95	326.80		
8091-TTVAX	Male	Yes	One year	No	Credit card (au	100.35	5681.10		
0280-XJGEX	Male	Yes	Month-to-mon	Yes	Bank transfer (103.70	5036.30		
5129-JLPIS	Male	Yes	Month-to-mon	Yes	Electronic check	105.50	2686.05		
3655-SNQYZ	Female	Yes	Two year	No	Credit card (au	113.25	7895.15		
8191-XWSZG	Female	No internet ser	One year	No	Mailed check	20.65	1022.95		
9959-WOFKT	Male	Yes	Two year	No	Bank transfer (106.70	7382.25		
4190-MFLUW	Female	No	Month-to-mon	No	Credit card (au	55.20	528.35		
4183-MYFRB	Female	No	Month-to-mon	Yes	Electronic check	90.05	1862.90		
8779-QRDMV	Male	No	Month-to-mon	Yes	Electronic check	39.65	39.65		
1680-VDCWW	Male	No internet ser	One year	No	Bank transfer (19.80	202.25		

Initial CRM
No Customer Churn risk estimate per customer

Example: Customer Churn CRM & Customer Churn scenario

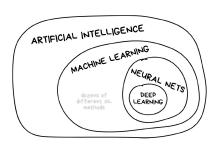
Customers Das	hboard							_	RES
CUSTOMERID -	GENDER -	STREAMI	CONTRACT A	PAPERLE	PAYMENT A	MONTHL	TOTALCH A	CHURN?	2000
7590-VHVEG	Female	No	Month-to-mon	Yes	Electronic check	29.85	29.85	0.21	SCOR
5575-GNVDE	Male	No	One year	No	Mailed check	56.95	1889.50	0.26	
3668-QPYBK	Male	No	Month-to-mon	Yes	Mailed check	53.85	108.15	0.37	REFRI
7795-CFOCW	Male	No	One year	No	Bank transfer (42.30	1840.75	0.39	
9237-HQITU	Female	No	Month-to-mon	Yes	Electronic check	70.70	151.65	0.44	
9305-CDSKC	Female	Yes	Month-to-mon	Yes	Electronic check	99.65	820.50	0.27	
1452-KIOVK	Male	Yes	Month-to-mon	Yes	Credit card (au	89.10	1949.40	0.23	
6713-OKOMC	Female	No	Month-to-mon	No	Mailed check	29.75	301.90	0.18	
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6388-TABGU	Male	No	One year	No	Bank transfer (56.15	3487.95	0.17	
9763-GRSKD	Male	No	Month-to-mon	Yes	Mailed check	49.95	587.45	0.15	
7469-LKBCI	Male	No internet ser	Two year	No	Credit card (au	18.95	326.80	0.23	
8091-TTVAX	Male	Yes	One year	No	Credit card (au	100.35	5681.10	0.34	
0280-XJGEX	Male	Yes	Month-to-mon	Yes	Bank transfer (103.70	5036.30	0.19	
5129-JLPIS	Male	Yes	Month-to-mon	Yes	Electronic check	105.50	2686.05	0.13	
3655-SNQYZ	Female	Yes	Two year	No	Credit card (au	113.25	7895.15	0.20	
8191-XWSZG	Female	No internet ser	One year	No	Mailed check	20.65	1022.95	0.13	
9959-WOFKT	Male	Yes	Two year	No	Bank transfer (106.70	7382.25	0.46	
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1680-VDCWW	Male	No internet ser	One year	No	Bank transfer (19.80	202.25	0.35	

Augmented CRM with AI including Customer Churn risk estimate

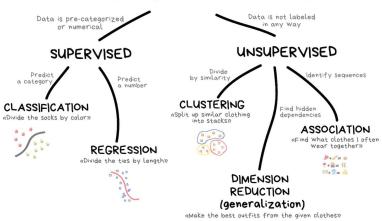


Machine Learning

- Born in the 1950's , ML relies on computing power, mathematics libraries and Data
- Learns from data, no explicit programming
- Establishes causation relationships in the data by 'observing' a phenomenon
- Data knowledge & expertise are necessary

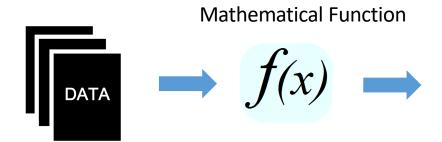


CLASSICAL MACHINE LEARNING





Machine Learning



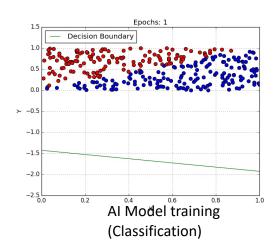


- Credit card transaction
- Loan application
- MRI image
- House data

- Fraudulent vs. legitimate
- Approve vs. reject
- Tumor benign vs. malignant
- House appraisal value
- Representing pattern by a mathematical function
- Machine learning is just a bunch of math

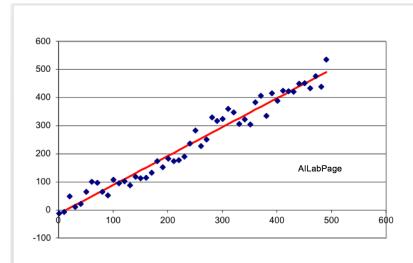
Machine Learning Loan Application Example

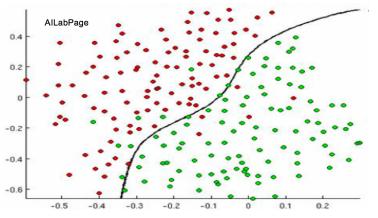
Feature	Feature	Feature	Feature	Label (values to predict in the future)	
Loan Requested	Income	Own House	Outstanding Debt	Decision	
\$20,000	\$100,000	Υ	0	Approve	
\$50,000	\$70,000	N	\$20,000	Reject	
\$5,000	\$150,000	Υ	\$10,000	Approve	
x ₁ x ₂	x ₁₁ x ₂₁	X ₁₂ X ₁₃ X ₂₂ X ₂₃	x ₂₄	У ₁ У ₂	
x ₃ x _n	 X _{n1}	X ₃₂ X ₃₃ X _{n2} X _{n3}		y ₃ y _n	
	_	X	7	Υ	



- → Learn from your data without explicitly coding the rules & relationships between X (features) and the decision Y
- → Once trained, the optimal parameters (here, A and B matrices) are fixed and materialize the model (function f).

Machine Learning







Regression

The system attempts to predict a value for an input based on past data.

Example – 1. Temperature for tomorrow



Classification

In classification, predictions are made by classifying them into different categories.

Example – 1. Type of cancer 2. Cancer Y/N

AlLabPage



Al Tooling trends



1. Ready to Use – simple customization

Possible model re-training with your own data. (specialization)

Ex: Integration w/ Watson API (Chatbot, Image Processing, Text Processing etc.)



2. Auto ML

Bespoke accelerated modeling. Chargeable Licensed products

- H2O Driverless AI, SAS Viya, ...
- Cloud Pak for Data: IBM Watson Studio & AutoAl / CLPEX Optimization Studio



3. Al Libraries

Bespoke models, total control, free or licensed products.

- ML/DL: Scikit learn, R, Rstudio, H2O-3, PyTorch, TensorFlow, Caffe, Chainer...
- Apache Spark, Nvidia CUDA, KubeFlow, IBM Snap ML, Appliances (Nvidia, IBM AC922...), OpenCE, etc.

Relax...All you need is available on IBM i!



Al Applications on IBM i

Dashboard (Web Query...) with additional insight, Augmented ERP, CRM, or Core business applications, etc.



1. Ready to Use – simple customization

Ex: Watson Conversation/Translation REST API integrated with ILE/PASE/Db2 Training in the Cloud (IBM Cloud PaaS or Private with Cloud Pak for Data)



2. Al Libraries & tools in PASE

PASE Libraries : Scikit learn, R, NLTK , etc.



3. Assisted ML: Auto ML

Ex: AutoML models running (inference) in PASE (Scikit learn auto-sklearn etc)
H2O Driverless AI Scoring Pipeline running in PASE (Java, Python) / Db2 Java UDF
Inference: PASE ←→ ILE Integration
Training in PASE or offloaded on Linux/other Clouds.

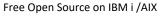
IBM Solutions : AI/ML/DL

AutoML Multi-Cloud (Private/Public)











Data Engineer

Structured Data

Tabular data: IoT/Logs/Database Acceleration & AutoML with Driverless AI



AutoML "Data Scientist in a Box" Training on Linux, Scoring on AIX/IBM i!

All use cases

Program your models w/ Open source & Accelerated HW . Scale w/o limit

Vision/Acoustic

Retail/Inspection /Security, Quality...

All use cases AI Cockpit **AutoML**

	Open Source Libraries AIX / IBM i	H₂ ^{O.ai} Driverless AI	Open-CE	WML-A	Visual Insights	WML Studio
Description	ML on IBM i /AIX No GPU, runs on CPUs Traditional datasets, Relevant for Model Inference (production) Enrichable with Watson Services	model "explainability",	Free, Easy to install High Performance DL LMS (Large Model) DDL (Distributed) TensorFlow* PYTÖRCH Chainer	Enterprise Solution for Deep Learning – Advanced Features : DDL 4+ Nodes HPC Job Scheduler	Ready to use Deep Learning with Video tools Train/Infer/Edge	Collaborative ML/DL WML dev & deployment AutoAI Use WMLCE if installed Can Use WML-A plugin
Support	Avaialable from IBM	H20 L 1-3	Available from IBM	IBM L 1-3 Included	IBM L1-3 Included	Available from IBM
Environments Server Technologies	AIX , IBM i, Linux	x86, IC/AC922, LC9xx	x86/AC922/ IC922	x86/AC922/ IC922	x86/AC922/ IC922	x86/AC922/ IC922

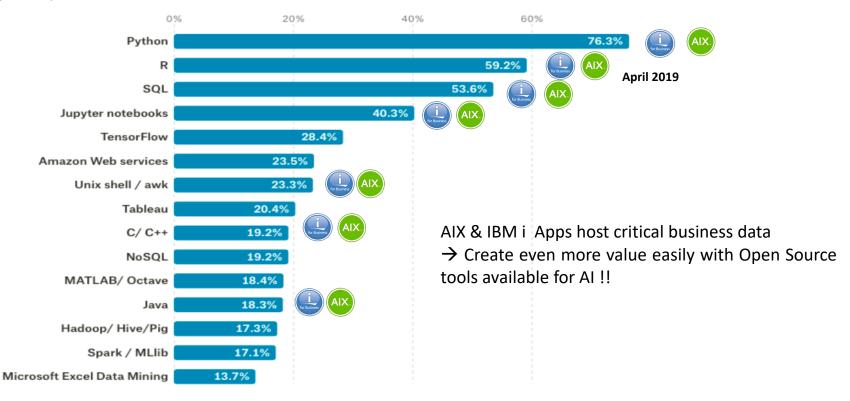




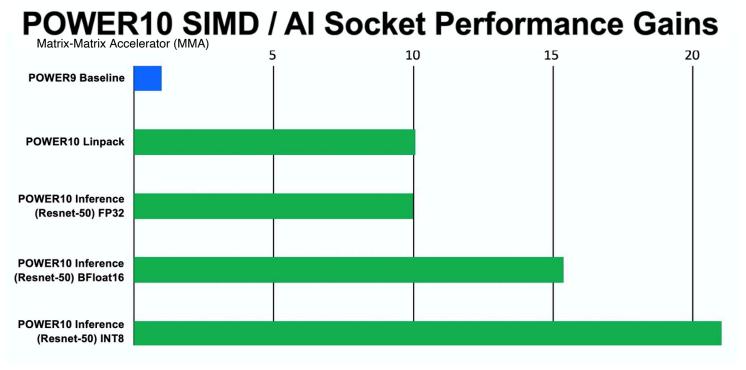
Al technologies available on IBM i

What tools do you use for AI & Analytics projects?

AI - Kaggle Survey results 2017



Analytics, AI & IBM Systems: What's next after POWER9?



- → Al Infusion into Enterprise Databases (Oracle, SAP, DB2)
- → AutoAl & Advanced Analytics : H2O, SAS Viya, IBM Cloud Pak for Data

Al Applications on IBM i

Dashboard (Web Query...) with additional insight, Augmented ERP, CRM, or Core business applications, etc.



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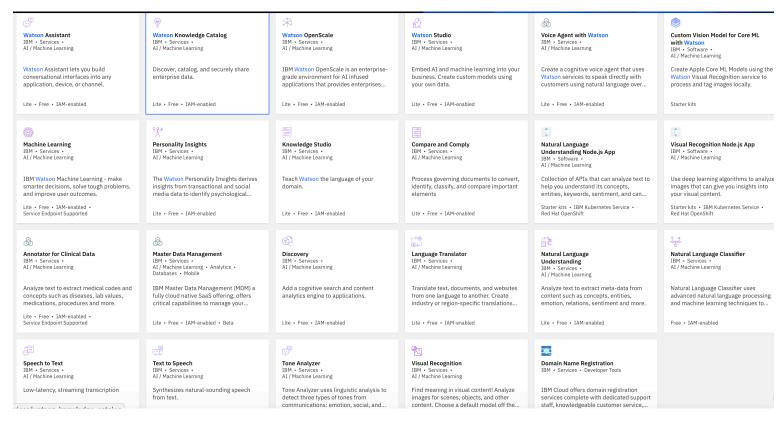
Ex: AutoML models running (inference) in PASE (Scikit learn auto-sklearn etc)
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Training in PASE or offloaded on Linux/other Clouds.

1_ Al Model consumption – IBM Watson API Example

- NLP, translation, tonality
- Documents (annotations..)
- images
- Voice

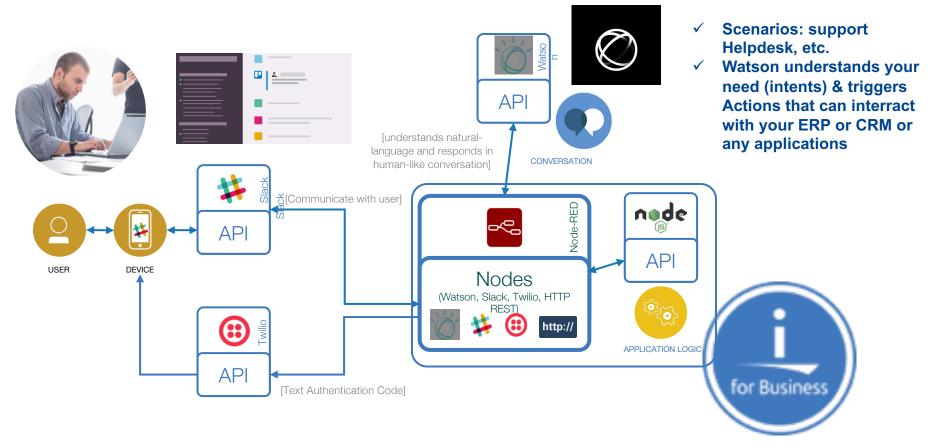
Integration via REST API or via Db2 for I data connector

Available on any Cloud (AWS, IBM, Azure, On Premise...)

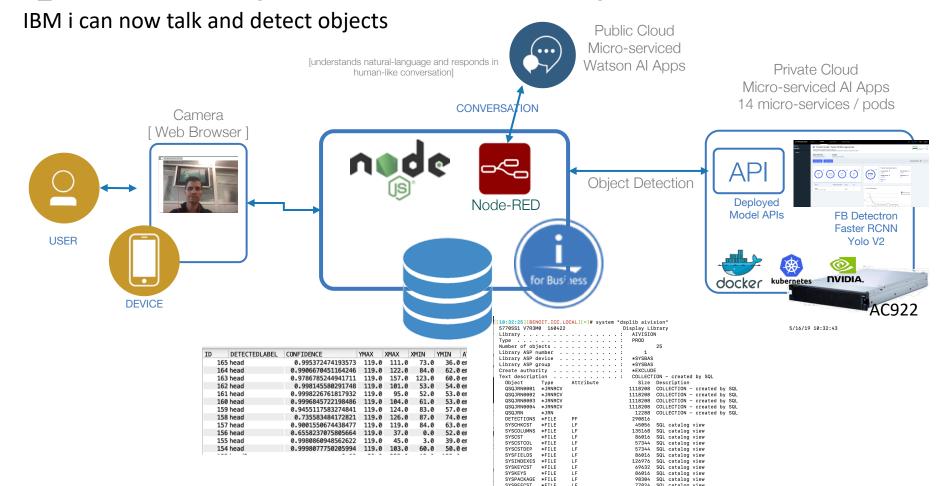


1_ Al Model consumption – IBM Watson API Example

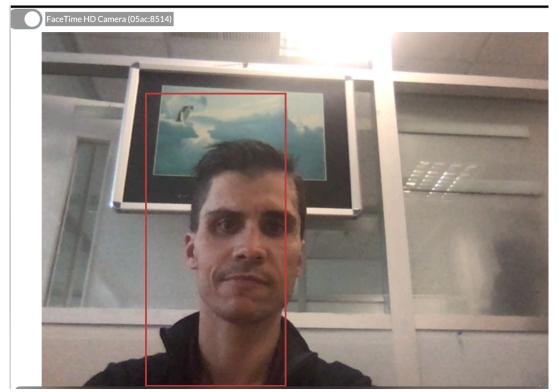
Chabot (Assistant) on IBM i



1_ Al Model consumption – IBM Watson API Example



Computer Vision on IBM i IBM i can now talk and detect objects



Visual Recognition Demo on IBM i + PowerAI Vision





Do try this at home!

https://github.com/bmarolleau/paiv-ibmi (need to be adapted if IBM Visual Insights is not used ☺)

Al Applications on IBM i

Dashboard (Web Query...) with additional insight, Augmented ERP, CRM, or Core business applications, etc.

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Librairies & languages on AIX & IBM i

Data & Scientific Packages Available

Numpy, Pandas: Data Processing

Scipy, Scikit-Learn

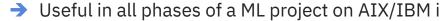
IPython: interactive Python

NLTK: Natural Language Processing

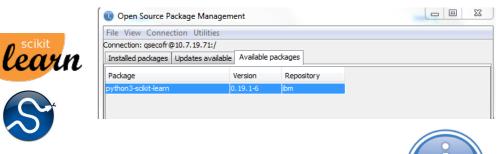
Matplotlib, Jupyter: Data Visualization

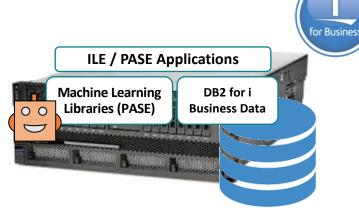
R Language (Interpreter, Runtime)

More to come? ©



- Data Visualization/ Preparation
- → Model creation , training, inference as well
- → GPU Acceleration often needed for large datasets (price / performance) → IC922 / AC922 + WML-CE





scikit-learn



Classification

Identifying to which category an object belongs to.

Applications: Spam detection, Image

recognition.

Algorithms: SVM, nearest neighbors,

random forest, ... — Examples

Regression

...

Predicting a continuous-valued attribute associated with an object.

Applications: Drug response, Stock prices. **Algorithms**: SVR, ridge regression, Lasso,

Examples

Clustering

Automatic grouping of similar objects into sets.

Applications: Customer segmentation, Grouping experiment outcomes

Algorithms: k-Means, spectral clustering,
mean-shift, ... — Examples

Dimensionality reduction

Reducing the number of random variables to consider.

Applications: Visualization, Increased

efficiency

Algorithms: PCA, feature selection, nonnegative matrix factorization. — Examples

Model selection

Comparing, validating and choosing parameters and models.

Goal: Improved accuracy via parameter tuning

Modules: grid search, cross validation,
metrics.

— Examples

Preprocessing

Feature extraction and normalization.

Application: Transforming input data such as text for use with machine learning algorithms. **Modules**: preprocessing, feature extraction.

- Examples

• Packages to install – 1.2 GB

Package (RPM)	Description
Tcl	TCL language support
Tk	TK package for GUI support of TCL
Python3	Python 3 support
Python3-devel	Python3 development package
python3-ibm_db	DBAPI support backage for IBM i
Pyton3-numpy	Numpy package
Python3-scipy	Scipy package
Python3-scikit-learn	Scikit-learn package
Libzmq	Zero-MQ library





Package Python (user)

Yum Install (*ALLOBJ)

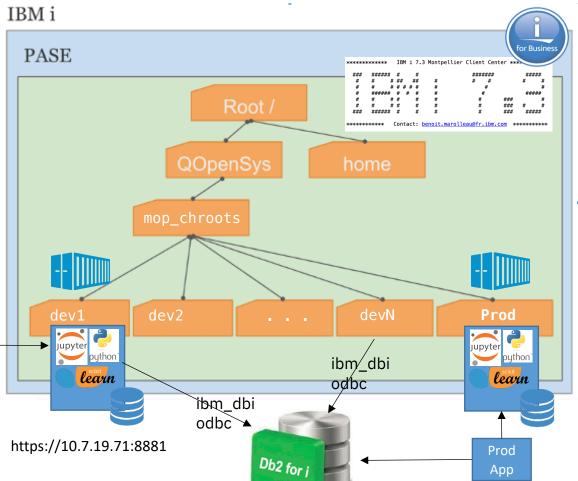
Package Python (PIP)	Description
juypterlab	Includes jupyter (Julia/Python/R IDE)
joblib	Model and large objects persistence on Filesystem

SSH

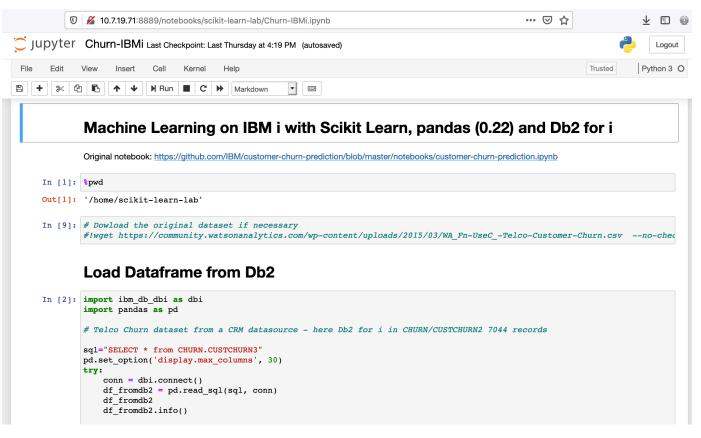
DevOps setup for Al

- Chroot jails (Isolation in PASE/IFS) ☺
- n Data Scientists = n sand boxes (chroot)
- Dev + Ops Environments
- 1.2GB per **Scikit-learn** environment
- Rest of the system (/) isolated

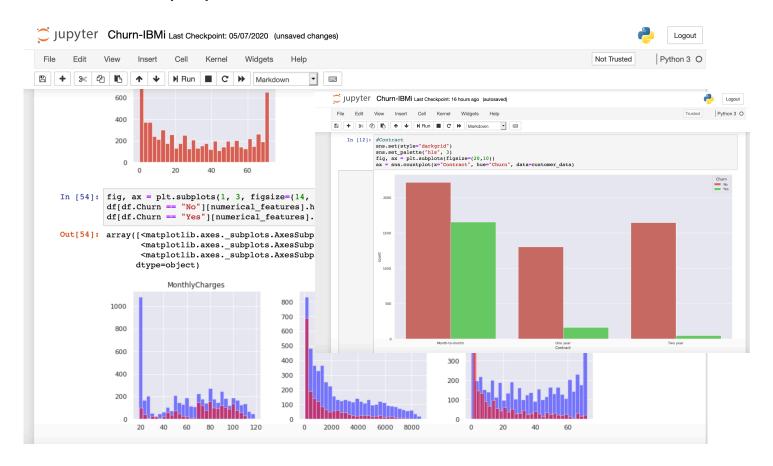
```
yum install
--installroot /QOpenSys/mop chroots/dev1
tk
pvthon3
python3-devel
python3-ibm db
python3-numpy
python3-scipy
pvthon3-scikit-learn
python3-pyzmq
libzmg
python3-pip
pvthon3-pandas
opensssl
bash
ssh dev1@BENOIT.ICC.LOCAL
bash-4.4$ pip3 install jupyterlab
bash-4.4$ pip3 install joblib
bash-4.4$ jupyter notebook --port 8881 --ip 10.7.19.71
```



Free Open Source on IBM i Model Training with data from Db2 for i

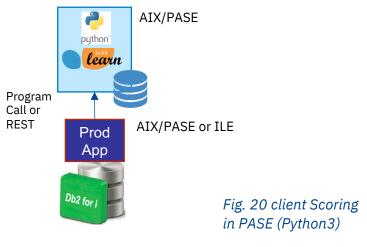


Analytics Dashboard, Data preparation

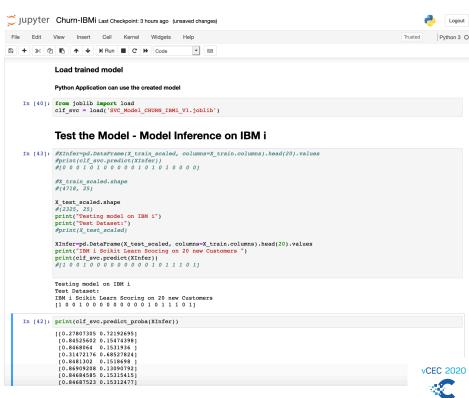


Model integration with your applications (GO LIVE)

- 1. Model Training/Validation (on IBM i or not)
- Model transfer & loading to IBM i (IFS/JFS)
- 3. Scoring (Inference) on IBM i (PASE)
- 4. Integration of the model in your app



Using production data



Free Open Source on IBM i: Smarter apps made easy CRM & Customer Churn scenario

Customers Dasi	hboard						1	^	RES
CUSTOMERID	GENDER -	STREAMI	CONTRACT A	PAPERLE	PAYMENT	MONTHL	TOTALCH	CHURN?	sco
7590-VHVEG	Female	No	Month-to-mon	Yes	Electronic check	29.85	29.85	0.21	500
5575-GNVDE	Male	No	One year	No	Mailed check	56.95	1889.50	0.26	
3668-QPYBK	Male	No	Month-to-mon	Yes	Mailed check	53.85	108.15	0.37	REF
7795-CFOCW	Male	No	One year	No	Bank transfer (42.30	1840.75	0.39	
9237-HQITU	Female	No	Month-to-mon	Yes	Electronic check	70.70	151.65	0.44	
9305-CDSKC	Female	Yes	Month-to-mon	Yes	Electronic check	99.65	820.50	0.27	
1452-KIOVK	Male	Yes	Month-to-mon	Yes	Credit card (au	89.10	1949.40	0.23	
6713-OKOMC	Female	No	Month-to-mon	No	Mailed check	29.75	301.90	0.18	
7892-POOKP	Female	Yes	Month-to-mon	Yes	Electronic check	104.80	3046.05	0.30	
6388-TABGU	Male	No	One year	No	Bank transfer (56.15	3487.95	0.17	
9763-GRSKD	Male	No	Month-to-mon	Yes	Mailed check	49.95	587.45	0.15	
7469-LKBCI	Male	No internet ser	Two year	No	Credit card (au	18.95	326.80	0.23	
8091-TTVAX	Male	Yes	One year	No	Credit card (au	100.35	5681.10	0.34	
0280-XJGEX	Male	Yes	Month-to-mon	Yes	Bank transfer (103.70	5036.30	0.19	
5129-JLPIS	Male	Yes	Month-to-mon	Yes	Electronic check	105.50	2686.05	0.13	
3655-SNQYZ	Female	Yes	Two year	No	Credit card (au	113.25	7895.15	0.20	
8191-XWSZG	Female	No internet ser	One year	No	Mailed check	20.65	1022.95	0.13	
9959-WOFKT	Male	Yes	Two year	No	Bank transfer (106.70	7382.25	0.46	
4190-MFLUW	Female	No	Month-to-mon	No	Credit card (au	55.20	528.35	0.17	
4183-MYFRB	Female	No	Month-to-mon	Yes	Electronic check	90.05	1862.90	0.52	
8779-QRDMV	Male	No	Month-to-mon	Yes	Electronic check	39.65	39.65	0.15	
1680-VDCWW	Male	No internet ser	One year	No	Bank transfer (19.80	202.25	0.35	

Augmented CRM with AI including Customer Churn risk estimate

Al Applications on IBM i

Dashboard (Web Query...) with additional insight, Augmented ERP, CRM, or Core business applications, etc.

1. Ready to Use – simple customization

Ex: Watson Conversation/Translation REST API integrated with ILE/PASE/Db2 Training in the Cloud (IBM Cloud PaaS or Private with Cloud Pak for Data)

2. Al Libraries & tools in PASE

PASE Libraries : Scikit learn, R, NLTK , etc.



3. Assisted ML: Auto ML

Ex: AutoML models running (inference) in PASE (Scikit learn auto-sklearn etc)
H2O Driverless AI Scoring Pipeline running in PASE (Java, Python) / Db2 Java UDF
Inference: PASE ←→ ILE Integration
Training in PASE or offloaded on Linux/other Clouds.

3_ Auto ML

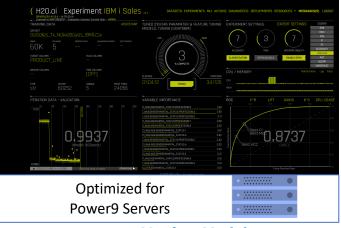
H₂O.ai

http://ibm.biz/PowerVUG

Example with H2O Driverless Al

Import & Visualize Data Create & Test Models





Monitor Models & Iterate

Une Solution simple et abordable :

- "Expert Data Scientist in a Box"
- Months down to Hours
- **Explainability & Transparency**

Model Export







Recommendation Engine

Scoring Pipeline



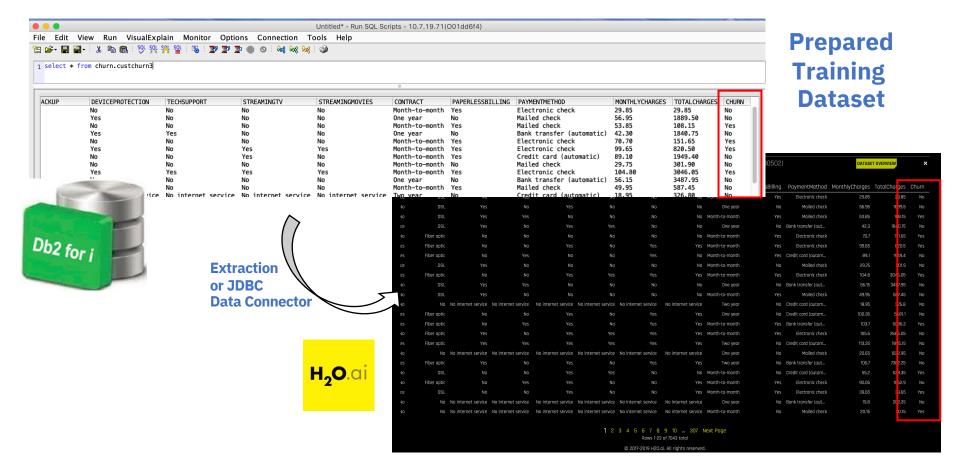
Augment IBM i Applications Model Inference



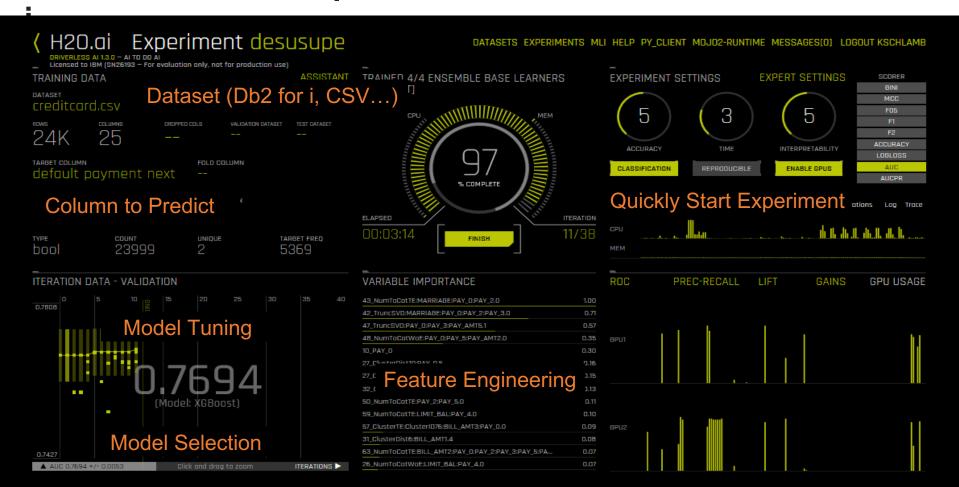


Smarter IBM i apps made easy with Driverless AI

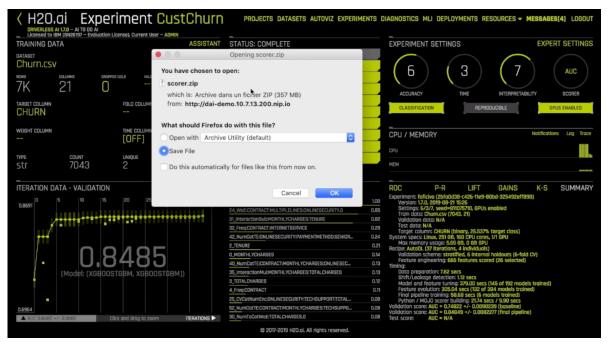
Dataset creation from Db2 for i



The Driverless Al Experience – Dataset from IBM



Export Scoring Pipeline



Scoring Pipeline export

Automatic Scoring Pipelines

Export ultra-low latency Python or Java Automatic Scoring Pipelines that include feature transformations and models.

Here, the Java/ Python Scoring Pipeline "scorer.zip" to be deployed on the inference system

Inference on the Edge , datacenter, Cloud ... on any device.

Export the Java/Python Scoring Pipeline (model) and run it on IBM i!

3_ Auto ML: Demo: Augmented CRM with Al

VouTube

Client Scoring for evaluating the churn risk / probability base on historical data

https://www.youtube.com/watch?v= QemqAzpyJPc&feature=youtu.be

Scoring Model Precision: 84%

Manually, with basic ML knowledge: 0.79%

Driverless AI: Feature Engineering like a Kaggle grandmaster.

Time to market: reduce it from months to days

Auto ML solutions with Hardware acceleration allow you to create high quality models, comparable to a ww ML grandmaster, very quickly, with only a good knowledge on ML.

With Driverless AI, The shaped model (Scoring Pipeline) can be executed on AIX or IBM i close to the core business applications and database.

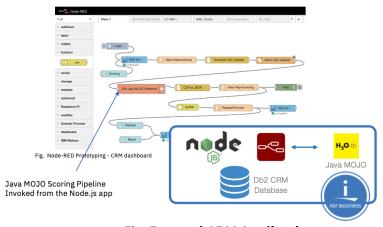


Fig. **Boosted CRM Application**w/ integration of a real time scoring
Pipeline (Java MOJO)
ILE ←→ PASE Integration

H2O Driverless AI on AC922 Client Reference in Financial Services

Improving Profitability of Core Financial Services



H₂O.ai

- Data extracted from IBM Db2 for i
- Improved years of expert feature engineering
- Increased accuracy of existing credit risk scoring in less time
- 2x propensity to buy for new bank products
- Accelerated by IBM Power Systems AC922

"By using Driverless AI on IBM Power Systems AC922, we have been able to build out ML/AI solutions in less time, with improved quality and accuracy."

"We were also able to double the propensity for our banking customers to accept an offer of credit products, such as credit cards... We plan to use the platform for more use cases in the future."

"This awesome result is because Driverless Al uses the expertise of "Kaggle Grand Masters in a Box."

<u>Vison Banco -</u> <u>IBM Customer</u> <u>Case Study</u>

Ruben Diaz Data Scientist, Visión Banco





How to get started?

How to Start With Machine Learning on IBM i (by Gan Zhang, IBM - Nov 2019)

Accessing the rest of your IBM i from Python (Kevin Adler) ibm dbi driver presentation

Join the #IBMiOSS Linkedin Group

Check out the IBM i OSS Examples - Github Repository

YiPS http://yips.idevcloud.com/wiki/

Bitbucket IBM i Open Source

My blog (K8s, App modernization, ML, H2O, Watson, Node-RED, IBM i)

Videos 2020 (AIX/IBM i in IBM Cloud , Terraform) My Youtube Channel





How to get started?











https://cognitiveclass.ai/courses/data-analysis-python

https://www.coursera.org/specializations/data-science-python

How to get started? Demos

→ 'Scikit-learn on IBM i' Demonstration on Youtube

https://youtu.be/Uw ePb8Hz3o?t=1296



- → Do it Yourself: Install yum packages and git clone
 - https://github.com/bmarolleau/firstdemo-scikitlearn-ibmi/
 - Free Auto-ML? Auto-sklearn from Scikit-learn
 - Need support or a live demo? Contact me!



→ IBM i & AutoML: H2O Driverless AI

https://www.youtube.com/watch?v=QemqAzpyJPc&feature=youtu.be



Modernisation des Applications IBM i & Open Source (webinaire) Jeudi 25 mars - 14h00 - 17h00



Smarter IBM i Applications made easy with Al

Benoit MAROLLEAU – Cloud/Al Architect

IBM Garage for Systems - Montpellier, France benoit.marolleau@fr.ibm.com

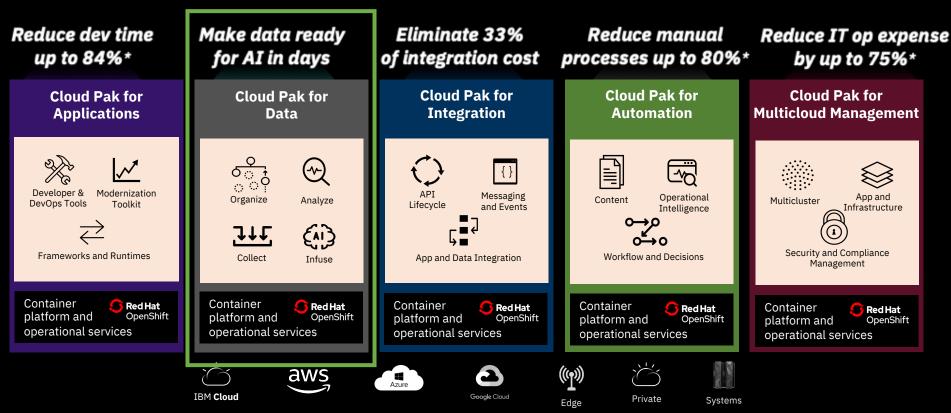


#IBMi #IBMiOSS Fan



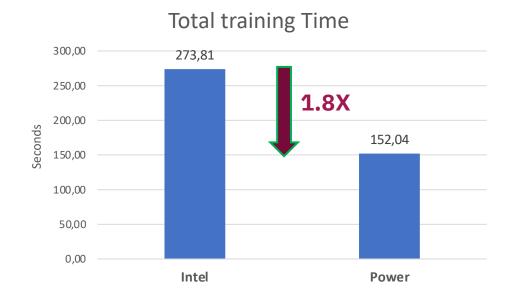
Cloud Paks – Pre-integrated for cloud use cases

Today, IBM offers clients the first five Cloud Paks...



LightGBM Algorithm is Faster on Power

H2O Driverless AI uses LightGBM as one of its main ML Algorithms



Advantages of LightGBM

- 1. Faster training speed and higher efficiency
- 2. Lower memory usage
- 3. Better accuracy than any other boosting algorithm
- 4. Compatibility with Large Datasets
- 5. Parallel learning supported.

LightGBM on Power (CUDA) is 1.8X faster than x86 (OpenCL)

Machine Learning

Use training data to derive f(x) so that

or mathematically
$$\min \quad 1/n \sum_{i=1}^n (y_i - f(x_i))^2$$

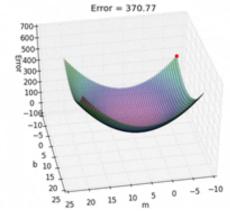
	Feature	Feature	Feature	Feature	Label (=f(X))		
	Loan Requested	Income	Own House	Outstanding Debt	Decision		
Vector X Input	\$20,000	\$100,000	Υ	0	Approve	Output	
	\$50,000	\$70,000	N	\$20,000	Reject	:t	
	\$5,000	\$150,000	Υ	\$10,000	Approve		
	x ₁	x ₁₁ x ₁₂	x ₁₃ x ₁	4 y 1	1		
	X ₂	X ₂₁ X ₂₂	X ₂₃ X ₂	4 y 2			
	X ₃	x ₃₁ x ₃₂	X ₃₃ X ₃	4 y 3			
	x _n	\mathbf{x}_{n1} \mathbf{x}_{n2}	X _{n3} X _n	4] <u>Y</u> n	_		

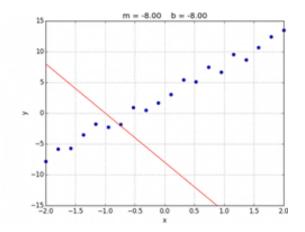
- Learning phase: minimize the distance between the current result f(x) and the actual result (supervised learning)
- Volume & Data quality are crucial (Training Dataset). Knowledge of the observed phenomenon, business.
- Each algorithm has **parameters**, adjusted (learned) according to the obtained errors during training.
- The choice of the algorithm is also key, as well as **hyperparameters** initialization.
- · At the end of the training, the parameter values are fixed
- The execution of the validated, tested model on new examples is called **inference**
- Accelerators GPU (or FPGA) are useful in the Training Phase, sometimes in Inference : thousands of cores, good at matrix and vector calculation (tensors) can divide a training time by 50

→ Solutions exist to assist in all these phases (Relax ②)

Machine Learning

Error Function to minimize (almost) at each iteration.



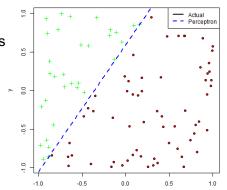


Regression

The model is getting better at each training iteration (while minimizing the error function)

Classification

- In reality , hundreds/thousands of dimensions (or features or columns)
- Dimensionality Reduction : eliminate redundant, correlated, useless dimensions
- Feature Extraction: creation of additional features to better undertand the phenomenon. A neural network does this extraction automatically, major difference vs. Classic Machine Learning.
- Model quality: precision on test dataset. Does it work fine with new data? overfitting vs. underfitting, bias & variance. Regularization methods...



Online Perceptron Iteration 1

→ Solutions exist to assist in all these phases (Relax ②)*

Machine Learning Solutions

On Premises? Public Cloud?

AI libraries & frameworks

- ML: scikit learn, R, Rstudio, SQL, H2O-3, H2O Driverless AI ...
- DL: PyTorch, TensorFlow, Caffe, Theano, Chainer ...
- Distributed: Apache Spark, H2O Sparking Water, HPC tooling...
- · Accelerated ML: Nvidia CUDA, Rapids, IBM Snap ML
- → Hundreds to come every month...most of them are **free open source**

Desktop & Server Appliance with Accelerators (GPU)

Nvidia DGX , Intel ... GPU / FPGA based etc.

IBM AC922: Oak Ridge 'Summit' SuperComputer Building Block



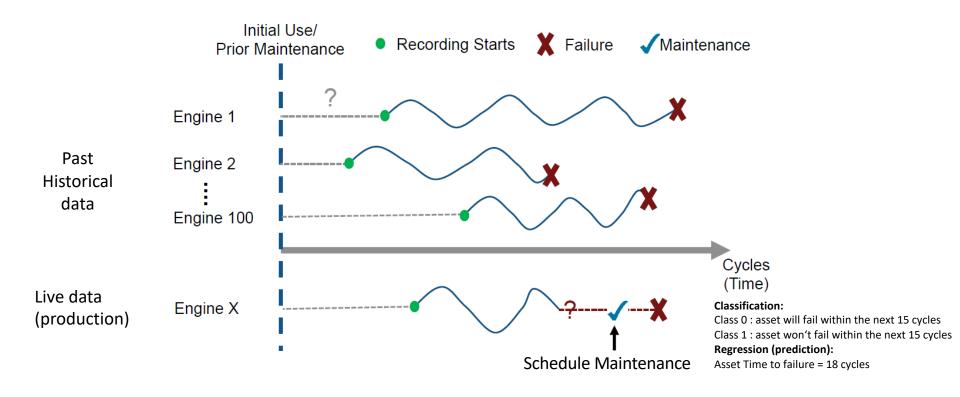




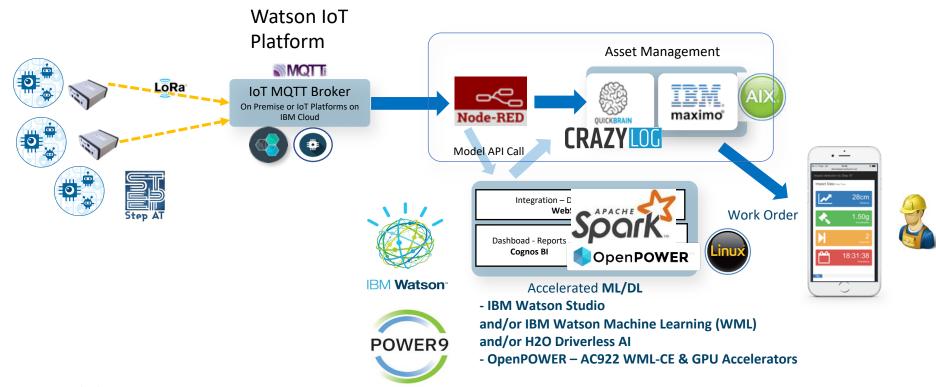
Microsoft



Example: Predictive Maintenance

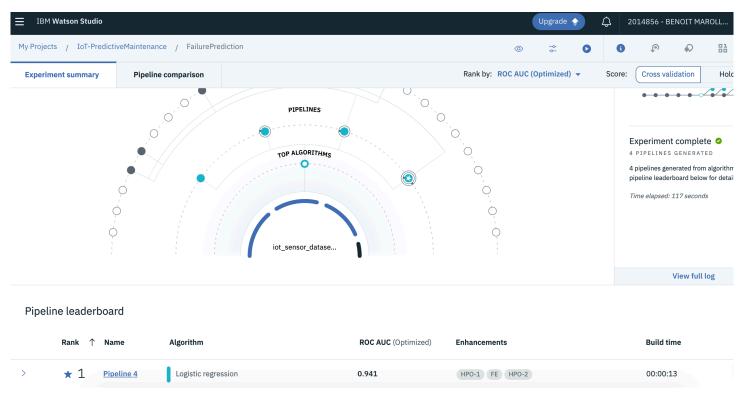


Example: Predictive Maintenance



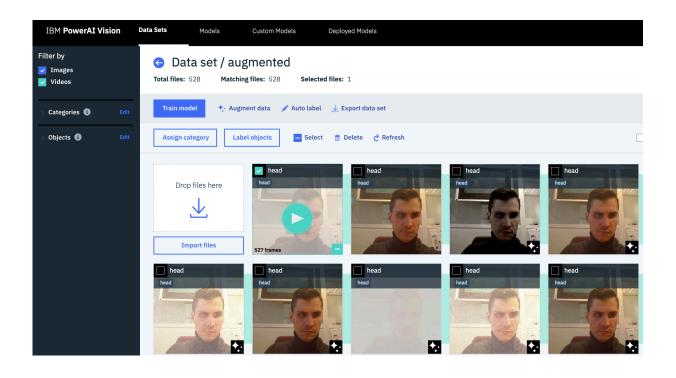
Example: Predictive Maintenance

with Watson Studio



Model created with AutoAI on Studio. Can be deployed on WML (REST API...)

Computer Vision with IBM Visual Insights & IBM i



• IBM POWER SYSTEMS

AC922



An Acceleration Superhighway

Unleash state of the art IO and accelerated computing potential in the post "CPU-only" era



Designed for the AI Era

Architected for the modern analytics and AI workloads that fuel insights



Delivering Enterprise-Class Al

Flatten the time to AI value curve by accelerating the journey to build, train, and infer deep neural networks

