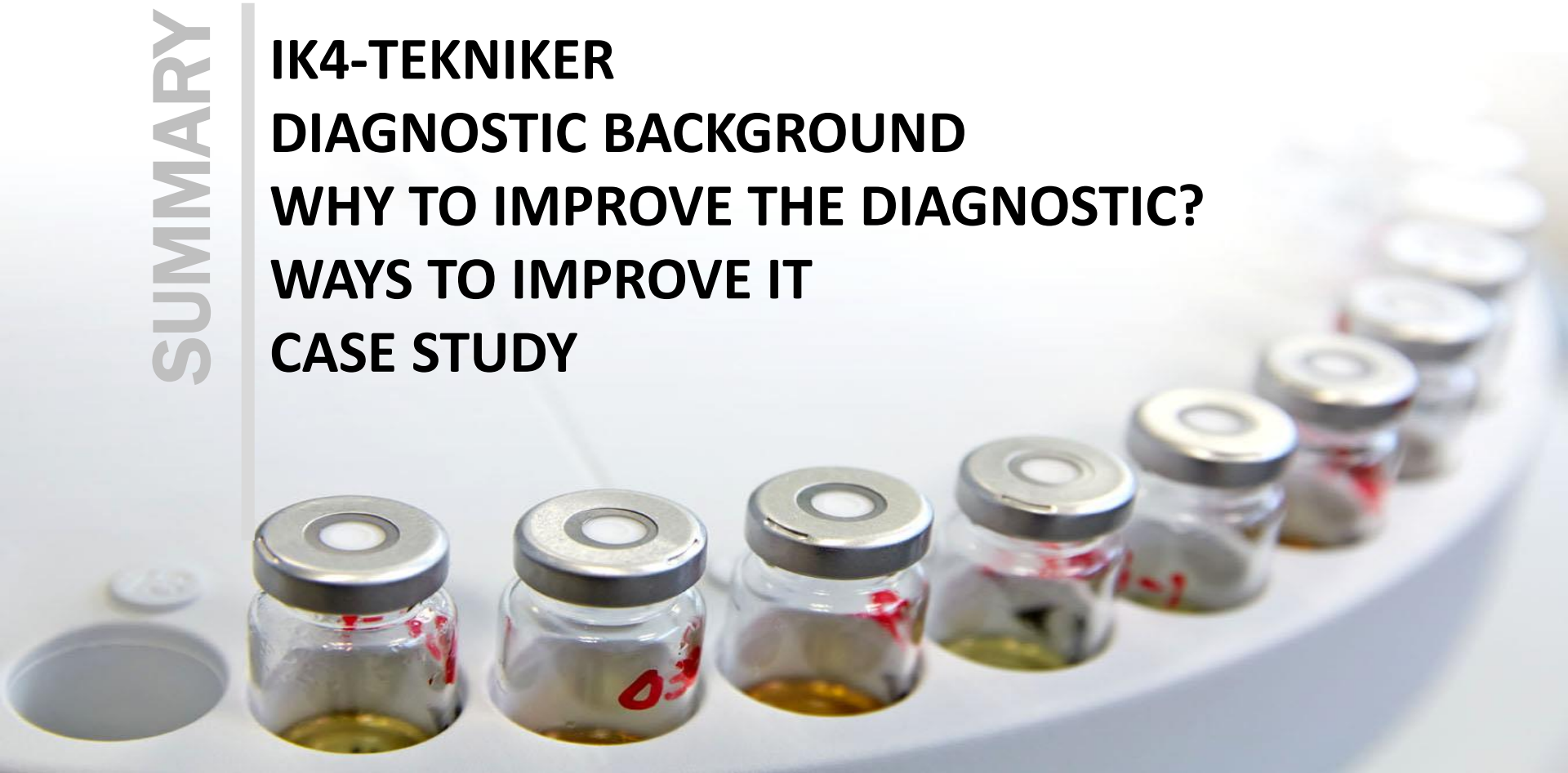


Improving the diagnostic capabilities with Big Data Analytics and Prognostics

JORGE ALARCON | JESUS TERRADILLOS
IK4-TEKNIKER

SUMMARY

**IK4-TEKNIKER
DIAGNOSTIC BACKGROUND
WHY TO IMPROVE THE DIAGNOSTIC?
WAYS TO IMPROVE IT
CASE STUDY**



INDUSTRIAL MILESTONES

**Connected
and Smart
devices**



1800

Industry 1.0
Steam-driven
production



1900

Industry 2.0
Mass production
supported by
specialization
and electricity



1970

Industry 3.0
Production
automation,
electronics and
IT



2016

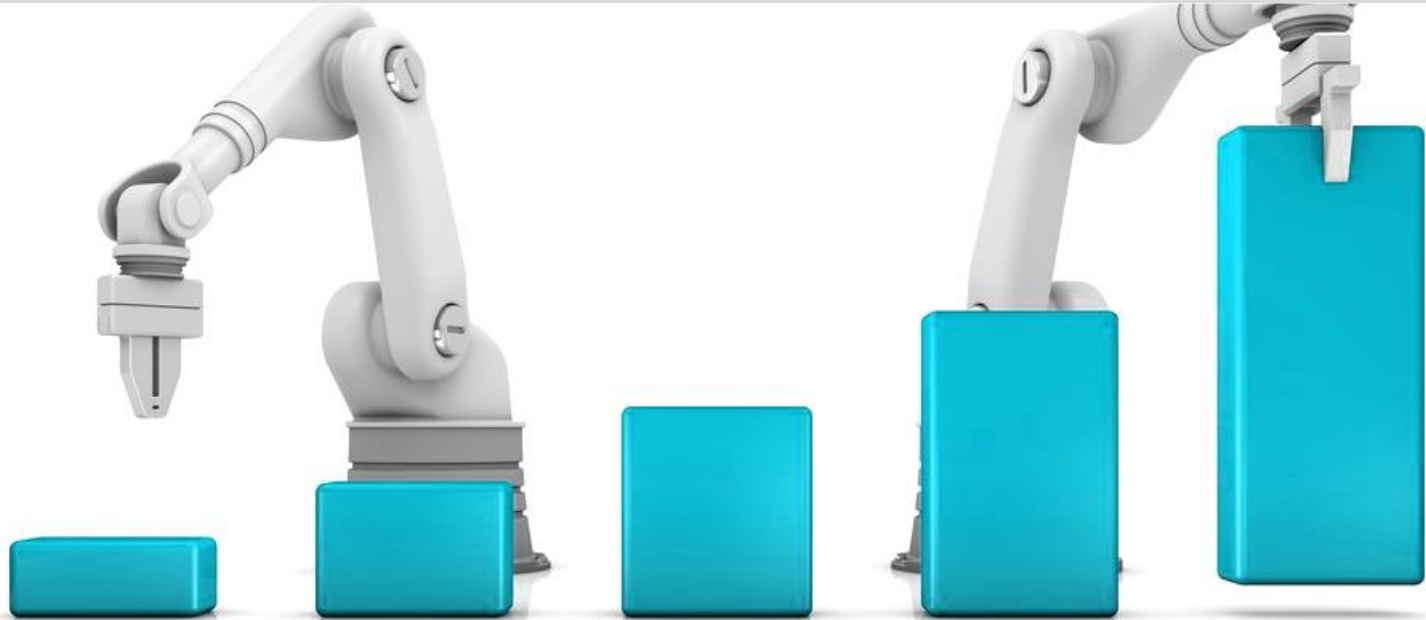
Industry 4.0
Production by
cyber-physical
systems

EVOLUTION OF MANUFACTURING

NEW DIGITAL
AGE

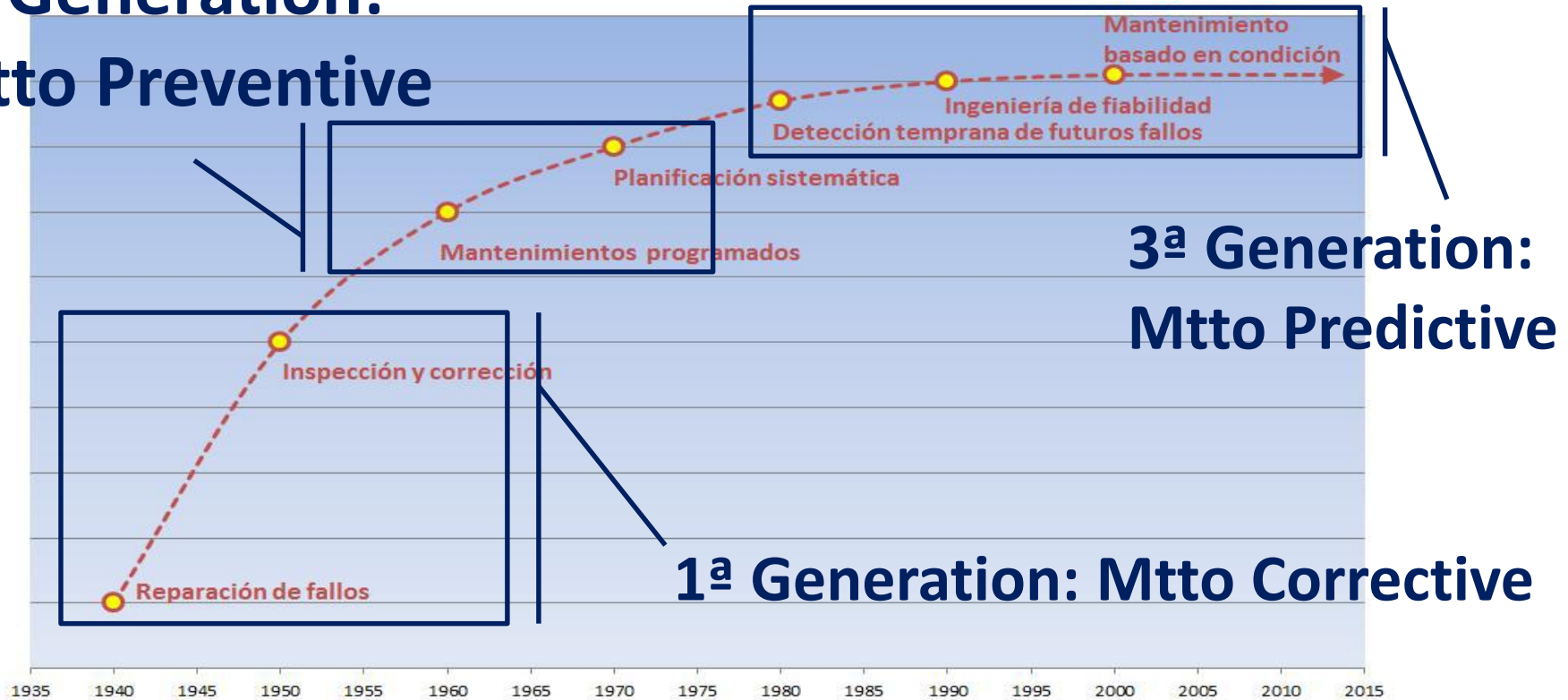
NEW MARKET
EXPECTATIONS

CHANGES IN
CUSTOMER
BEHAVIOR

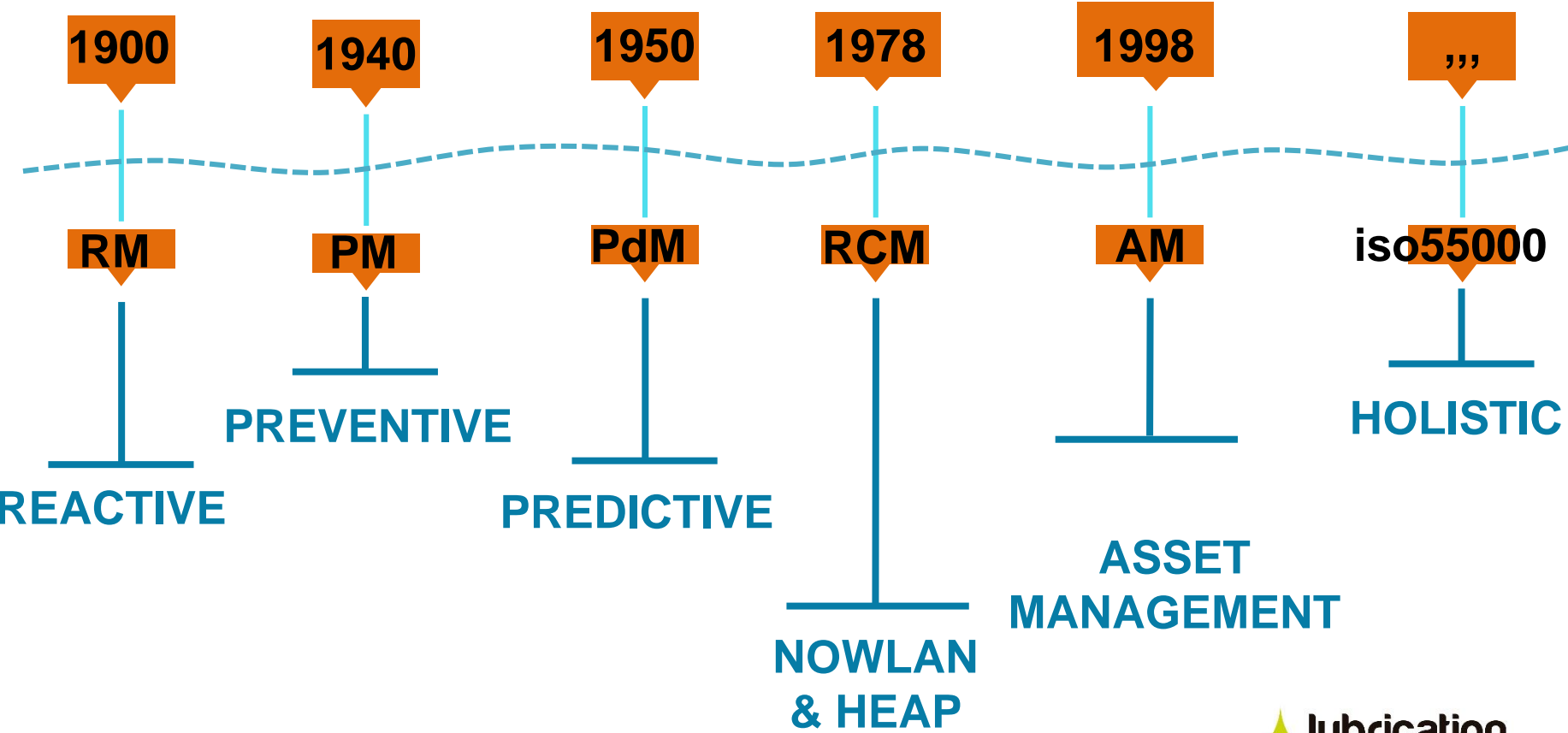


WE ARE MOVING TOWARDS 4TH MAINTENANCE GENERATION

2ª Generation: Mtto Preventive



MAINTENANCE MILESTONES



32%

Appropriate Lubrication processes in plant

Ref: Machinery Lubrication Reader Survey - March, 2011 (n: 347)

40%

Average time spent on corrective maintenance activities

Ref: Think Act - November 2014

WORLD CLASS < 6%

50%
mechanical components

IS THIS A GOOD REASON TO
IMPROVE DIAGNOSTIC ON
MONITORED EQUIPMENT?

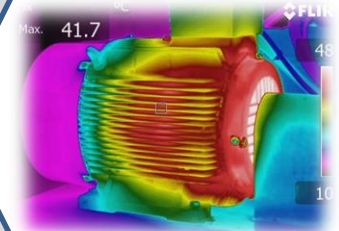
are replaced due to **premature wear**

DATA SOURCES

CONDITION MONITORING TOOLS



**HOW
WHEN**



NEW SCENARIO?



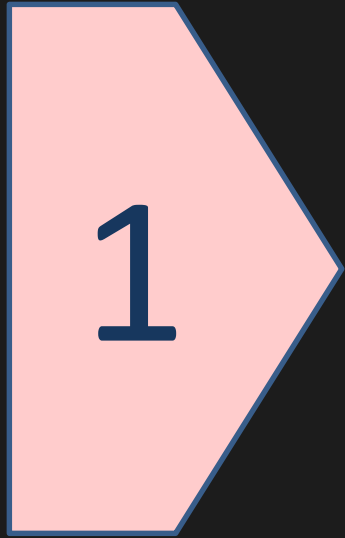
1970's →

COMMON DIAGNOSTIC

→ 2015



WHAT IS DIAGNOSTIC?



**EQUIPMENT
IN SERVICE**



**DATA
CAPTURE**



**SOFTWARE
ANALYSIS**



**HUMAN
DECISION**

In 20 years

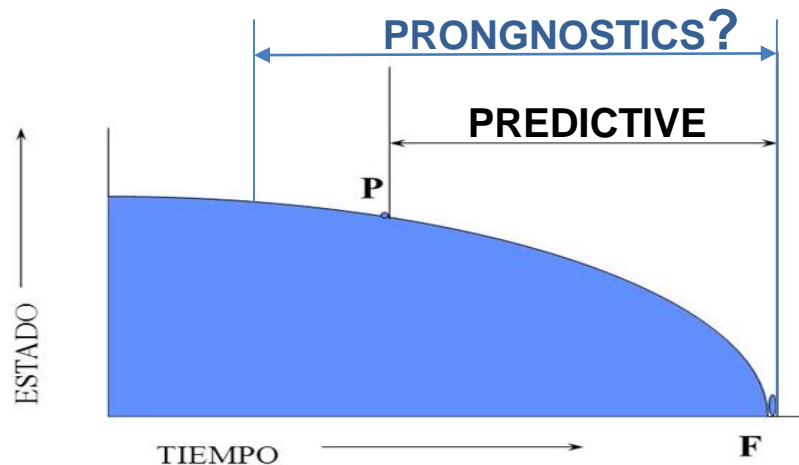
of **jobs** will be automated
by **machines** or
software



47%

Source :
KwonMads, *John Moravec*

There is enough information (Big Data)
Strange but important events
Multidisciplinary team: Experts + analysts



76%

Industrial companies are **not aware** of the potential on the analysis and their data

Ref: SAS Report on Big Data 2013

WAYS TO IMPROVE IT

COMMON DIAGNOSTIC

+

BIG DATA ANALYTICS

+

PROGNOSTICS



BIG DATA

Is a new paradigm that represents the search for solutions to store and analyze structured and unstructured data together for an affordable and scalable data mode.



PREDICTIVE ANALYTICS

A variety of statistical techniques from modeling, machine learning, and data mining that analyze current and historical facts to make predictions about future, or otherwise unknown, events.

PROGNOSTICS

Algorithms to detect interesting patterns in data.

Statement about the way things will happen in the future

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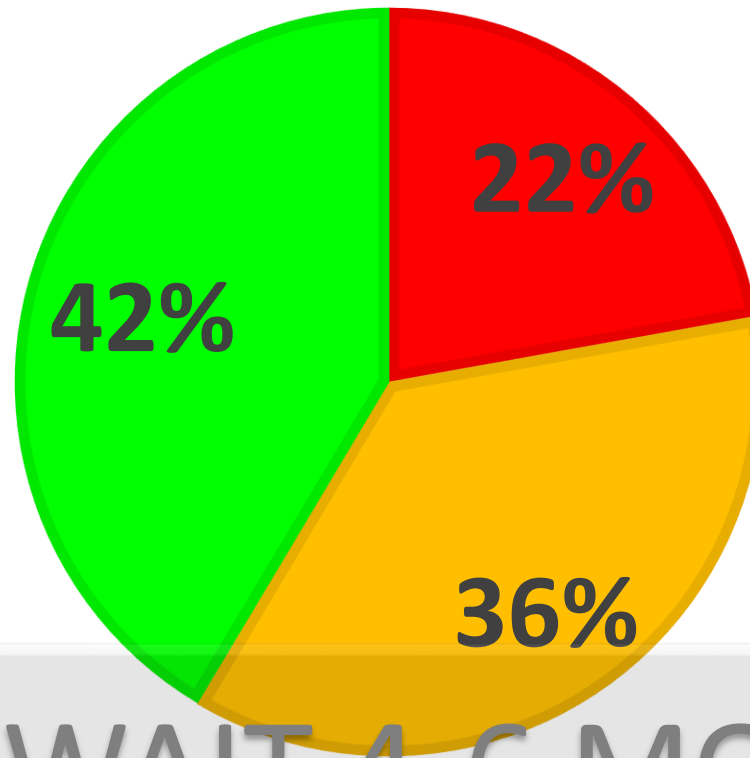


CASE STUDY

- Critical bearings in Plant: **320**
- Schedule Maintenance Shutdown: **4-6 months**
- 5 years of grease analysis of each bearing
- 2 different type of greases
- 3 different bearing models

BEARING CONDITION

■ CRITICAL ■ ALERT ■ GOOD



CAN WE WAIT 4-6 MONTHS?

Critical 71: DO SOME MAINTENANCE!

Alert 115: BE AWARE OF CONDITION

Good 134: ENOUGH TIME

- 1 bearing shutdown = **1.500 € per day**
- Time to repair a bearing = **3 days**
- Potential **production** losses = 319.500 €
(critical) + 517.500 € (alert)

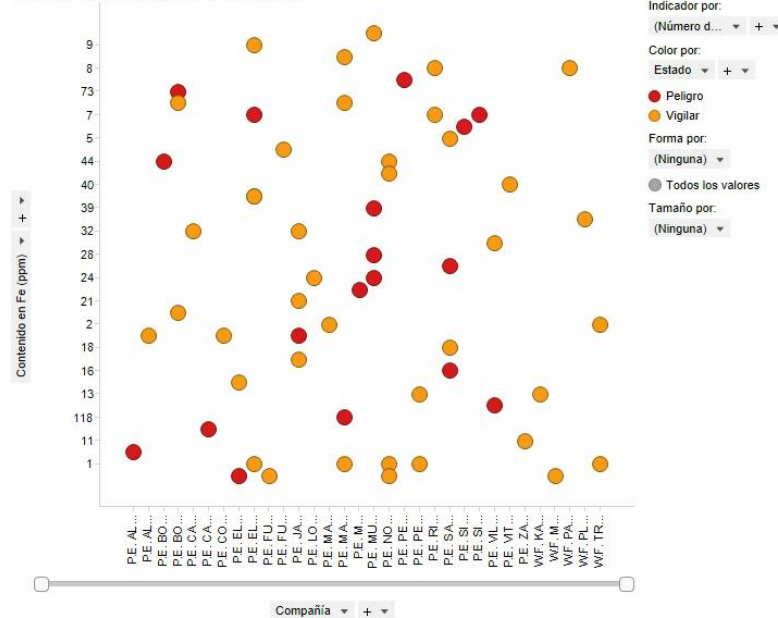
INFORMATION

- Around 4.000 reports
- 3 different type of failures detected in the past

1	996566	P.E. LA FE - SAN MARTIN	05/12/2015	03/09/2015			23277 H	The oi
2	1012344	P.E. ALENTISQUE	03/11/2015	03/11/2015	NULL	50815 H	50815 H	There wear
3	1012991	P.E. CASTELLANETA BKW	04/11/2015	04/11/2015			18648 H	The oi
4	1013334	P.E. FUENTE VAIN	30/10/2015	05/11/2015			46200 H	contar
5	1013416	P.E. SIERRA DE OLIVA	30/10/2015	06/11/2015			54916 H	Viscos
6	1013417	P.E. SIERRA DE OLIVA	30/10/2015	06/11/2015			62054 H	2091.
7	1013419	P.E. SIERRA DE OLIVA	30/10/2015	06/11/2015			62054 H	The oi
8	1013421	P.E. SIERRA DE OLIVA	30/10/2015	06/11/2015			54916 H	contar
9	1013423	P.E. SIERRA DE OLIVA	30/10/2015	06/11/2015			62054	The oi
10	1013425	P.E. LEZUZA	30/10/2015	06/11/2015			35605 H	contar
11	1014295	W.F. SOUDAN	10/11/2015	10/11/2015			27163 H	Contac
12	1014305	P.E. VIRGEN DE LA PEÑA	30/10/2015	10/11/2015			76138 H	The oi
13	1014312	P.E. LUCERA	10/11/2015	10/11/2015			27526 H	contar
14	1014324	P.E. LUCERA	10/11/2015	10/11/2015			27526 H	The oi
15	1014348	P.E. FOIANO-BN EDISON	10/11/2015	10/11/2015			35781 H	contar
16	1014352	P.E. FOIANO-BN EDISON	10/11/2015	10/11/2015			137483 H	The oi
17	1014357	P.E. GINESTRA STAR WIND	10/11/2015	10/11/2015			138897 H	contar
18	1014570	P.E. JANDA III	30/10/2015	11/11/2015		26272	26272	The oi
19	1014571	P.E. JANDA III	30/10/2015	11/11/2015		26272	26272	contar
20	1014572	P.E. EL PINO	05/11/2015	11/11/2015		80069 H	80069 H	The oi

BIG DATA ANALYTICS & OIL ANALYSIS

Contenido en Fe (ppm) frente a Compañía

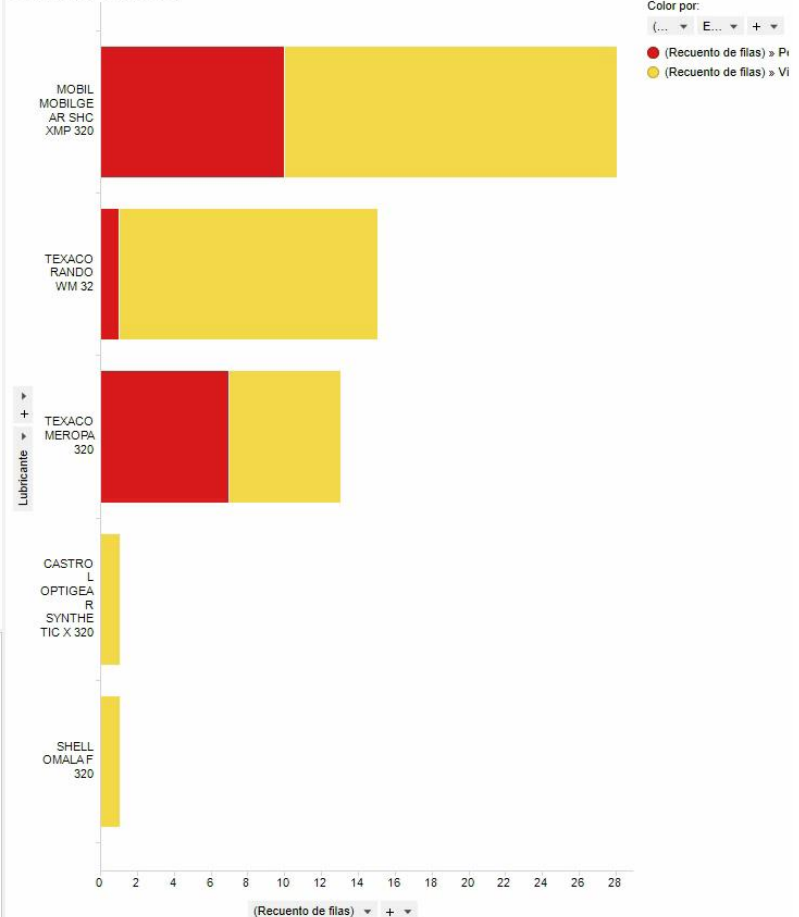


Recuento de filas según Lubricante y Estado

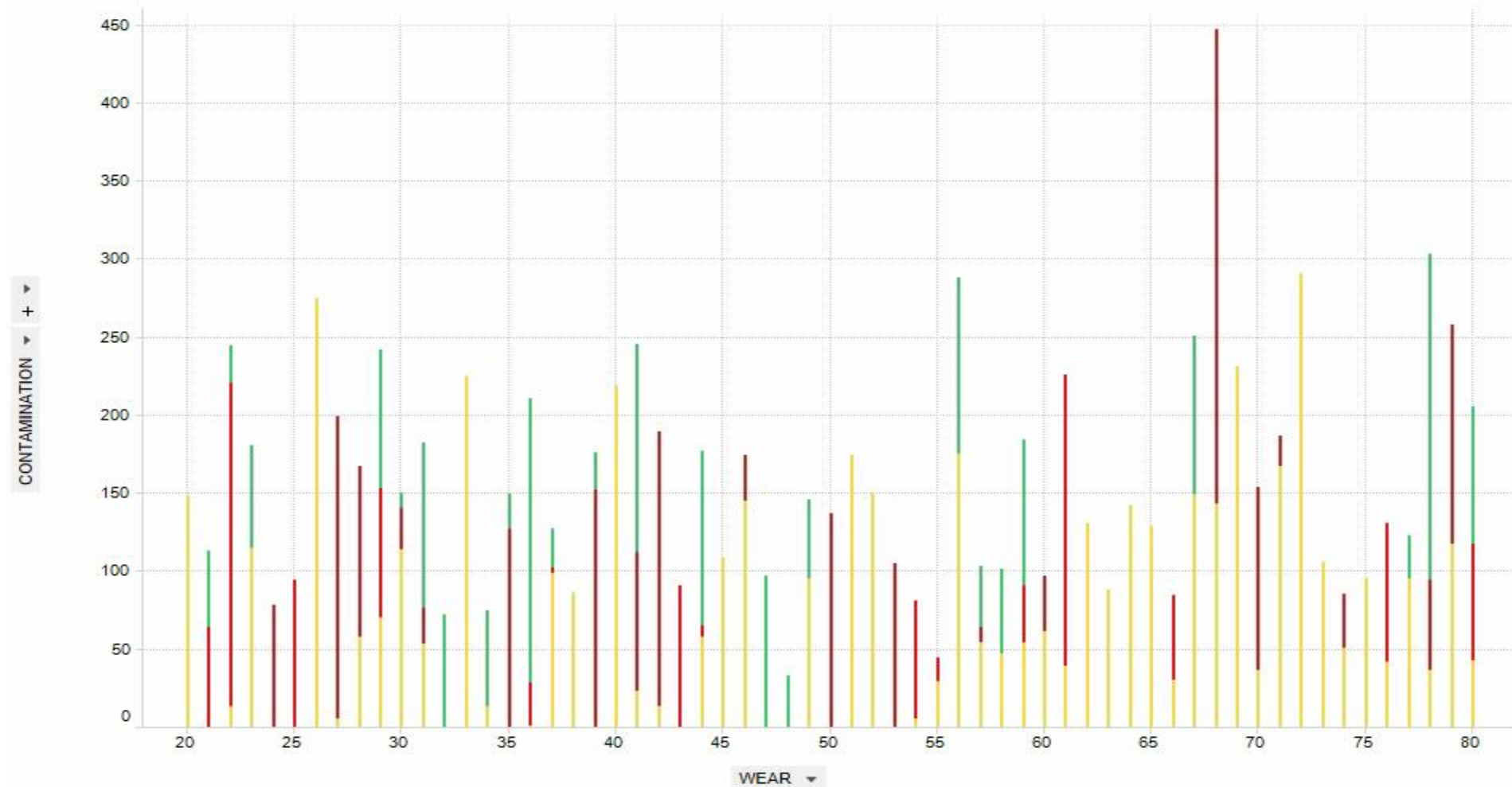
Lubricante	Estado	
	Vigilar	Peligro
TEXACO RANDO WM 32	14	1
TEXACO MEROPA 320	6	7
MOBIL MOBILGEAR SHC XMP 3...	18	10
CASTROL OPTIGEAR SYNTHETI...	1	---
SHELL OMALA F 320	1	---

(Recuento de filas) + -

Distribución – Lubricante

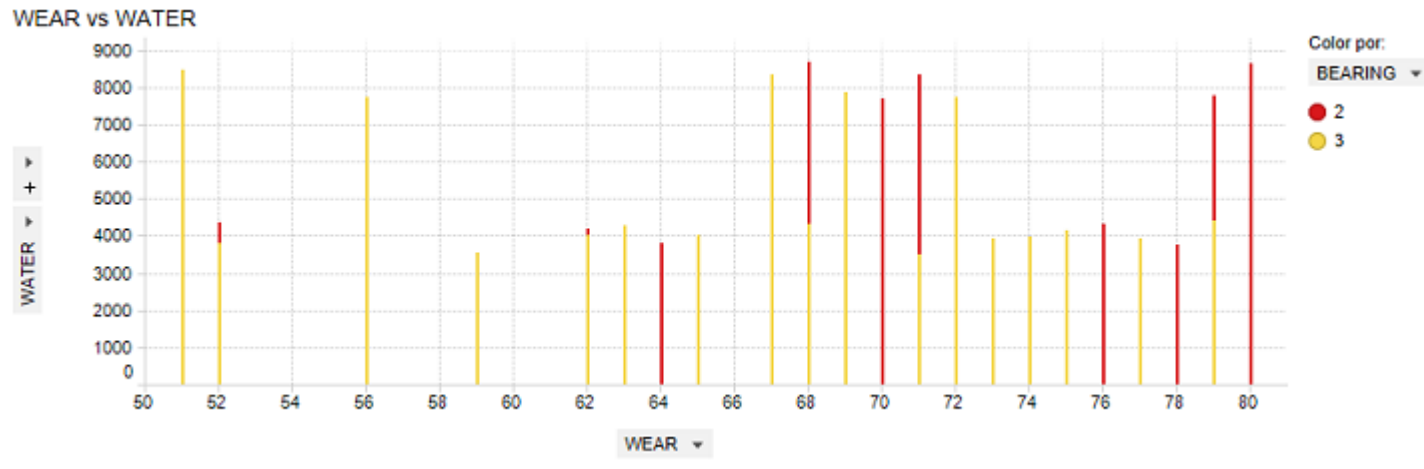


WEAR vs CONTAMINATION



HOW is going to fail?

BIG DATA ANALYTICS



9 bearings of type 2 - Failure type b
17 bearings of type 3 - Failure type a

PROGNOSTICS

Especificación de flota

Nombre del operador	Acciona Energia
Nombre de flota	VED
Localización flota	Monte Vedadillo (Falces)
Tipo de flota	Parque Eólico
Capacidad de la flota	49,5 MW
Flota ID	VED
Unidades monitoreadas	VA

Diagnóstico de condición

VA605	VA505	VA603	VA606	VA503
VA601	VA204	VA301	VA306	VA602
VA604	VA202	VA302	VA303	VA304
VA305	VA301	VA303	VA304	VA501
VA502	VA504	VA607	VA101	VA102
VA104	VA201	VA203	VA205	VA206
VA103	VA105	VA202		

Condición



Pronósticos de mal funcionamiento

Selección	Equipo crítico	Capacidad	En'16	Abr'16	Jul'16	Oct'16	En'17	Abr'17	Jul'17	Oct'17	En'18	Abr'18	Jul'18	Oct'18
<input checked="" type="checkbox"/>	VA605	MPA (L1.M1.5)	1,5 MW											
<input checked="" type="checkbox"/>	VA505	MPA (L1.M1.2)	1,5 MW											
<input checked="" type="checkbox"/>	VA603	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA606	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA503	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA601	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA204	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA301	MPA (L1.M1.1)	1,5 MW											
<input checked="" type="checkbox"/>	VA306	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA602	MPA (L1.M1.1)	1,5 MW											
<input checked="" type="checkbox"/>	VA604	Condición normal	1,5 MW											
<input checked="" type="checkbox"/>	VA202	Condición normal	1,5 MW											

- Riesgo de tiempo inactiv.: >25%
- Riesgo de tiempo inactiv.: 10% - 25%
- Riesgo de tiempo inactiv.: <10%
- Pronósticos alta resistencia
- Fuerza media pronóstico
- Baja resistencia pronóstico

Horario



WHEN is going to fail?

PROGNOSTICS

Pronósticos de mal funcionamiento

Seleccione	Modos de funcionamiento	Fuentes de d.	En'16	Abr'16	Jul'16	Oct'16	En'17	Abr'17	Jul'17
<input checked="" type="checkbox"/> M1.1	Fractura de diente	L1							
<input checked="" type="checkbox"/> M1.2	Contaminacion con agua	L1							
<input checked="" type="checkbox"/> M1.3	Contaminacion con particulas	L1							
<input checked="" type="checkbox"/> M1.4	Degradacion de aceite	L1							
<input checked="" type="checkbox"/> M1.5	Desgaste	L1							
<input checked="" type="checkbox"/> M1.6	Peridida de aditivacion	L1							
<input checked="" type="checkbox"/> M1.7	Contaminacion cruzada	L1							

Critical 71:

31% will fail in the next 3 months

Case Study Conclusions

DIAGNOSTICS



BIG DATA
ANALYTICS



PROGNOSTICS

To sum up

JORGE ALARCON

More Data doesn't just let us see more

More data allow us to see new

More data allow us to see better

More data allow us to see different
