Project0

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A. General Purpose:

| Sys | tem | Pros | Cons |
|-----|------------------------------------|--|--|
| 1 | Lotus Notes: ¹ | Better security control than its competitor Supports more platforms than its competitor Support hardware as well as software virtualization | It's not very flexible in term of using because it's a little bit too secure |
| 2 | Actian Versant ² | It's Object Oriented! It's simple and straight forward to build complex data models which are hierarchical. No need to use a query language No primary keys Make developer's life easier by allowing database to process complex objects models without writing mappings. Allows you to change database schema while your service is on-line. Good Scalability Full support for transactions, logging and locking. | Lack of solid theoretical basis. API is language dependent Schema change may need a system-wide compile(not in Versant's case) |
| 3 | InterSystems Cache ³ | A fully persistent database with high throughput even comparing with in-memory database Good scalability and performance when hosted | • |

| | | on inexpensive machines. |
|----|----------------------------|--|
| 4 | McObject: | Memory data base with good scalability. ACID-Compliant It's Object Oriented! It's simple and straight forward to build complex data models which are hierarchical. No need to use a query language No primary keys Lack of solid theoretical basis. API is language dependent Schema change may need a system-wide compile(not in Versant's case) |
| 5 | ObjectStore: | It's Object Oriented! It's simple and straight forward to build complex data models which are hierarchical. No need to use a query language It's Object Oriented! It's Lack of solid theoretical basis. API is language dependent Schema change may need a system-wide compile(not in Versant's case) No need to use a query language No primary keys |
| 6 | WakandaDB: ⁴ | Everything(schema, server-side processing, querying) can all be done in JavaScript Open Source and not mature |
| 7 | IBM IMS: ⁵ | It's a full function database Optimized for high transection rates. High Availability |
| 8 | Adabas ⁶ | Is able to work close with previously existing database system No access control in term of native network encryption |
| 9 | UniVerse: ⁷⁸ | Multi-valued database High Availability Good Scalability Intuitive database design High Performance Does not constrain by 1st Normal Form The fact that it does not adhere to 1FN can be abused |
| 10 | UniData Documentation xDB | Secure Similar to UniVerse XML-based Database Allows the schema to be Similar to UniVerse Not ACID-compliant |

| | | • | Flexible schema compared | | |
|----|------------------------|--------|--------------------------------------|---|--|
| | | | to relational database | | |
| | | | | | |
| | | • | EMC^2's disaster-recovery | | |
| | | | options | | |
| | Tamino XML | * | XML-based Database | • | Not ACID-compliant like many |
| | Server | | advantage described in | | other document based |
| | | | Documentation xDB | | database |
| 13 | Ipedo XML | • | XML-based Database | • | Not ACID-compliant like many |
| | Database | | advantage described in | | other document based |
| | | | Documentation xDB | | database |
| 14 | OrientDB:910 | • | Document oriented | • | Not Mature, API changes over |
| | | | database with graph | | time. |
| | | | database feature | | |
| | | • | Open Source | | |
| | | • | Can be queried using SQL | | |
| | | • | AICD | | |
| 15 | SQLite | • | Easy to use | • | Doesn't scale very well |
| | | • | Consumes less resources | | · |
| 16 | Firebird ¹¹ | • | Free open source | • | Not very scalable in term of |
| | | • | Relational database | | horizontal scaling |
| | | • | Well established and | • | Difficult to model complex |
| | | | tested based on solid | | data model because it is table |
| | | | theoretical foundation | | based |
| | | | ACID | | Sasca |
| | | • | SQL as access language | | |
| | | | Join! | | |
| | | | Large Throughput | | |
| 17 | SAP Sybase | • | Relational database | • | Relational database |
| 17 | ASE: ¹² | • | advantages as described in | ľ | |
| | ASE. | | Firebird | | disadvantages described in Firebird |
| 10 | CAD COL | _ | | _ | |
| | SAP SQL | • | Relational database | * | Relational database |
| | Anywhere | | advantages as described in | | disadvantages described in |
| | | | Firebird | | Firebird |
| | | • | Embed: consume less | | |
| | | | resources. | | |
| | | • | | | |
| 19 | Postgres-XL: | • | ACID | * | Too complex for simple stuff |
| | | * | Open Source | | |
| | | * | Cluster-wide Consistency | | |
| | | * | Secure | | |
| | | • | SQL | | |
| | | | | | |
| | | • | Horizontal scaling | | |
| | | • • | Horizontal scaling Rich feature set. | | |

| 21 | MySQL | • | Relational database advantages as described in Firebird | ٠ | Not very Scalable |
|----|-----------------|---|---|---|--|
| 22 | Oracle Database | • | Relational database advantages as described in Firebird Commercial | * | Price Relational database disadvantages as described in Firebird |
| 23 | IBM DB2 | • | Relational database advantages as described in Firebird Commercial | • | Price Relational database disadvantages as described in Firebird |

B. Specialist analytic

| System | Pros | Cons | |
|----------------------------|---|---|--|
| 25. Google BigQuery | Allows users to use SQL-like queries to query massive datasets Rest API Google's infrastructure makes the operation fast and economic. Makes ad-hoc and trial-and-error interactive query on large dataset possible. | Not able to do complex data processing. You can't update your data, only appending is allowed. No large Join. | |
| 26. InfluxDB ¹³ | on large dataset possible Open Source SQL-like language Native HTTP API Can process big amount of data. Focus on time series. | Maturity | |
| 27. 1010data | Interactive analytical service. A rich set of analytic functions are integrated. Suit for very large data set. Decent performance with high scalability. | Price. Should support more complex database operations. | |
| 28. BitYota: ¹⁴ | Data warehouse as a serviceSQL | • | |

| 29. AWS RedShift ¹⁵ * Scaling is easy * You can use SQL to query * Can work seamlessly with other AWS services 30. SpaceCurve: 16 * Focus mainly on spatial data. * Good at real-time location analysis * Query: Some patent-pending strategies that optimize quires * View: Algorithm that make updating views more efficient * Supports serializable concurrent transection 31. LogicBlox: * View: Algorithm that make updating views more efficient * Supports serializable concurrent transection 32. MonetDB: 17 * Column oriented store: * Good at OLAP scenario * Higly Compressed * Supports both row and column-oriented store disadvantage * Increased disk seek time | | | • | Real time data analysis | | |
|--|-----|-----------------------------|---|---|---|---|
| data. Good at real-time location analysis Public Response serializable concurrent transection 1. LogicBlox: Query: Some patent-pending strategies that optimize quires View: Algorithm that make updating views more efficient View: Algorithm that make updating views more efficient Supports serializable concurrent transection 1. Supports serializable concurrent transection 2. MonetDB: 17 Column oriented store: Good at OLAP scenario Higly Compressed Column-oriented storage Highly scalable Highly scalable Column oriented store disadvantage Highly scalable Increased disk seek time | 29. | AWS RedShift ¹⁵ | • | Scaling is easy You can use SQL to query Can work seamlessly with | • | maintain the uniqueness of data, programmer are responsible for their data |
| pending strategies that optimize quires View: Algorithm that make updating views more efficient Supports serializable concurrent transection Column oriented store: Good at OLAP scenario Higly Compressed Supports both row and column-oriented storage Highly scalable Highly scalable Pending strategies that optimize quires Increased disk seek time Insertion costs more Increased disk seek time Insertion costs more Increased disk seek time Insertion costs more | 30. | SpaceCurve: ¹⁶ | • | data. Good at real-time location | • | strategies that optimize quires View: Algorithm that make updating views more efficient Supports serializable |
| Good at OLAP scenario Higly Compressed Supports both row and column-oriented storage Highly scalable Highly scalable Insertion costs more Column oriented storage Increased disk seek time Insertion costs more | 31. | LogicBlox: | • | pending strategies that optimize quires View: Algorithm that make updating views more efficient Supports serializable | • | |
| GreenPlum: ¹⁸ column-oriented storage • Highly scalable • Increased disk seek time • Insertion costs more • | 32. | MonetDB: 17 | • | Good at OLAP scenario | • | |
| + | 33. | | | column-oriented storage | | disadvantage Increased disk seek time |
| 4 Column Oriented 4 Highly compressed 5 Good at log parsing 6 Immaturity 7 Increased disk seek time 8 Insertion costs more | 34. | HP Verica | • | Highly compressed | • | |
| 35. SAP Sybase IQ ¹⁹ | 35. | SAP Sybase IQ ¹⁹ | • | Column Oriented | | |
| 36. ParStream:²⁰ Real time analysis Focus on IOT(Internet of Things) data | 36. | ParStream: ²⁰ | | Focus on IOT(Internet of | | |
| 37. IBM InfoSphere Real-time analytic platform Merge diverse data Column-oriented database In memory Highly compressed | 37. | IBM InfoSphere | | , , | • | Column-oriented database In memory |
| 38. Kx Systems: ²¹ Column store database advantages described in MonetDB Column store database disadvantages described in MonetDB | 38. | Kx Systems: ²¹ | • | advantages described in | • | Column store database disadvantages described in |
| 39. LucidDB: ²² Column store database Column store database | | | + | | | |

| | advantages described in MonetDB Open Source | disadvantages described in MonetDB |
|--------------------------------------|--|---|
| | Bitmap indexingHash join/aggregation | |
| 40. Kognitio ²³ : | Multiversioning In memory Support SQL integrated with HADOOP | • |
| 41. Actian Vector | A high performance analytic frame built on Hadoop Developer can use SQL to interact with the system | • |
| 42. MetaMarkets Druid: ²⁴ | A distributed real-time data store Real time ingestion Column-oriented storage's advantage Bitmap indexing Fault tolerance | Column-oriented storage's disadvantage |
| 43. Teradata ²⁵²⁶ | A decent data warehouse system Developers can choose to store the data either based on row or column | • Price |
| 44. SQream | Scalable SQL data base GPU based database brings high parallel processing ability Column oriented storage advantages | Column oriented storage disadvantages |
| 45. RainStor | Can work with different data types | • |
| 46. HPCC ²⁷ | Introduced a new programming language: ECL It is more complex than a key-value pair storage. High availability, scalability and consistent | Still growing. |
| 47. Teradata Aster: ²⁸ | Allows users to write map reduce code that manipulate relational data base data. | • |

| | | • | Graph analytics engine | | |
|-------|------------------------------|---|--------------------------------|---|---|
| | | • | Support massive parallel | | |
| | | · | | | |
| 40 | SciDB ²⁹ | | processing | | |
| 48. | 2CIDR ₅₃ | • | Array data model | * | Focus mainly on Mathematic |
| | | * | Supports complex | | operations |
| | | | mathematic processing on | | |
| | | | the arrays | | |
| | | * | Can model uncertainty | | |
| 49. | Hadapt ³⁰ | • | Brings SQL to Hadoop, | • | No transections |
| | | | which allows users to write | | |
| | | | SQL to query on massive | | |
| | | | amount of data | | |
| | | • | Uses a hybrid storage | | |
| | | | engine which stores | | |
| | | | structured data in a | | |
| | | | traditional relational | | |
| | | | database while | | |
| | | | unstructured data in HDFS. | | |
| 50. | JethroData ³¹ | • | Like Hadapt, it builds a layer | ٠ | No transections |
| | | | on top of Hadoop that | | |
| | | | allows user to write SQL on | | |
| | | | Hadoop | | |
| | | • | Unique index strategy | | |
| | | • | Scalability that comes with | | |
| | | | HDFS | | |
| 51. | CitusDB: 32 | • | SQL on Hadoop | ٠ | No transections |
| | | • | Also suppor semi- | | |
| | | | structured data | | |
| | | • | Optimized specially for | | |
| | | | time-series data. | | |
| 52. | Impala: ³³³⁴ | • | SQL on Hadoop | • | No transections |
| | | • | It's supported by Cloudera | • | Data need to be stored in a |
| | | | it o capported by cioudera | | specific data format |
| 53. | IBM Big SQL ³⁵ | • | SQL on Hadoop | • | No transections |
| | Presto | • | Sgl on Hadoop | • | No transections |
|] 34. | 110310 | • | Can query data from | | No transcettons |
| | | | different source and bring | | |
| | | | them together | | |
| 55 | Apache Drill: ³⁶ | • | SQL on Hadoop | • | No transections |
| 55. | Apache Dilli. | • | Apache license | | ivo transcottons |
| | | | Can work with semi- | | |
| | | | structured or nested data | | |
| | | | | | |
| - | A 1 2720 | • | Low latency | | N L L L L L L L L L L L L L L L L L L L |
| 56. | Apache Hive: ³⁷³⁸ | • | Data warehouse built on | • | No update and delete |

| | Hadoop | operation |
|-------------------------------|--|----------------------------------|
| | Use a SQL like language | No access control |
| | Bitmap index | The overhead brought by Map |
| | Supports different storage | Reduce make it a little bit slow |
| | type. | |
| | Data are compressed | |
| | • | |
| 57. Apache Tajo ³⁹ | Fully distributed SQL | No transections |
| | Various query optimization | |
| | Supports ANSI/ISO SQL | |
| | Has a shell | |

C. Big Tables

| System | • Pros | • Cons |
|----------------------------------|--|---------------------------------|
| 58. Google Cloud | No schema is needed, Aims | No database layer caching |
| Datastore ⁴⁰⁴¹ | at storing non-relational | No Join |
| | data | Filter results using a subquery |
| | Write scale and read scale. | is not supported |
| | Supports transection | |
| 59. Google App | Key-value pair store makes | Does not support ACID |
| Engine | it more flexible | transactions |
| Datastore ⁴² | | No join |
| 60. Cassandra.io ⁴³⁴⁴ | Linear scalability(All nodes | No Join |
| | are identical) | Does not support ACID |
| | Fault-tolerance on | |
| | inexpensive hardware | |
| | The language it uses(CQL3) | |
| | is very similar to SQL | |
| | Constant-time writes | |
| | Integrated with Hadoop | |
| 61. Accumulo ⁴⁵⁴⁶ | Wode Column Store DB | • |
| | similar to Cassandra and | |
| | HBase | |
| | Better Performance(can | |
| | scan 800k entries per | |
| | second per node) compare | |
| | to HBase | |
| | Provides cell-level security | |
| 62. Hbase ⁴⁷ | Works hand in hand with | No strict ACID |
| | Hadoop | Because of its master and |

| | • | Specially optimized for real time analysis | | slave architecture, Hbase has the problem of single point |
|------------------------------|---|--|---|---|
| | • | Also linear scalability | | failure |
| | • | Consistent reads and writes | ٠ | No join |
| | * | Row level Atomic | | |
| 63. HyperTable ⁴⁸ | + | Implements using c++ | * | |
| | • | Runs on haddop | | |
| | • | SQL like language | | |
| | • | Faster and smaller than | | |
| | | HBase | | |
| 64. DataStax | * | Built on Cassandra | * | Similar to Cassandra |
| Enterprise ⁴⁹ | • | Comercial | | |
| | • | Similar to Cassandra | | |

D. Key value stores

| System | | Pros | | Cor | Cons | |
|--------|---------------------------------|------|---|-----|--|--|
| 65. | AWS DynamoDB ⁵⁰⁵¹ | • | Supports both document and key-value data | • | Consume more resource because its stronger | |
| | | * | Low latency | | Consistent constrain | |
| | | • | Highly scalable | • | No join | |
| | | • | Highly available | • | No support for transection | |
| | | • | Strong consistency on read | • | No ACID | |
| | | • | Supports atomic counters | | | |
| | | • | Secure: find access control | | | |
| 66. | AWS | • | Fit for smaller workloads | • | Table cannot grow over 10 GB | |
| | SimpleDB ⁵²⁵³ | • | Automatically index all | • | Not as scalable as DynamoDB | |
| | | | things | • | No joins | |
| 67. | MagnetoDB ⁵⁴ | • | A key-value storage for | • | No join | |
| | | | open stack | • | No support for transection | |
| | | • | Highly scalable | • | No ACID | |
| | | • | Supports both eventual and | | | |
| | | | strong consistency reads | | | |
| | | • | Fault tolerance | | | |
| 68. | Redis Cloud ⁵⁵ | ٠ | In memory non-relational | ٠ | No ACID | |
| | | | database | | | |
| | | • | Scale out seamlessly | | | |
| | | • | Zero Down time | | | |
| | | • | Secure | | | |
| 69. | Redis Labs ⁵⁶ | ٠ | Similar to Redis Cloud | ٠ | Similar to Redis Cloud | |
| 70. | AWS | ٠ | You can choose from two in | ٠ | Disadvantages are similar to | |
| | ElastiCashe ⁵⁷ | | memory cache options: | | Redis | |
| | | | Redos or Memcached | | | |

| | | | | 1 | |
|-----|----------------------------|---|-----------------------------|---|-------------------------------|
| | | • | The advantages are similar | | |
| | | | to those two | 1 | |
| 71. | | • | A redis management tool | • | Redis' disadvatages |
| 72. | RedisGreen ⁵⁸ | • | A redis hosting service | • | Redis' disadvatages |
| 73. | ObjectRocket | • | A redis hosting service | • | Redis' disadvatages |
| | Redis ⁵⁹ | | | | |
| 74. | HyperDex ⁶⁰ | • | Key-value storage | • | No Join |
| | | • | Strong consistency | | |
| | | • | Fault tolarence | | |
| | | • | ACID | | |
| 75. | LevelDB ⁶¹ | • | Key-value | • | No indexes |
| | | • | Comparison function can | • | It only allows one process to |
| | | | be customized | | access the database at a time |
| | | • | Compressed | | |
| 76. | BerkeleyDB ⁶² | • | Provides building blocks | • | |
| | | | that can help you develop | | |
| | | | your own data | | |
| | | | management solution | | |
| 77. | Oracle NoSQL ⁶³ | • | Key value storage with | • | No Join |
| | | | secondary indexes | | |
| | | • | ACID | | |
| | | • | Secure | | |
| 78. | Voldemort | • | | • | |
| 79. | Redis ⁶⁴ | • | In memory non-relational | • | No ACID |
| | | | database | | |
| | | • | Scale out seamlessly | | |
| | | • | Zero Down time | | |
| | | • | Secure | | |
| 80. | Couchbase ⁶⁵ | • | Key-value | • | No join |
| | | • | Document(Json) | • | No transection |
| | | • | | | |
| 81. | FatDB | ٠ | Tight intergration with SQL | • | |
| | | | Server | | |
| 82. | Riak ⁶⁶ | • | Buck key together | • | ACID |
| | | • | Strongly consistent | • | Join |
| | | • | Non-key based query use | | |
| | | | map reduce to get the | | |
| | | | answer | | |
| 83. | ArangoDB ⁶⁷ | ٠ | Multi-model database:S | • | |
| | | | upport documents, graphs | | |
| | | | and key-values data model | | |
| | | • | SQL-like | | |
| | | • | Joins like operation | | |
| | | • | Transections | | |
| L | | 1 | | 1 | |

| 84. Aerospike ⁶⁸ | • | Handle real time data | ٠ | Join |
|-----------------------------|---|-----------------------|---|------|
| | • | ACID | | |
| | • | Flash as storage | | |
| | • | Mainly key-value | | |
| | • | Map reduce | | |

E. Hadoop

| System | Pros | Cons |
|----------------------------------|------------------------------|--|
| 85. GridGain ⁶⁹ | In Memory data faric | • history |
| | Can act as a cache layer to | |
| | accelerate Hadoop | |
| | Realtime streaming | |
| | Linearly scale out | |
| | ACID transection | |
| 86. ScaleOut 70Software | In memory storage | • |
| 87. Pivoutal GenFire | Helps SQI to scale out using | Transection and ACID |
| XD ⁷¹ | Hadoop | |
| | High availability | |
| | In memory | |
| | • | |
| 88. Sqrrl Enterprise | Based on Apache Accumulo | disadvantages similar to |
| | Advantages similar to | Accumulo |
| | Accumulo | |
| 89. LucidWorks Big | A big data platform brings | • |
| Data ⁷² | together Hadoop solr and | |
| | etc. | |
| 90. Trafodion ⁷³ | SQL on Hbase | • |
| | ACID Transection | |
| | • scaling | |
| 91. Splice Machine ⁷⁴ | Full function SQL on | • |
| | Hadoop | |
| | Scale out | |
| | Transection | |
| | High concurrency | |
| | Real time updates | |
| 92. Apache Tajo - | Already described in | • Already described in Specialist |
| Pivaotal HD | Specialist analytic | analytic |

F. Appliance

| System | | Pros | Cons | |
|--------|--|--|---|--|
| 93. | Oracle Big Data Appliance ⁷⁵ | Cloudera Enterprise Technology software Oracle NoSQL database Integrated solution | Integrated solutionprice | |
| 94. | Oracle Exalytics ⁷⁶ | In memory integrated solution | Integrated solutionprice | |
| 95. | Microsoft SQL Server PDW | Integrated solution | Integrated solution | |
| 96. | SAP HANA ⁷⁷ | In memoryColumn orientedRelational databse | Increased disk seek timeInsertion costs more | |
| 97. | IBM Pure Data | Integrated solution | Integrated solution | |
| 98. | Oracle Exadata | Integrated solutionflash | Integrated solution | |

G. Graph

| System | Pros | Cons |
|---------------------------------|--|--|
| 99. infiniteGraph ⁷⁸ | Distributed graph database graph specific queries Policy-driven consistent Data visualization is integrated | Does not support map reduce Does not support data compression Eventual consistency |
| 100. HypergraphDB ⁷⁹ | Graph oriented graph specific queries Transection | Does not support map reduce |
| 101. Allegrograph ⁸⁰ | ACID transection Automatic indexing SOLR and MongoDb are integreted Secure Sharding | Does not support map reduce |
| 102. Giraph ⁸¹ | Data analysis tool on graph data Apache Used by Facebook Runs as map reduce jobs | • |
| 103. SPARQLBASE ⁸² | Graph databse In memory | • |

| | Uses HDFC to store data | |
|-----------------------------|-----------------------------|---|
| 104. Trinity ⁸³ | Embed or distributed graph | Not mature |
| , | storage | |
| | In memory | |
| | Data compressed | |
| 105. Titan ⁸⁴ | Distributed graph database | + |
| | Support Transection and | |
| | eventual consistency | |
| | Can use Cassandra, HBse or | |
| | BerkeleyDB to store data. | |
| | Support geo, numeric and | |
| | text search | |
| | Map reduce | |
| 106. Objectivity:85 | Graph NoSQL database | • |
| | Good at exploring | |
| | relationships in data. | |
| | Suits for areas like social | |
| | networks. | |
| 107. Stardog | Graph database | • |
| | • ACiD | |
| 108. FlockDB 86 | Graph database | • Fewer function cause it's |
| | Twitter uses it to store | simpler(maybe it's an |
| | social graphs | advantage) |
| | Designs for websites | |
| 109. GrapheneDB | Cloud hosting Neo4j | ◆ Same as Neo4j |
| 110. Sparksee ⁸⁷ | Data compression(use | • |
| | bitmap to represent data) | |
| 111. Neo4j ⁸⁸ | High Availability | Does not support map reduce |
| | Data compression | Has Max size value limitation |
| | Fully ACID | |
| | • | |
| 112. CortexDB ⁸⁹ | Multiple data model: key- | • |
| | value, graph, multi value | |
| | column | |
| | Distributed | |

H. Data Caching

| System | Pro | S | Cor | าร |
|-------------------------------|-----|--------------------------|-----|---------|
| 113. MemCachier ⁹⁰ | • | In memory scalable key | ٠ | No ACID |
| | | value pair cache | | |
| | • | Better reliability and | | |
| | | usability than memcached | | |

| 114. Redis | In memory non-relational No ACID | |
|-------------------------------|--|------------|
| | database | |
| | Scale out seamlessly | |
| | Zero Down time | |
| | Secure | |
| | • Persistent | |
| 115. Redis Labs | Cloud hosting Memcached Similar as Memcached | hed |
| Memcached | Similar as Memcached | |
| Cloud ⁹¹ | | |
| 116. IronCache ⁹² | Key value cache No ACID | |
| | Cloud service | |
| | Can persist the data | |
| 117. AWS ElastiCache | You can choose from two in Disadvantages are s | similar to |
| | memory cache options: Redis and Memcac | hed |
| | Redos or Memcached | |
| | The advantages are similar | |
| | to those two | |
| 118. BigMemory ⁹³ | In memory data store | |
| | Supports SQL | |
| | Runs Ehcache | |
| 119. Ehcache ⁹⁴ | Im memory data store | |
| | Schema less | |
| | • ACID | |
| | Sharding and replication | |
| 120. InfiniSpan ⁹⁵ | In memory key value data | |
| | store | |
| | Support Map reduce | |
| | Support data compression | |
| | Support full text search, | |
| | and graph data | |
| | • Persistent | |
| | ACID transection | |
| 121. RedHat JBoss | In memory distributed | |
| Data Grid ⁹⁶ | caching | |
| | Support map reduce | |
| | Supports replication | |
| | Transection | |
| | Redhat support | |
| 122. Memcached ⁹⁷ | In memory key value pair Value is limited to 1 | IMB |
| | cache | |
| | Simpler than Redis makes it | |
| | easier to scale out | |
| | + ACID | |

I. Document

| System | Pros | Cons |
|-------------------------------|---|--|
| 123. Informix ⁹⁸ | Real time processing Availability: zero down time Supports SQL, JSON, and even time/special data Support Rest API | Comercial |
| 124. JumboDB ⁹⁹ | Supports indexing on Json Supports data compression Supports complex data model | No sharding and replication yet Immaturity Join No ACID Transection |
| 125. RethinkDB ¹⁰⁰ | Use Json as storage Supports complex data model Sharding and replication Fault tolerance MapReduce Schema-less | No Join No ACID Transection |
| 126. CouchDB ¹⁰¹ | JSON as storage Supports features that important to web development such as real time change notification Supports complex data model MapReduce Eventual consistency Schema-less | No ACID Transection No join |
| 127. RavenDB ¹⁰² | Schema-lessData compressionACIDMapReduce | • |
| 128. TokuMX ¹⁰³ | A high performance distribution of MongoDB Better caching strategy Optimized IO Supports document-level locking allows better concurrency | No ACID Transection |
| 129. MongoDB | Use Json as storageSupports complex data | No ACID No Join |

| | model | |
|--------------------|---|--|
| | Supports immediate and | |
| | strong consistency | |
| | Supports Sharding and | |
| | replication | |
| | Schema-less | |
| 130. Compose | Cloud hosting mongodb Similar as mongo db | |
| | Similar as mongodb | |
| 131. Iris Couch | Cloud hosting CouchDB Similar as CouchDB | |
| | Similar as CouchDB | |
| 132. MongoLab | Cloud hosting mongodb Similar as mongo db | |
| | Similar as mongodb | |
| 133. Object Rocket | Cloud hosting mongodb Similar as mongodb | |
| | and redis | |
| | Similar as mongodb | |
| 134. Azure | Use Json as storage No Join | |
| DocumentDB | Supports complex data | |
| | model | |
| | Schema-free | |
| | Supports different level of | |
| | consistency | |
| | Transection | |
| 135. Cloudant | ◆ Use Json as storage | |
| | Supports complex data | |
| | model | |
| | Schema-free | |
| | Supports Full-text search | |
| | Supports sptial indexes | |
| | Data compression | |

To be continued...

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