

Future World Enters Twilight Zone

Machine Learning, AI, VR, Robotics & Quantum Computing Ahead



(Welcome to the Fifth Dimension)





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Special Thanks:

Kay Cavender, Charles Kim, Monica Lee, & Linda Hoover

Rich's Overview *(a)***richniemiec**

ORACIE ACE Director



- Chief Innovation Officer, Viscosity North America: rich.niemiec@viscosityna.com •
- Former CEO of TUSC
 - Inc. 500 Company (Fastest Growing 500 Private Companies)
 - 10 Offices in the United States (U.S.); Based in Chicago •
 - **Oracle Advantage Partner in Tech & Applications**
- Former President Rolta TUSC & President Rolta EICT International
- Author (5 Oracle Best Sellers #1 Oracle Tuning Book for over a Decade):
 - Oracle Performing Tips & Techniques (Covers Oracle7 & 8i)
 - Oracle9i Performance Tips & Techniques •
 - Oracle Database 10g Performance Tips & Techniques ۲
 - Oracle Database 11g Performance Tips & Techniques •
 - Quick Start Guide to Oracle Query Tuning (2015) ٠
 - Oracle Database 12cR2 Performance Tips & Techniques •
- Former President of the International Oracle Users Group ٠
- IOUG Top Speaker in 1991, 1994, 1997, 2001, 2006, 2007 ٠
- MOUG Current President & Top Speaker Twelve Times ٠
- National Trio Achiever award 2006 ۲
- Oracle Certified Master & Oracle Ace Director ۲
- Chris Wooldridge Award 1998, 2012 ٠
- Chicago Entrepreneur Hall of Fame 1998 ٠
- MIT Machine Learning & AI •
- Purdue Outstanding Electrical & Computer and Engineer Alumni 2007 ٠
- Board Member TEC, Entrigna, Ask DB Experts ٠
- E&Y Entrepreneur of Year & National Hall of Fame 2001





Oracle Database 12c === 12 Release 2 Performance Tuning Tips and Techniques Seat Practices for Optimaling Database Porturnance

Hickard Niterates Carthad Manne, Charles Mill Street

ORACLE **Quick Start Guide to** Oracle Query Tuning: **Tips for DBAs and**

Developers Kol Number



Viscosity Pillars and Delivery Models



Developer Resource Center / **Oracle ACE Program**

Viscosity's Oracle ACES

Oracle ACE Program

The Oracle ACE Program recognizes and rewards community members for their technical contributions in the Oracle community.



Charles Kim, CEO & Co-Founder

Twitter: @racdba

ORACLE ACE Director



Rich Niemiec, Chief Innovation Officer

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Craig Shallahamer **Applied AI Scientist**

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Sean Scott, Consultant

Twitter: @oraclesean







Gary Gordhamer, Consultant

Twitter: @ggordham





Julio Ayapan, Consultant



VISCOBITY NORTH AMERICA

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Implementing 50+ Years of Innovation, **Expertise, and Excellence**



Quick FREE notes

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Agenda – Goal is Apply Technology & W.I.N.

- □ You've Just Entered the Twilight Zone
- □ Twilight Zone that We See Now
- □ Big Data and IOT Data Is Coming Fast
- □ ML & Oracle Overview & ADB
- □ Applications of ML Algorithms & AutoML
- □ Machine Learning Future, Robots & VR

□ Future Tech from the Twilight Zone+

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Where is Everybody?







Where is Everybody?





The Fever







Deprived of Sports, Gamblers Wager Outside Their Home Turf



regular sports bettors, Le. betting money on quorts events at least news a month. Source: Morning Consult

2020 SP

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much is too much?



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Sports betting ads: Industry weighs how

Number 12 Looks Like You (2020)





tioned Panda TikToker Goes Viral By Exposing The Truth Behind Heavily-Edited Pictures

> TikTok guidelines said not to promote posts by ugly and poor users



Campuflage Makeup - Amazon.com



hide or disguise the presence of















The Brain Center at Whipple's







Robotics/Automation Impact to Jobs

109



The New York Times

Uber's Self-Driving Trucks Hit the Highway, but Not Local Roads





Share of workers in occupations at high risk of automa 30% Middle incom 25% 20%

Low-Income Jobs at Highest Risk











Leverage – DB, GPS & Robotics! Impact to Jobs?



The Obsolete Man

















Autonomous Database – Replacing the DBA?



Rich Niemiec @RichNiemiec - Oct 2 I'll ask #Pepper If she can tune my #database in 11 AM session on #innovation #ich #robolics #cloud at #oow17 #ioug #viscosityna #oracleace



Are DBAs Obsolete?

Posted on February 20, 2014

Before we go any further, let me briefly ans "No Way!" OK ... with that out of the way, le

Every so often, some industry pundit gets h "Database administrators are obsolete" or t hear this, it makes me shake my head sadly can be



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OCCUPATIONAL OUTLOOK HANDBOOK

Occupational Outlook Handbook > Computer and Information Technology >

Database Administrators

Statement of the local division of the	Minat Those the	Work Environment	How to Recomp Dee	Draw	Joh
Summary	AALIGE LINEA PAD	WORK ENVIRONMENT.	now to become one	ray	200

Summary

Quick Facts: Database Adr	ninistrators		
2017 Median Pay 😗	\$87,020 per year \$41.84 per hour		
Typical Entry-Level Education 🕢	Bachelor's degree		
Work Experience in a Related Occupation 😨	None		
On-the-job Training 😳	None		
Number of Jobs, 2016 🕢	119,500		
Job Outlook, 2016-26 🜍	11% (Faster than average)		
Employment Change, 2016-26 🥥	13,700		

What Database Administrators Do

Database administrators (DBAs) use specialized software to store and organize data, such as information and customer shipping records. They make sure that data are available to users

Good News: DBA +11% Increase (2016-2026)

Biju Thomas at ODTUG - Emerging Jobs

#8 Data Engineer

Linked DE 2020 Emerging Jobs Report

- 33% annual growth
- Skills
 - Apache Spark
 - Hadoop
 - Python
 - ETL
 - AWS
- Industries
 - Computer Software
 - Information Technology
 - Financial Services
 - Healthcare & Hospitals

- Primary job responsibilities involve preparing data for analytical or operational uses.
- Works as part of an analytics team, providing data in a ready-to-use form to data
- scientists.
 - Commonly deal with both structured and unstructured data sets

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consumes over 80% of the time in Al projects.

What You Need; Nick of Time (Predictive Analytics)

















Characteristics of Big Data - The Five V's



Big Data Themes

- HW & SW technologies for large data volumes
- Focus on Web 2.0 technologies
- Database Scale-out
- Relational & Distributed Data Analytics
- Real Time Analytics

Big Data Domains

- Digital Marketing Optimization
- Data Exploration & Discovery
- Fraud Detection & Prevention
- Social Network & Relationship Analysis
- Machine-generated Data Analytics

Converged Database - Oracle Multi-Model Database*

Benefits of Oracle's Converged Database are broad

*Slide from Oracle's, Nitin Vengurlekar



- Integrated development tools (Apex, SQL Dev, Spatial Studio)
- 3rd party and Open Source development tools
- Machine Learning
- Node.js, Python, many others
- In-memory database
- Spatial, Graph support
- NoSQL (JSON, key-value, wide column, XML)
- Containers, microservices, virtualization (Docker, MT)
- Integrated Security
- Deployment choice (on-prem, cloud, hybrid)
- Integrated High Availability and Disaster Recovery



Become a DA (Data Administrator): Not just DBA



The Hitchhiker (Many Predictive Medical devices & IOT Data)











Pro-active Side



echodot

Add Alexa to any room





A Robot may not look like one!

Oh Yeah... they never complain, always happy to do more, work anywhere, get smarter as time goes on, leverages AI & ML, works 24x7, doesn't ask for a raise, no union (yet).



A Robot may not look one!

*Robots that Manage a Database (ADW)!

*Robots that secure a system and use ML & AI

Autonomous DB : Future DBA & Robot DB

- Self-Managing (Driving)
- Self-Securing
- Self-Tuning
- Self-Recovering
- Self-Scaling Administration

Fully automated patching, upgrades, backups, & availability architecture



Oracle Unveils World's First Autonomous Database Cloud



Reality of the Autonomous Database & ML

Will my job change?

- Absolutely...sure hope so!
- It has many times in the past...
- Closer to the business & Innovation
- Data Critical & ML & Al driven by it.



- Data Admin/Architect/ML instead of DBA
- ML/AI/Data Science Developer instead of just Oracle
- Security Expert instead of Security on the DB
- Watching over costs more
 - Cloud Hidden Costs: Cloud, Hybrid, or On-Site Decisions
 - Decide which databases should be Autonomous

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ATP – Provisioning a Database (2 minutes) **ORACLE** Cloud US East (Ashburn) V N 🕂 O 🌐 O Search resources, services, documentation, and marketplace Overview + Autonomous Database + Autonomous Database Details DB-20220510121752 0 1 0 ORACLE Cloud US East (Ashburn) V Search resources, services, documentation, and marketplace Overview » Autonomous Database » Autonomous Database Details ATP DB-20220510121752 C1 Database Actions DB Connection Performance Hub C Service Console More Actions -PROVISIONING Autonomous Database Information Tools Tags General Information Infrastructure Database Name: DB20220510121752 Dedicated Infrastructure: No AVAILABLE. Workload Type: Transaction Processing Compartment: richniemiec (root) Autonomous Data Guard (1) Status: Disabled Enable Created: Tue, May 10, 2022, 17:19:34 UTC **Oracle Autonomous Database** OCPU count: 1 Backup OCPU auto scaling: Disabled (1) 0 Last Automatic Backup: No active backups exist for this database. Support 10% Storage: 1 TB Manual Backup Store: Not Configured Storage auto scaling: Disabled (i) License Type: License included Network Q4 Growth Rate Database Version: 19c

Annualized Consumption Revenue - ACR

ELite and Privacy Chokie Preferences

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ATP – Scaling Database (50 sec.)



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Terms of Ose and	Database Version: 19c	Annese Time: Allow Serves Annese from supplicates		

ATP – Stop Database (25 sec.)



ORACLE Cloud Search resources, services, do	cumentation, and marketplace	US East (Ashbu
Overview » Autonomous Database » Autonomous Database Details		
DB-2022051	0121752	
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STOPPE	ED Workload Type: Transact	tion Processing Dedicated Infrastructure:

ADW - Provision Database (1 min. 20 sec.)

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ORACLE Cloud Set Overview * Autonomous Database * Autono ADDA PROVISIONING	arch resources, services, documentation, and ma proves Database Details DB-20220510174437 Data Data Oreaview + Autonomous Database + Autor Au G Da We Co oc Cri OC Str Str Lic Pa	earch resources, services, documentation, and marketplace Tomous Database Details DB-20220510174437 C Database Actions DB Connection Performance Autonomous Database Information Cols General Information Database Name: DB20220510174437 Workload Type: Data Warehouse Compartment: nchrisemiec (root) OCID:iabjpa Show Copy Created: Tue, May 10, 2022, 22:57:35 UTO OCPU count: 1 OCPU auto scaling: Disabled () Storege: 1 TB	US East (Ashburn) V N (Ashburn) V N (B) (C) (C) (C) (C) (C) (C) (C) (C) (C) (C
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Autonomous DB: ATP-ADW-AJD-APEX



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ATP – Start Database (30 sec.)



	Cloud	Search resources, service	s, documentation, and marketplace		US East	(Ashbum)		
Overview + Autonomous	Database » A	DB-20220	™ 510121752					
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			AVAILABLE		Workload Type: Trans	saction Processing		200.000

ATP - Service Console (Development/ML)



Oracle Focus: Cloud, Data, Apps, Security, Self-Service, ML, & Al! Innovation Components for Business!

ORACLE

"I admire risk takers. I like leaders – people who do things before





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they become fashionable or popular. I find that kind of integrity inspirational."

LAWRENCE J. ELLISON | Chairman & Chief Executive Officer, 2003

ORACLE



Oracle Machine Learning: Brief Highlights Only

ORACLE SaaS APPS w/EMBEDDED ML Pre-built, packaged ML and data-driven SaaS applications

READY-TO-GO

ORACLE AUTONOMOUS DATABASE Embedded AI in Databases to simplify enterprise data management

READY-TO-WORK

ready-to-Build

READY-TO-BUILD

ORACLE

DATA SCIENCE PLATFORM

Complete platform to build and

support ML-powered

applications

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Machine Learning Process (Supervised Learning)

- First: Clear Business Problem to Solve
- <u>Second</u>: *Function* to Perform
- Third: Algorithms to use
- <u>First</u>: Build/*Train the Model:* When you build it use about 60% of your data.
- <u>Second</u>: *Test/Score Model* for accuracy/precision using about 40% of data.

Compare Algorithms!



Business Understanding

Be Extremely Specific in Problem Statement:

Poorly Defined	Better	Data Mining Technique
Predict employees that leave	 Based on past employees that voluntarily left: Create New Attribute EmplTurnover → O/1 	
Predict customers that churn	 Based on past customers that left (churn): Create New Attribute Churn → YES/NO 	
Target "best" customers	 Recency, Frequency Monetary (RFM) Analysis Specific Dollar Amount over Time Window: Who has spent \$500+ in most recent 18 months 	
How can I make more \$\$?	What helps me sell soft drinks & coffee?	
Which customers are likely to buy?	 How much is each customer likely to spend? 	
Who are my "best customers"?	 What descriptive "rules" describe "best customers"? 	
How can I combat fraud?	 Which transactions are the most anomalous? Then roll-up to physician, claimant, employee, etc. 	

OAA Model Build and Real-time SQL Apply

Simple SQL Syntax - Attribute Importance

ML Model Build (PL/SQL)



Additional Detail:

drop table CUST_INSUR_LTV_SET;
exec
dbms_data_mining.drop_model('BUY_INSURANCE_AI');

create table CUST_INSUR_LTV_SET (setting_name varchar2(30), setting_value varchar2(4000));

insert into CUST_INSUR_LTV_SET values
('ALGO_NAME','ALGO_SUPPORT_VECTOR_MACHINES');
insert into CUST_INSUR_LTV values ('PREP_AUTO','ON');
commit;

Model Results (SQL query)

SELECT attribute_name, explanatory_value, rank
FROM BUY_INSURANCE_AI
ORDER BY rank, attribute_name;

ATTRIBUTE NAME	RANK	ATTRIBUTE VALUE
BANK FUNDS	1	0.2161
MONEY MONTLY OVERDRAWN	2	0.1489
N TRANS ATM	3	0.1463
N TRANS TELLER	4	0.1156
T AMOUNT AUTOM PAYMENTS	5	0.1095

OAA Model Build and Real-time SQL Apply Prediction

Oracle Advanced Analytics (OAA) Simple SQL - Classification

BEGIN DBMS_DATA_MINING.CREATE_MODEL(model_name => 'BUY_INSUR1', mining_function ': => dbms_data_mining.classification; ' data_table_name => 'CUST_INSUR_LIV', case_id_column_name => 'CUST_ID', target_column_name => 'BUY_INSURANCE', settings_table_name => 'CUST_INSUR_LTV_SET'); END; /

Additional Detail:

drop table CUST_INSUR_LTV_SET; exec dbms_data_mining.drop_model('BUY_INSUR1'); create table CUST_INSUR_LTV_SET (setting_name varchar2(30), setting_value varchar2(4000));

insert into CUST_INSUR_LTV_SET values
('ALGO_NAME','ALGO_SUPPORT_VECTOR_MACHINE';
insert into CUST_INSUR_LTV_SET values ('PREP_AUTO','ON');
commit;

Model Apply (SQL query)

```
Select prediction_probability(BUY_INSUR1, 'Yes'
```

USING 3500 as bank_funds, 825 as checking_amount, 400 as credit_balance, 22 as age, 'Married' as marital_status, 93 as MONEY_MONTLY_OVERDRAWN, 1 as house_ownership) from dual;

🖡 📇 🙀 🎭 SQL | All Rows Fetched: 1 in 0.043 seconds

PREDICTION_PROBABILITY(BUY_INSUR 1, YES'USING3500ASBANK_FUNDS, 825ASCHECKING_AMOUNT, 400ASCREDIT_BALANCE

1 0.9276956709910801

Create a Model (FYI Only)

DBMS DATA MINING CRE. model name mining function data table name IN VARCHAR2, case id column name IN VARCHAR2, target column name IN VARCHAR2 DEFAULT NULL, settings table name data schema name settings schema name xform list

IN VARCHAR2, IN VARCHAR2, IN VARCHAR2 DEFAULT NULL, IN VARCHAR2 DEFAULT NULL, IN VARCHAR2 DEFAULT NULL, IN TRANSFORM LIST DEFAULT NULL);

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DBMS_DATA_MINING Algorithms in Oracle*

Algorithm	Abbreviation	Function	*Oracle Database PL/SQL Packages and Types Referen
Apriori	AR	Association	1
CUR Matrix Decomposition	CUR	Attribute Importance	
Decision Tree	DT	Classification	
Expectation Maximization	EM	Clustering	
Explicit Semantic Analysis	ESA	Feature Ext	raction, Classification
Exponential Smoothing	ESM	Time Series	5
Generalized Linear Model	GLM	Classification, Regression	
<i>k</i> -Means	КM	Clustering	
Minimum Descriptor Length	MDL	Attribute Importance	
Naive Bayes	NB	Classification	
Neural Networks	ИИ	Classification, Regression	
Non-Negative Matrix Factorization	NMF	Feature Extraction	
Orthogonal Partitioning Clustering	O-Cluster	Clustering	
Random Forest	RF	Classification	
Singular Value Decomposition and Principal Component Analysis	SVD and PCA	Feature Extraction	
Support Vector Machine	SVM	Classificatio	on, Regression, Anomaly Detection 42

DBMS_DATA_MINING Algorithms in Oracle*

ALGO_NAME Value	Description	Mining Function	
ALGO_AI_MDL	Minimum Description Length	Attribute Importance	
ALGO_APRIORI_ASSOCIATION_RULE	Apriori	Association Rules	
ALGO_CUR_DECOMPOSITION	CUR Decomposition	Attribute Importance	
ALGO_DECISION_TREE	Decision Tree	Classification	
ALGO_EXPECTATION_MAXIMIZATION	Expectation Maximization	Clustering	
ALGO_EXPLICIT_SEMANTIC_ANALYS	Explicit Semantic Analysis	Feature Extraction	
		Classification	
algo_exponential_smoothing	Exponential Smoothing	Time Series	
ALGO_EXTENSIBLE_LANG	Language used for extensible algorithm	All mining functions supported	
ALGO_GENERALIZED_LINEAR_MODEL	Generalized Linear Model	Classification, Regression; also Feature Selection and Generation	
ALGO_KMEANS	Enhanced k_Means	Clustering	
ALGO_NAIVE_BAYES	Naive Bayes	Classification	
ALGO_NEURAL_NETWORK	Neural Network	Classification	
ALGO_NONNEGATIVE_MATRIX_FACTO R	Non-Negative Matrix Factorization	on Feature Extraction	
ALGO_O_CLUSTER	O-Cluster	Clustering	
ALGO_RANDOM_FOREST	Random Forest Classification		
ALGO_SINGULAR_VALUE_DECOMP	P Singular Value Decomposition Feature Extraction		
ALGO SUPPORT VECTOR MACHINES	Support Vector Machine	Classification and Regression	

These are the **actual** algorithm names that need to be use.

These are values for ALGO_NAME setting

There are <u>many</u> <u>additional settings</u> that go with each Mining Function (i.e. **Confidence, Length, Rules, Aggregates, Cost, Number of Clusters**...etc.)

assification and Regression *Oracle Database PL/SQL Packages and Types Reference

One-Class SVM (ML Anomaly Detection)*



The hypersphere containing the target data having center a and radius R. Objects on the boundary are support vectors, and two objects lie outside the boundary having slack greater than 0.

Support Vector Data Description (SVDD): Find the smallest hypersphere containing all data points (use supervised training to get it)



Linear SVM

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@richniemiec *Wikipedia

Machine Learning connection to ADW/ATP

ЛТР	DB Connection Performance Hub	Gervice Console Scale Up/Down	Stop Actions	
	Autonomous Database Information	Tags		
د		ng	III ADMIN	RICH2 Project [ADMIN_RIC *
A. 100	Example Templates			
Proces	+ Create Notebook			Search b
Cherrole Activity Admin Develo Betteene DB2025	Anomaly Detection This notebook shows how to detect Author: Date Added: 2/13/18 11:16 PM Tags: 'Anomaly Detection' 'Machin ★ 5 Likes 1600 第 70	Association Rules Notebook to show the use of Asso Author: Date Added: 2/13/18 11:156 PM Tags: 'SQL' 'Associations' 'Rules' M * 2 Likes % 772	Attribute Importance Notebook to identify key attributes Author: Date Added: 2/13/18 11:16 PM Tags: 'SQL' Attribute Importance' ' * 2 Likes \$\int_{530}\$\$\$29	Classification Prediction M Example notebook to predict custo Author: Date Added: 2/13/18 11:16 PM Tags: 'Classification' 'Prediction' 'De * 3 Likes
	Clustering This notebook shows how to ident Author: Date Added: 2/13/18 11.16 PM Tags: 'Clustering' 'K-Means' 'Expect ★ 1 Likes 🔍 597 💽 35	My First Notebook Oracle Machine Learning example Author Date Added: 2/13/18 11:16 PM Tags: SQL' Date 'Graph' ★ 4 Likes % 913	Regression This notebook shows how to predic Author: Date Added: 2/13/18 11.16 PM Tags: 'Regression' 'SVM' 'GLM' Log ★ 1 Likes % R17 第-31	Statistical Function Oracle Machine Learning example Author. Date Added: 2/13/18 11:16 PM Tags: "Statistics" 'ANOVA' 'T-test' 'T ★ 2 Likes 0 256 ★ 11

Machine Learning connection to ADW/ATP



Machine Learning connection to ADW/ATP

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= <u>u</u> o m	View Pre	diction_Details that explain why the record was	selected as anomalous	FRANKED II 🗐 🕀
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104360 0	CUST_ID	FIRST_ATTRIBUTE	SECOND_ATTRIBUTE	THIRD_ATTRIBUTE
	100646	"CUST_MARITAL_STATUS" actualValue="Widowed" weight=".226"	"CUST_YEAR_OF_BIRTH" actual/value="1941" weight=".118"	"CUST_CREDIT_LIMIT" actual/value="1500" weight="
	102922	"CUST_MARITAL_STATUS" actualValue="Widowed" weight=".222"	"CUST_YEAR_OF_BIRTH" actualValue="1931" weight=",169"	"CUST_CREDIT_LIMIT" actualValue="1500" weight="
	103441	"CUST_MARITAL_STATUS" actual/value="Widowed" weight=".222"	"CUST_YEAR_OF_BIRTH" actual/value="1941" weight=".117"	"EDUCATION" actual/value="Bach." weight=".076"
	104286	"EDUCATION" actual/value="9th" weight=".165"	"HOUSEHOLD_SIZE" actual/value="4-5" weight=".146"	"CUST_CREDIT_LIMIT" actual/value="1500" weight="

Decision Tree Algorithm (ML Classifier)

A decision tree consists of three types of nodes:^[1]

- 1. Decision nodes typically represented by squares
- 2. Chance nodes typically represented by circles
- 3. End nodes typically represented by triangles



Flow Chart calculates whether to settle a case or not based on costs/probabilities



OML (OAA) Oracle Data Mining SQL Sample (PARTIAL)

dmdtdemo.sql =DBMS_DATA_MINING package - Decision Tree



ORDER BY cust id, S.prediction;

OAA Oracle Data Mining SQL Sample

Starter SQL and PL/SQL Scripts for Learning and Fast-Starts



Data Mining Sample Programs

OAA Oracle Data Mining SQL Sample Programs Directory Listing of the Data Mining Sample Programs ORACLE SQL

dmaidemo.sql dmkmdemo.sql dmsvddemo.sql dmardemo.sql dmnbdemo.sql dmsvodem.sql dmdtdemo.sql

dmsvrdem.sql dmdtxvlddemo.sql dmocdemo.sql dmtxtnmf.sql dmemdemo.sql dmsh.sql dmtxtsvm.sql dmshgrants.sql dmglrdem.sql dmstardemo.sql dmhpdemo.sql dmsvcdem.sql

A Game of Pool (Talent / Luck / Work / Nerve) (Your Current Team Plays Robots built on the Best Past Teams or ML)







Gambling on virtual reality: the online casinos of the future



Invasian Credit: Silpitukit/hon





Machine Learning & AI - Oracle's Built-In Algorithms

Oracle's Machine Learning & Adv. Analytics Algorithms

- CLASSIFICATION – Naive Bayes
- Naïve Bayes
- Logistic Regression (GLM)
- Decision Tree
- Random Forest
- Neural Network
- Support Vector Machine
- Explicit Semantic Analysis

CLUSTERING

- Hierarchical K-Means
- Hierarchical O-Cluster
- Expectation Maximization (EM)

ANOMALY DETECTION

- One-Class SVM

TIME SERIES

ORACLE

- State of the art forecasting using Exponential Smoothing.
- Includes all popular models
 e.g. Holt-Winters with trends,
 seasons, irregularity, missing data

REGRESSION

- Linear Model
- Generalized Linear Model
- Support Vector Machine (SVM)
- Stepwise Linear regression

Minimum Description Length

- Principal Comp Analysis (PCA)

- Unsupervised Pair-wise KL Div

CUR decomposition for row & AI

- Neural Network
- LASSO *

200

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C Spark

TEXT MINING SUPPORT

FEATURE EXTRACTION

🟓 puthon

- Algorithms support text type
- Tokenization and theme extraction
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STATISTICAL FUNCTIONS



 Basic statistics: min, max, median, stdev, t-test, F-test, Pearson's, Chi-Sq, ANOVA, etc.

R PACKAGES



- CRAN R Algorithm Packages
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EXPORTABLE ML MODELS

- REST APIs for deployment

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SQL ANALYTICS

 SQL Windows, SQL Patterns, SQL Aggregates



Naïve Bayes (ML Classifier)

- Probabilistic classifier (*Bayes Theorem*):
 output = prior x likelihood/evidence (simplified)
- Word frequencies for text categorization (reduce spam)
- Also used in Medical diagnosis
- Probability as extension of logic: quantify knowledge shared (supervised learning)
- Probability as extension of logic: Includes personal beliefs.
- Good for supervised learning <u>Is it M / F based on variables</u>: Height/Weight/Shoe Size)



Logistic Regression (ML Classifier)

- Probability of two alternatives based on variables (also 2+)
- Estimating the parameters of logistic model
- Predicts mortality based on Injury score (TRISS)
- Predict customer tendencies of purchases
- Predict voting based on demographics (D or R)
- Predict mortgage default likelihood (0 or 1)
- Predict risk of developing disease
- Example:

Pass/Fail based on Hours Studied - logistic Childs Height changes 2"/yr - linear



Logistic Regression (ML)

Probability of passing exam versus hours of studying



1.00

Pass

Is it a 0 or 1; Log Odds or Logit

Where's line for multiple values (linear regression)





Decision Tree (ML Classifier)

- Tree model for decisions and consequences
- Great for decision making to reach a goal
- Great to calculate odds of different choices
- Can have chance outcomes
- Can calculate costs & "utility" (value)

 Could be used for chatbots (Get monthly payments)



Random Forest (ML Classifier)

- Multitude of Decision Trees (picks the mode)
- Output: Class of decision tree most likely
- *Mode* of the classes is mean or *most likely*
- Mode is found during training time
- Separate the Real Data from the Noise! (divide and conquer)
- Fix decision tree "overfitting" or <u>"too close of fit"</u> with supervised learning training set.
- Miss the outliers if overfitting occurs
- *Bagging* will smooth things out (stability)

richniemiec@gmail.com

viscosityna.com

TowardsDataScience.com



Tally: Six 1s and Three 0s Prediction: 1

@richniemiec

Fixing Random Forest (ML Classifier)





Fixing Random Decision Forest by **avoiding overfitting** and provide stability to the model or graph.



Bootstrap aggregating, also called bagging, is a machine learning ensemble metaalgorithm designed to improve the stability and accuracy of machine learning algorithms used in statistical classification and regression. It also reduces variance and helps to avoid overfitting. Although it is usually

Neural Network (ML Classifier)

- Artificial Neural Networks (ANN) make ML/AI fast
- <u>Deep Learning</u> (Google's **Tensor Flow** leverages this)
- Image Classification with Tensor Flow
- Classify patterns & sequences, regression testing...
- Data Mining, Spam Filtering, Robotics
- Supervised or Unsupervised Learning
- Autonomous Vehicles
- Use for Predictive Modeling
- Used with Speech Recognition



Neural Network (ML Classifier)



Use BP or <u>Backpropagation</u> to update weights to minimize loss; use gradient descent or stochastic gradient descent [differential]. (PyTorch autograd / Dropout so you don't overfit)

NN's also called ANN's /Artificial Neural Networks

neural networks, but artificial neural networks



y = vector/image

W=weight (each line weight/strength)

b = thresholds/vector or bias

y times W is just Vector x Matrix

h = rectified linear unit (>0)

Set: #layers / W's / b's / #neurons / features

Neural Network - CNN (ML Classifier)* *HeliFromFinland





https://towardsdatascience.com/why-deep-learning-isneeded-over-traditional-machine-learning-1b6a99177063

Support-Vector Machine (ML Classifier)

- Used to <u>Classify Data</u> (groups/separates objects)
- They analyze the data used for **regression**
- **<u>SVM</u>** as binary non-probabilistic linear classifier
- Classify images, text, & even hand written text.
- SVM maps training data into separated areas
- <u>Supervised Learning</u>: Uses Algorithms with your data
- Unsupervised Learning: Data clustering of random data





Explicit Semantic Analysis (ML Classifier)

- Used for NLP (Natural Language Processing)
- Used to Categorize Text
- Text represented as a Vector



- String of words as the Centroid of Vectors
- Semantic relatedness based on vectors
- Must use a Knowledge Base (Learning)
- NLP is Very Hard in world of ML (use pre-built)
- Uses weighting factors based on frequency...

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ASSOCIATION RULES

Hierarchical K-Means (ML Cluster)

- Use the K-means (Lloyd's) Algorithm
- Use observations of the Data Set
- Clusters generated based on each observation with nearest mean & <u>hierarchies of clusters</u>.



How Many? Elbow Method (4 in image)!

Centroid cluster is new mean - <u>Repeat</u>



Hierarchical O-Cluster (ML Cluster)

- Orthogonal Partitioning Clustering
- Oracle-Proprietary clustering algorithm
- Algorithm operates recursively
- Creates dense areas in attribute space



- Hierarchical grid-based with Axis-parallel (orthogonal) partitions
- Sensitivity defines baseline density level (to separate peaks/valleys)
- Areas with peak density above baseline are clusters
- Clusters used to generate Bayesian probability
- Outliers can cause issue; pre-clip outliers
- <u>K-Means is Distance Based; O-Cluster is Density Based!</u>

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One-Class SVM (ML Anomaly Detection)

- Support Vector Machine (SVM)
- One-Class Classification (OCC)
- Used to Classify Data
- Supervised Learning using Algorithms
- Identify specific objects of a class based on supervised learning based on objects of that class.
- SVM maps training data into separated areas
- Unsupervised Learning does data clustering
- They analyze the data used for **regression**
- SVM as binary non-probabilistic linear classifier
- Classify images, text, & even hand written text.



Fraud detection Use SVM OCC

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SQL Aggregates

ASSOCIATION RULES

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SQL ANALYTICS

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Seasonal, Irregular & Missing Data: Time Series Algorithm



Week

Exponential Smoothing (Time Series)

- State of the art Forecasting Tool
- Keep the pattern & lose the "noise."
- Instead of Simple Moving Averages (SMA), recent data gets higher weighted.



- SMA all data equal & no smoothing occurs.
- Older data is weighted lower based on smoothing factor.
- As the smoothing factor nears zero, there is a greater smoothing effect on older data.
- A smoothing factor of 1 causes no smoothing.
- Smoothing gets rid of the "noise" in a signal
- Also, Holt-Winters double exponential smoothing

viscosityna.com

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C Spark **FEATURE EXTRACTION**



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PREDICTIVE QUERIES






Linear Model (Regression)

Linear Regression (below):

Using simple **Linear Regression** we find the line between the points. There is one independent and one dependent variable. This helps us get to predictive analytics.



Linear Regression (above): Unemployment goes down & GDP goes up



Generalized Linear Model (GLM)

- GLM: *Flexible* generalization of linear regression
- GLM allows output that doesn't match normal linear regression & generalizes it by a link function based on variance from predicted value.
- GLM for varying <u>situations without Normal</u>
 <u>Distributions</u>
- GLM assumes observations uncorrelated
- GLM assumes random vs. normal effects
- GLM mixed models allow random effects that can be correlated through some probability.



Support-Vector Machine Regression (SVR)

- <u>NOT</u> used to <u>Classify Data</u> (which groups/separates objects), but for regression.
- Both Linear and Non-linear SVM Regression
- The Model is Produced by SVM Classification, but depends on only a subset of training data
- SVM analyzes the data used for **regression**
- Cost function to build model is <u>not concerned with training</u> points that are beyond the margin.
- <u>A margin of tolerance (Epsilon) is used for an</u> approximation (the higher epsilon the less error)
- Supervised Learning uses Algorithms
- Unsupervised Learning uses Clustering



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Minimum Description Length (Attribute Importance)

Use MDL to find the general form of a model and its parameters



- Formalization of Ocam's Razor (keep it simple) by compressing the data (describe things using fewer symbols)
- Hypothesis that includes the main things (MDL) & removes things that are only relatively important
- If methods would lead to poor results, the method should be dismissed.
- <u>Similar to Bayesian inference</u>; Model and data correspond to prior probability & marginal likelihood in Bayesian framework (it's close to Bayes)
- Similar to MML (Minimum Message Length), but MML is subjective Bayesian & MDL is avoids assumptions about Data generating process (there are other differences)

Principal Component Analysis (Attribute Importance)

- PCA: A dimension-reduction tool taking large set of attributes to small set (reduction) with most important information.
- Use Feature Selection of Most Important Attributes
- PCA speeds up your ML reducing dimension input
- "The eigenvectors and eigenvalues of a covariance (or correlation) matrix are the "core" of PCA and...
- The **eigenvectors** (principal components) determine the directions of the new feature space, and the **eigenvalues** determine their magnitude."
- PCA is the simplest eigenvector based analysis (multivariate)

(see next slide for example & later slides for example on PCA Feature Extraction)

Unsupervised Pair-Wise KL Div (Attribute Importance)

- Kullback-Leibler Divergence (KL Divergence)
- A measure of how one Probability Distribution is different from another (the divergence).
- <u>Applications include</u>: time-series randomness, entropy of information, information gain when comparing models
- Also called **Relative Entropy** (decline to disorder)
- Identical KL divergence is zero
- Measurement of *surprise*

• **Example:** Compare NBA Players



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R PACKAGES CRAN R Algorithm Packages

Apriori / Market Based (Association Rules)

- Apriori Algorithm is used for <u>association rule</u> learning databases. <u>Oracle Excels here!</u>
- Identifies <u>frequently associated</u> database items



- Frequent items determine <u>Association Rules (trends)</u>
- Helps with <u>Market Basket analysis (purchase</u> <u>behavior of customers</u> (retail) by using information to discount/promote items
- Also shows items bought together (cheese/crackers)
- The algorithm may scan database many times (downside)
- Finds too many matches (must limit this at times)

A Priori / Market Based (Association Rules)



T4Tutorials: Finding best Items Frequently Bought together.

beer that close to the Bread, Diapers & Milk

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Principal Component Analysis (Feature Extraction)

- PCA Feature extraction reduces large data sets, by building new set of attributes making it faster to process
- **Present** the <u>same information with Fewer Variables</u>
- Ensure you preserve the structure (variance)
- Use Feature Extraction vs. Feature Selection



Visual Feature Extraction Extract features from the face that can be used for the recognition task

- PCA is a dimension-reduction tool taking large set of variables to small set with most important information.
- First component has largest variance and succeeding variables are orthogonal & highest variance.
- PCA uses orthogonal transformation to convert possibly correlated variables (numeric) into linear uncorrelated variables (principle components)
 - New Attributes a Combination of Old Attributes



Singular Value Decomposition (Feature Extraction)

- SVD is the most widely used Matrix decomposition method. Used in Linear Regression, Image Compression, PCA, Least Squares and De-Noising Data (take the noise out)
- **SVD** is factorization (**decomposition**) of a Matrix.
- The key is to get the constituent elements from the Original Matrix (through an iterative process)
- SVD can find singular vectors & values from the Original Matrix
- SVD lowers the amount of features (<u>feature reduction</u>) of the large original matrix (a type of <u>feature extraction</u>)
- Used in widely in Statistics to see more than the original matrix is showing
- Many books on SVD using Python (Jason Brownlee)



Singular Value Decomposition (Feature Extraction)



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SQL ANALYTICS SQL Windows, SQL Patterns,

ASSOCIATION RULES



 OAA (Oracle Data Mining + Oracle R Enterprise) and ORAAH combined OAA includes support for Partitioned Models, Transactional, Unstructured, Geo-spatial, Graph data, etc. OAA is Oracle Advanced Analytics; ORAAH is Oracle R Advanced Analytics for Hadoop





SQL Analytics (Windows / Patterns / <u>Aggregates</u>)*



Product Mgr. View



Financial Mgr. View





Regional Mgr. View



Ad Hoc View

Cubes show many <u>dimensions for</u> <u>various users</u> of the system.

In-Memory will use Vector Group By Aggregation



Statistical Functions in Oracle (Partial List)*

Numeric Functions	Analytic Functions		Aggregate Functions		Data Mining Functions
ABS ACOS ASIN ATAN ATAN ATAN2 BITAND CEIL COS COSH EXP FLOOR LN LOG MOD NANVL POWER REMAINDER ROUND (number) SIGN SIN SINH SQRT TAN TANH TANH TRUNC (number) WIDTH_BUCKET	AVG * CLUSTER_DETAILS CLUSTER_DISTANCE CLUSTER_ID CLUSTER_PROBABILITY CLUSTER_SET CORR * COUNT * COVAR_POP * COVAR_SAMP * CUME_DIST DENSE_RANK FEATURE_DETAILS FEATURE_DETAILS FEATURE_SET FEATURE_SET FEATURE_SET FEATURE_VALUE FIRST FIRST_VALUE * LAG LAST LAST_VALUE * LEAD LISTAGG	MAX * MIN * NTH_VALUE * NTILE PERCENT_RANK PERCENTILE_CONT PERCENTILE_DISC PREDICTION PREDICTION_COST PREDICTION_DETAILS PREDICTION_PROBABILITY PREDICTION_PROBABILITY PREDICTION_SET RANK RATIO_TO_REPORT REGR_ (Linear Regression) ROW_NUMBER STDDEV * STDDEV_SAMP * SUM * VAR_POP * VAR_SAMP * VARIANCE *	APPROX_COUNT APPROX_COUNT_DISTINCT APPROX_COUNT_DISTINCT_AGG APPROX_COUNT_DISTINCT_DETAIL APPROX_MEDIAN APPROX_PERCENTILE_AGG APPROX_PERCENTILE_DETAIL APPROX_RANK APPROX_SUM AVG COLLECT CORR CORR* COUNT COVAR_POP COVAR_SAMP CUME_DIST DENSE_RANK FIRST GROUP_ID GROUPING GROUPING GROUPING_JD JSON_ARRAYAGG JSON_OBJECTAGG LAST LISTAGG MAX	MEDIAN MIN PERCENT_RANK PERCENTILE_CONT PERCENTILE_DISC RANK REGR_(Linear Regression) Function STATS_BINOMIAL_TEST STATS_CROSSTAB STATS_F_TEST STATS_KS_TEST STATS_MODE STATS_MW_TEST STATS_ONE_WAY_ANOVA STATS_T_TEST_* STATS_ONE_WAY_ANOVA STATS_T_TEST_* STATS_WSR_TEST STDDEV_SAMP SUM SYS_OP_ZONE_ID SYS_XMLAGG TO_APPROX_PERCENTILE VAR_POP VAR_SAMP VAR_SAMP VAR_SAMP VARIANCE <u>Also:</u> T-Test, XMLAGG (Analysis of * Oracle Datab	CLUSTER_DETAILS CLUSTER_DISTANCE CLUSTER_ID CLUSTER_PROBABILITY CLUSTER_SET FEATURE_COMPARE FEATURE_DETAILS FEATURE_ID FEATURE_SET FEATURE_VALUE ORA_DM_PARTITION_NAME PREDICTION_BOUNDS PREDICTION_BOUNDS PREDICTION_COST PREDICTION_DETAILS PREDICTION_PROBABILITY PREDICTION_SET , F-Test, Chi-Squared, ANOVA Variance), Pearson's Test
				UIALIE DALAD	ase such Language Neithelitte

unctions

Machine Learning Functions* - Oracle Docs.

- <u>First</u>: Clear Business Problem to Solve
- <u>Second</u>: *Function* to Perform
- Third: Algorithm to use
- <u>First</u>: Build/*Train the Model* when you build it using about 60% of data.
- <u>Second</u>: *Test/Score Model* for accuracy/precision using about 40% of data.

Compare Algorithms!



Oracle Example: Business Issue, Build & Score Model

```
-- PL/SQL procedure to Build a Decision Tree
```

-- classification model to predict customer

```
-- acceptance of offer
```

BEGIN DBMS_DATA_MINING.DROP_MODEL('PRED_AFFINITY_DT'); EXCEPTION WHEN OTHERS THEN NULL; END; / DECLARE

v_set1st DBMS_DATA_MINING.SETTING_LIST;

```
BEGIN
v_set1st('PREP_AUTO') := 'ON';
v_set1st('ALGO_NAME') := 'ALGO_DECISION_TREE';
```

```
DBMS_DATA_MINING.CREATE_MODEL2(
    'PRED_AFFINITY_DT',
    'CLASSIFICATION',
    'SELECT * FROM MINING_DATA_BUILD',
    v_setlst,
    'CUST_ID',
    'AFFINITY_CARD');
```

-- SQL to dynamically score and return all

- -- customers with likelihood > 50% to be
- -- AFFINITY_CARD responders from MINING_DATA_APPL

```
SELECT * FROM (
   SELECT CUST_ID,
    PREDICTION_PROBABILITY(PRED_AFFINITY_DT, '1'
        USING A.*) PROBABILITY
   FROM MINING_DATA_APPLY A)
WHERE PROBABILITY > 0.5
ORDER BY PROBABILITY DESC;
```



Time Enough at Last for ML with Autonomous DB



AutoML for Autonomous Database- 3/18/2021

ORACLE' Machine Learning



AutoML is here for Autonomous Database





Exadata Cloud Machine: ALL Features (Oracle's Juan Loaiza presentation on Exadata Cloud)

Exadata Cloud: Compatible, Scalable, Available, Secure Decades of Database Innovation Proven at Millions of Mission-Critical Deployments



Oracle Machine Learning Algorithms and Analytics in Oracle Database

CLASSIFICATION



- Naïve Bayes
- Logistic Regression (GLM)
- Decision Tree
- Random Forest
- Neural Network
- Support Vector Machine (SVM)
- Explicit Semantic Analysis
- XGBoost^{*}

ANOMALY DETECTION

- One-Class SVM
- MSET-SPRT^{*}

CLUSTERING

- Hierarchical K-Means
- Hierarchical O-Cluster
- Expectation Maximization (EM)

TIME SERIES MAN

- Forecasting Exponential Smoothing
- Includes popular models e.g. Holt-Winters with trends, seasonality, irregular time series

OML Algorithm Cheat Sheet Algorithm Documentation



- Generalized Linear Model (GLM)
- Support Vector Machine (SVM)
- Stepwise Linear regression
- Neural Network
- XGBoost*

ATTRIBUTE IMPORTANCE

- Minimum Description Length
- Random Forest
- Unsupervised Pairwise KL Divergence
- CUR decomposition for row & AI

ASSOCIATION RULES

SOL ANALYTICS

- SOL Windows SOL Patterns
- SQL Aggregates

outhon



 R^{n}



AutoML

FEATURE EXTRACTION

- Principal Comp Analysis (PCA)
- Non-negative Matrix Factorization
- Singular Value Decomposition (SVD)
- Explicit Semantic Analysis (ESA)

ROW IMPORTANCE

CUR Decomposition

RANKING

XGBoost



TEXT MINING SUPPORT

- Algorithms support text columns
- Tokenization and theme extraction
- Explicit Semantic Analysis (ESA)

STATISTICAL FUNCTIONS

 min, max, median, stdev, t-test, F-test, Pearson's, Chi-Sq, ANOVA, etc.

Includes support for partitioned models, integrated text mining, automated data preparation





 \bigcirc





Types of ML Algorithms in Healthcare on PubMed*



*Stroke and vascular neurology (snj.bmj.com)

<u>Note</u>: Markov is Bayesian, NN is k-NN, DA close to PCA

Machine Learning has many parts



From Agnes with Love (Computers try to help)



Biju Thomas at ODTUG - *Emerging* Jobs

#1 AI / ML Specialist (Engineer)

Linkedin

2020 Emerging Jobs Report

- 74% annual growth
- Skills
 - Machine Learning
 - Deep Learning
 - TensorFlow
 - Python
 - Natural Language Processing
- Industries
 - Computer Software
 - Information Technology
 - Higher Education
 - Consumer Electronics

- Take the data scientists' code and making it more effective and scalable.
- Introduce various programming rules and good practices.
- Join the results from potentially unrelated tasks to enhance the models performance even more.
- Glue all the "data science" and "software" parts together.

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Python is #1 Language for Machine Learning*



Oracle integrates Python with OAA



- Similar architecture to OAA's Oracle R Enterprise
- OML4Py Transparency Layer
 - Use Oracle Database as High Performance Computing environment
- OML4Py OAA Model Build and Apply
 - Use OAA parallel and distributed ML algorithms
 - Manage Python scripts and Python objects in Oracle Database
- OML4Py Embedded Python
 - Make callout to Python packages
 - Integrate Python results into applications via SQL



George Mines ADW with Oracle Analytics Cloud (OAC)

OAC: Data Visualization Cloud Services (DVCS), Bus. Intell. CS (BICS), Essbase, Smart View (EE/SE)





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www.viscosityna.com

Oracle Analytics Cloud (OAC) to Cluster Data



ML & Business Apps - Today at 4:15; San Antonio 1



Better World – Leverage Tech!

Steve Wozniak on AI: Will we be pets or mere ants to be squashed our robot overlords?

Apple co-founder Steve Wozniak predicts a 'scary' future when artificial intelligence takes


Robots We Grew up With...







Robots Now... Closer to the Future



Actual Robots Now... Beyond Science Fiction



Living Doll

















These dolls are spying on your kids, consumer groups say



Published 3:46 PM EST, This Docomber 9, 2016







"Privacy advocates try to keep 'creepy,' 'eavesdropping' Hello Barble from hitting shelves," Sarah Halzack, Washington Post

Mirror Image













Mirror Image





Prof. Hiroshi Ishiguro, right, inspects his robotic twin, Geminoid HI-1, also known as 'Gemmy,' who handles his teaching duties at Osaka University in Japan. Photo courtesy of ATR Intelligent Robotics and Communication Laboratories.











Fun with Machine Learning! (@CitizenPlain)

MY MODERN MET

PERMANNIA ALL'INFIDENCE CONTRACTOR A handler

THE NET AND DRAFT AND T

CONTRACT - MARCHINA - O Designation -

de Constantia





INTY MADE PERSON NAMES

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The Lonely (Robot Companionship)



Leveraging – DB & Robotics!



Pepper the Robot as Host at Sushi Restaurant (SynchroJapan) Leverage Robotics & Database



Rich Niemiec @RichNiemiec · Apr 3

The #Future is here! This #Robot at @synchrojapan in charge of reservations & seating at #sushi #restaurant in #Japan. Spend time on #Innovation, #BigData, #BigDataAnalytics, #IOT, #MachineLearning, #DeepLearning, #AI with the #database #oracledatabase

🚯 SynchroJapan • Food & Restaurant Culture in Japan



🚯 SynchroJapan • Food & Restaurant Culture in Japan



Use Oracle Virtual Assistant with Robots



Oracle Virtual Assistant Interface

DRALLE	R Accountance					
4ow		updated every 30 secs	Intents		Upr	lated every 30 secs
2	3 innier	12% 🔺	348 Total	332 Published		
Committees	4 American Exchanges			16 Not Published		
Den Ween, * April 8: 2018 for April 14, 2018					125 Conversations	405 66 Exchanges Issues
where the second are you how do i reset my password how old are you how do i reset my password how old are you how do i reset my password where do i find the heart vide agent where do i find the heart vide but yes my coffice maker makes weak esp coll you brattos		Women haves With booes	Most Popular Intents O	Sort By Total	Without Issues	with Issues
		Prefer Can I talk to an agent? I need to make my petronord My coffee makes makes week way. I'm seeing error metroge ma I'm still seeing the same error me. When can I load the macros s too.				
nues Q						

Twilight Zone that's not here yet... fully



















Five Characters in Search of an Exit



A World of Difference (Getting Closer)



Leveraging – DB, Al & Virtual Reality!



V_{irtual} **R**eality



- > Immersion in virtual worlds
- > Total interaction with virtual
- > E.g. Oculus Rift

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Mixed Reality



- > Virtual World integrated to reality
- > Interaction between reality and virtual
- > E.g. Microsoft HoloLens



Augmented Reality



- > Virtual on top of reality
- > Limited interaction with the virtual
- > E.g. Smartphones & tablets

actimage

digital intelligence 124

Elegy (Build Archive to Communicate with)





Elegy (Future?)

Funeral parlour looking for 'volunteers' for 'Black Mirror' project to create immortal AI

SCIENTISTS are looking for volunteers for a macabre experiment which will allow deceased family and friends to live forever as robots in an eerie scenario similar to Netflix's Black Mirror, it has been revealed.

By MATT DRAKE 23:43 Sat. Feb 24, 2018 | UPDATED: 01:53; Sun. Feb 25, 2018





Lifestyle / Family & Relationships

'It's heartbreaking': how coronavirus pandemic is changing funerals, grieving and ways we say goodbye to loved ones



Pepper Now Available at Funerals as a More Affordable Alternative to Human Priests > In Japan, thrifty Buddhists can now hire the Pepper robot as a funeral assistant





In His Image (Available Now - 100K Waiting List)







ETER9

The most daring experience ever made on the Internet.

Let your counterpart work for you. The very first-ever intelligent virtual self of you.

STOCKELS YOURSEL

Connecting, was have inter-1985;XXXI registered states when one and the weaking lief. For \$15339 to aircfulld delaydicates, for a part of 1510 mean table phase tool:

CBS Aol. EDE The Celegraph theguardian Sun





Artificial Intelligence

ETERS is a social network that relies an Artificial Initiallyseca as a cancel of element, and it's currently in the IRETA stage. Even in your absence, the virtual beings will publish, comment and interact with you initializently.



Counterpart

The Counterpart is your Virtual Self their will stay in the system and interacts with the work/gates like you worked if you energy present. Your Counterpart will learn more with each action you take. The more your interact in the new social network, the more your Counterpart will learn



Cyber Eternity

Elemiting is a way of keeping your thoughts and posts for all time. Are you surlocal/tame meets your Counterpart and become etermal. Challenge the impressible.



JOB BOARD

Search



AI GUEST

It's time for workers to worry about AI

GARY GROSSMAN, EDELMAN @GARYG02 APRIL 7, 2019 2:22 PM



MOST READ



It's time for workers to worry about AI



Future Tech is in AI:

You can Leverage this NOW!







The After Hours (Future Sentience Issues Ahead)



The Digital Transformation Ahead

Digital Transformation 2000 to 2050 A historically significant change in humanity...





Wearing Digital Implanting Digital The Hive Mind

Gerd

The Future of Implant Technology



Connection to the Brain – Currently Working



C Johns Hopkins University

Walking Distance (Create your past to view)









Developer Creates VR Time Machine To Relive His Past Year

uly 13. 2020 - by finishity Carillan







A Nice Place to Visit (Mystery Solve / Mean Afterlife VRs)

















Leveraging – DB & Robotics!



Blog: Science

Mini Insect Robocops: Engineers, Government Work on RC Insects

Jaron Midi (blog) - October 9, 2007 10:02 AM

The Fifth Trumpet

...⁹They also had thoraxes like breastplates of iron, and the sound of their wings was like the roar of many horses and chariots rushing into battle. ¹⁰They had tails with stingers like scorpions, which had the power to injure people for five months. ¹¹They were ruled by

Experimental Vaccine Daily Blog





Third Planet from the Sun (Elon Musk)











SPACEX







Twilight Zone - Maybe you can build it...







Digital – How did we go from Magical to Toxic?





Black Mirror - Dystopian

• Nosedive: (Social Meeting Addiction & Ratings)

• Hated in the Nation (Cancel Culture)



WIRE Our World is Going Full Black Mirror'

Nosedive (season 3, episode 1)





• Be Right Back (ETER9)



- The Entire History of You
 (Kapture / Google Glass / Google Clips)
- Metalhead (Robot Dog)





A Thing About Machines - People Frustrated













AVE U TRIED TURNING IT O AND ON AGAIN⁹







Oracle Database 12c 12: 12: Release 2 Performance Tuning Tips and Techniques Fully Encrypted Database Tablespace Encryption (TDE

18c

Oracle

Database

ORACLE

Database

Oracle Database Security Built over MANY years...

Encryption in the Silicon (M7) Oracle Multi-Tenant Security (PDBs)

19

Oracle Audit Vault

Oracle Database Vault

DB Security Evaluation #19

Transparent Data Encryption

EM Configuration Scanning

Fine Grained Auditing (9i)

Secure application roles Client Identifier / Identity propagation

Oracle Label Security (2000)

Proxy authentication

Enterprise User Security

Global roles

Virtual Private Database (8i)

Database Encryption API

Strong authentication (PKI, Kerberos, RADIUS)

Native Network Encryption (Oracle7)

Database Auditing Government customer

1977

<u>18c:</u> User-defined <u>Master</u> <u>Encryption Key</u> (bring own key - keystores)

<u>19c</u>: Oracle DB supplied <u>schema-only accounts</u> <u>have passwords removed</u> (not sample accounts)

21°

Global Fault-Tolerant Key Vault & Oracle Data Safe

2022 +

Gartner Hype Cycle August 2018 All about Tech Creating a New Reality



^{© 2018} Gartner, Inc.


🔿 less than 2 years 🕥 2 to 5 years 🌑 5 to 10 years 🛕 more than 10 years 🛞 obsolete before plateau

What's comes *after* the Exadata Zone? YOU will soon be in for more...

	Directly Addressable	Indirect
4 Bit:	16	(640)
8 Bit:	256	(65,536
16 Bit:	65,536	(1,048,5
32 Bit:	4,294,967,296	
64 Bit:	18,446,744,073,709,551,616	
128 Bit:	3.4 x 10e+38 quantum leap!	

Indirect/Extended (640) (65,536) (1,048,576)



- Qubits allow multiple states so that you can look at all of the possibilities/probabilities at one time (IBM Q is 20 qubits).
- The "Quantum Zone" next (Quantum Physics is incomplete Einstein)
 - Just 512 qubits would store 512-bits of addressable memory or 2⁵¹² (which is well over a googol or 1 with 100 zero's after it a googol is about 2³³²).
 - Brush up on your Eigenvectors, Eigenvalues, Pauli Matrices & Grover's Algorithm
 - Create Singularity ... all atoms of a person by 2045 (I think earlier); 12-Monkeys
 - Private universes Is there one for each person? (Schroeder's cat I think not)
 - Rearranging atoms to create new objects; Nanotech + Quantum Physics coming!

The Age of Entanglement!

"A masterful account of the phenomenon Einstein thought so crazy it could not possibly be true. (Only it is!)" -- DR. MARCUS CHOWN, author of The Universe Next Door

THE GOD EFFECT

QUANTUM ENTANGLEMENT, SCIENCE'S STRANGEST PHENOMENON

> TIME TRAVEL, TELEPORTATION, AND THE ULTIMATE COMPUTER

BRIAN CLEGG



We are now shifting from the Information Age to:

The Age of Entanglement



Gartner 2021 hype cycle for Emerging Tech

Hype Cycle for Emerging Technologies, 2021



3 Types of Artificial Intelligence

Artificial Narrow Intelligence (ANI)



Stage-1

Machine Learning

 Specialises in one area and solves one problem





Artificial General Intelligence (AGI)



Stage-2

Machine Intelligence

 Refers to a computer that is as smart as a human across the board

Artificial Super Intelligence (ASI)



Stage-3

Machine Consciousness

 An intellect that is much smarter than the best human brains in practically every field

Final Thoughts... world changing fast!

"Those who use things of the world should not become attached to them. For the world in its present form is passing away."

1 Corinthians 7:31



Star Trek

- Communicator Motorola Flip Phone
- Phaser EPM or Stun Gun / Taser
- Tablets (Medical) Tablet Computers
- Tricorders Many Medical Devices (below)
- Translators Google Translate (others)
- Tractor Beam MIT has it in concept
- **Telepresence** Zoom Virtual Conferences
- Geordi's Visor Robotic Eyes / Implants
- Communicator Badges Many Security Badges
- Food Replicator 3D Printer
- Holodeck VR
- **Teleportation Quantum Entanglement**
- Big Screen TV Everyone has it
- Hands Free Phone Bluetooth headset / Airpods

















Hololens VR Apps Above



Microsoft working on Holodeck





The Digital Transformation Ahead

Digital Transformation 2000 to 2050 A historically significant change in humanity...







Wearing Digital

Implanting Digital

The Hive Mind

Summary – Goal is Apply Technology & W.I.N.

- □ You've Just Entered the Twilight Zone (20+)
- □ Twilight Zone We See Now
- □ Big Data and IOT Data Is Coming Fast
- □ ML & Oracle Overview & ADB
- □ Applications of ML Algorithms & AutoML
- □ Machine Learning Future, Robots & VR

□ Future Tech from the Twilight Zone+

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"Spooky action-at-a-distance" (A. Einstein)

Final Thoughts... Catch your Ride!



"Things may come to those who wait, but only the things left by those who hustle." — Abraham Lincoln





Oracle never caught from behind Oracle's 45th Anniversary in 2022

- Great Sales/Marketing
- Great Database
- Applications Leader
- BI Leader
- In the lead except Cloud
- GAME OVER
- Hardware/Software Engineering!
- Have Everything to Win in Cloud + Al!

In Memory of Ken Jacobs, Dr. DBA & Joel Kallman



Chief Innovation Officer, Viscosity North America 2a - O Dr. DBA, Ken Jacobs, He will be missed!! Oracle's b

chard Niemiec

Dr. DBA. Ken Jacobs. He will be missed!! Oracle's best advocate of User Groups: Kind, Gwing, Positive, Honest, Humble & Smartest Guy in the Room? Sadly, Ken suffered a pulmonary embolism and passed away last Friday at the age c __see more.

Who is the Dr. pictured below?



O 🗣 🖸 104 - 43 comments

12c R2 Book – Available Now!

3an 2022	0ec 2021	Jan 2021	DBMS	Database Model	Jan 2022	0ec 2021	Jan 2021
1.	1.	1	Oracle 😃	Relational, Hulti-model 🐻	1266.89	-14.85	-56.05
2.	- 2.	2.	MySQL 📫	Relational, Multi-model 👩	1206.05	+0.01	-46.01
3,	э,	3.	Microsoft SQL Server	Relational, Multi-model 👩	944.81	-9,21	-56.42
4.	4.	4.	PostgreSQL 😆 👰	Relational, Multi-model 😈	606.56	-1.86	+54.33
5.	5.	3.	MongoDB C	Document, Multi-model 👔	488.57	+3.89	+31.34
ő.	6.	* 7.	Redis 📫	Key-value, Multi-model	177.98	++.44	+22.97
7.	- 3.	46.	18M Db2	Relational, Multi-model 👩	164.20	-2,98	+7.03
В.	8.	8.	Elasticsearch	Search engine, Hutti-model 🗃	160.75	+3.03	+9.50
9.	1 11.	₱11.	Microsoft Access	Relational	128.95	+2.96	+13.61
10.	4 %	49.	SQLite C	Relational	127.43	-1.25	+5.54
11.	11.	4 10.	Cassandra 🛄	Wide column	123.55	+4.35	+5.47
\$2.	12.	12.	MariaDib 😋	Relational, Multi-model 🛅	106.42	+2.06	+12.63
13.	13.	13	Splunk	Search engine	90.45	-1.87	+2.79
14.	14.	1 15.	Microsoft Azure SQL Database	Relational, Multi-model 👩	86.32	+3.87	+14.96
15.	15.	1 6.	Hive 👩	Relational	83.45	+1.52	+13.02
16.	16.	4 .17.	Amazon DynamoDB 🗧	Multi-madel	79.85	+2.23	+10.72
17.	17.	4 37.	Snowflake	Relational	76.82	+5.79	+61.30
18.	18.	4 14.	Teradata 😆	Relational, Multi-model 🚺	69.13	-1.17	-3.46
19.	† 20.	4 20.	Solr	Search engine, Hulti-model 👩	58.53	+0.80	+6.04
20.	4 19.	4 19.	Neo4) 🚦	Grieh	58.03	0.00	+4.25
21.	21.	21.	SAP HANA 😆	Relational, Multi-model 👸	56.92	+2.34	+6.85
22.	22.	22.	FileMaker	Relational	55.86	+1.99	+11.47
23.	23.	4 18.	SAP Adaptive Server	Relational, Hulti-model 👩	51.05	-0.33	-3.56
24.	24.	24.	Google BigQuery	Relational	45.62	-0.18	+9.62
25.	25,	4 23.	HBase 🗰 🖗	Wide column	43.99	-1.55	-2.29
26.	26.	4 25.	Microsoft Azure Cosmos DB 👩	Multi-model 🖬	40.04	+0.33	+7.07
27.	27.		PostGIS	Spatial DBMS, Multi-model 👩	31.87	-0.57	
28.	1 29.	\$ 27.	InfluxDB 🖸	Time Series. Hulti-model 🚺	30.09	+1.70	+3.77
29,	\$ 29.	4 26.	Couchbese 😋	Document, Hulb-model 👔	28.86	+0,42	-2.77
30.	30.	30.	Firebird	Aeletional	27.28	-0.31	+4.52
31.	♠ 32.	4 79.	Amazon Redshift	Belatunal	25.85	11.48	72.93

Top New Release amazon niemiec All -Try Prime Your Amazon.co Departments -Books Advanced Search New Releases Books > Computers & Technology > Databases & Big Oracle Database 12c by Richard Niemiec (Author) #1 New Release (in Oracle Databases ORACLE Oracle Database 12c CRACLE 12" **Release 2 Performance** Tuning Tips and Techniques

Best Practices for Optimizing Database Performance

Calendar Training Tools Blog Support Login



FREE Membership

Join Thousands Of Happy Customers And End The Frustration In Tuning Oracle Databases

Our services give you the training you need so your Oracle database runs faster and more efficiently. And, you get the credit!



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> Follow us on LinkedIn, Facebook or Twitter: @viscosityna

Email me at katie.barnes@viscosityna.com



Objectives.

Rich Mierniec Charlonnachar (Phan, Chara A'd Dengar

-Show important features to leverage in 21c from other versions.

-Show examples of features

-Show new features of 21c.

References

- The Emerging Technology Roadmap, Scott Klososky
- Futurist Gerd Leonhard The Futures Agency, Technology vs. Humanity, Gerd Leonhard, oracle.com & Juan Loaiza / Doug Hood presentations, amazon.com, smartcitiescouncil.com, youtube.com, business coach, libelium.com, monetate, en.wikipedia.org, Netflix, Black Mirror, hometoys.com, FPOV, huffingtonpost.com, thegardian.com, nationalgeographic.com, newscientist.com, enswmu.blogspot.com, dailymail.co.uk, FutureRobot, theguardian.com, thinkhealthwireless.blogspot.com, ge.com, cmswire.com, runningsupplement.co.uk, quickmeme.com, nike.com, thisiswhyimbroke.com, businessinsider.com, slideshare.com, forrester.com, spiceworks.com, mwaintel.com, humancapitalist.com, wired.com and Wired Magazine, shodanhq.com, developer.nokia.com, extremetech.com, Getty Images, Dr. Quantum, Leo Kouwenhoven, Telecom Tech News, Actimage & any other company products are the property of their respective companies.







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- References include Rich Niemiec's Exadata Presentation & Oracle 12cR2
 Database Performance Tuning Tips & Techniques book, <u>www.oracle.com</u>, en.wikipedia.org, slashgear.com, gifsoup.com, <u>www.amazon.com</u>, Tech
 Crunch, <u>www.rolta.com</u>, The Twilight Zone, Information Week, Gartner,
 Computerworld, Quest, Data and Technology Today, Forbes, Quest,
 Orbit, Computer Weekly, Redmond, Database Trends & Applications,
 dsp, Dataversity, zdnet, DBVisit, Steve Jones, Kerry Osborne, Julian
 Dontcheff, Accenture, Quora, Brent Ozar, & Oracle OpenWorld



Getting Started—Oracle ML/AI Resources

ORAC	LE.	Oracle Advanced Analytics Overview Information						
	•	Oracle Machine Learning Newest Features and Road Map.pptx presentation						
	•	Blog post: Simple Guide to Oracle's Machine Learning and Advanced Analytics						
	•	Oracle Advanced Analytics Public Customer References						
	•	Oracle's Machine Learning and Advanced Analytics Data Management Platforms white paper on OTN						
-	÷							
. Þ.		YouTube recorded Oracle Advanced Analytics Presentations and Demos,	White Papers					
	 Oracle's Machine Learning & Advanced Analytics 12.2 & Oracle Data Miner 4.2 New Features YouTube video 							
	•	Library of YouTube Movies on Oracle Advanced Analytics, Data Mining, Machine Learning (7+ "live" Demos						
		e.g. Oracle Data Miner 4.0 New Features, Retail, Fraud, Loyalty, Overview, etc.)					
ORAC	LE	Overview YouTube video of Oracle's Advanced Analytics and Machine Learning						
UNIVERSI	TY	Getting Started/Training/Tutorials						
•	OA	A/Oracle Data Miner Workflow GUI Online (free) Tutorial Series on OTN						
•	<u>O</u> A	A/Oracle R Enterprise (free) Tutorial Series on OTN						
•	Trv	the Oracle Cloud Now!	four Tube Search					
•	Ge	tting Started w/ ODM blog entry						
•	Ne	w OAA/Oracle Data Mining 2-Day Instructor Led Oracle University course	N					
	Ora	acle Data Mining Sample Code Examples	Oracle's Machine Learning					
ORACI	LEHe		and Advanced Analytics					
		Additional Resources, Documentation & OIN Discussion Foru	Oracle Data Miner 4 2 New Features					
	•	Oracle Advanced Analytics Option on OTN page	oracie pata miner 412 new reatores					
	•	OAA/Oracle Data Mining on OTN page, ODM Documentation	Accelerate Your					
	•	OAA/Oracle R Enterprise page on OTN page, ORE Docume	Schwert Ander Mangement					
	•	Oracle SQL based Basic Statistical functions on OTN	Phartie berger (Durink Law)					
	•	Oracle R Advanced Analytics for Hadoop (ORAAH) on OTN	X 1 M 1 4 0 007/412 00 0 0					

Analytics and Data Summit, All Analytics, All Data, No Nonsense.

March 12-14 2010 Redwood Shores CA

Quick FREE notes

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