# 光學檢測 Optical Methodologies for Mechanics and Industrial Applications

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# About Myself

- Researcher Fellow, Taiwan Instrument Research Laboratory, NARLabs
- AdCom Member, IEEE Instrumentation and Measurement Society, 2016-2024
- Chair of Optical Methods Technical Division, Society of Experimental Mechanics, 2023/06-2025/05
- Chair of TC 18, IEEE Instrumentation and Measurement Society
- Secretary, IEEE Taipei Section Instrumentation and Measurement Society Chapter
- Board Members, Taiwan Photonics Society, 2017-2019
- Vice President for Education, IEEE Instrumentation and Measurement Society, 2023-2024
- Vice President for Conferences, IEEE Instrumentation and Measurement Society, 2017-2018
- Associate Editor-in-Chief, IEEE Transections on Instrumentation and Measurement, 2024/05-
- Associate Editor, IEEE Transections on Instrumentation and Measurement, 2020-2024

# Topics

- General Introduction
- Why optical methodologies?
- Define Requirement-Knowing Limitation of a Method
- Key Element
  - Light Source
  - Sensors
  - Optical Lens& Optical Components
- Principle of Basic Interferometry

- Spectrum and Its Applications
- Laser Triangulation Measurement Method
- Structure Lights and Its Application
- Moiré Method/ Sampling Moiré Method
- Astigmatic Method and Applications
- Principle of White Light Scanning Interferometer and its Applications
- Principle of Confocal Microscopy and its applications
- Principle of Conoscopic Holography and its applications
- Introduction to optical image processing and deep learning

First Report (40%)

Due: 2024/10/23 (-10% for a week delay)

Topic: Proposal for implementing an optical inspection system

**Content Required:** 

Figure out a scenario for implementing optical measurement to replace/ to upgrade/ to be installed to a machine for on-machine measurement; the reports should include the following items;

- (1) Descriptions for current problems/limitations/requirements;
- (2) Discussion for candidate methods based on description (1);
- (3) Detail working principle of selected method;
- (4) Risk/ Risk mitigation/ residual risk (risk can be associated with technology, finance, operation(bottleneck introduced), ...);
- (5) Conclusions-convince your virtual boss"!

1,000



# Grading Policy (2)

Final Report (60%)

Due: 2024/12/25 (No delay is allowed)

**Topic:** Self-Defined

**Content Required:** 

- (1) Selecting an optical measurement method;
- $(2) \ge 30$  papers are required;
- (3) Working principle, milestones (roadmap..), progressing, applications ...;

1,000

(4) Potential application/ new chance for on-machine measurement with possible technical barriers included



Contents lists available at ScienceDirect

#### Journal of Manufacturing Systems

journal homepage: www.elsevier.com/locate/jmansys



## Application of sensing techniques and artificial intelligence-based methods to laser welding real-time monitoring: A critical review of recent literature



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#### ARTICLEINFO

Keywords:
Laser welding
Real-time monitoring
Sensing techniques
Multi-sensor fusion technology
Artificial intelligence
Multi-monitoring objectives

#### ABSTRACT

Laser welding has been widely utilized in various industries. Effective real-time monitoring technologies are critical for improving welding efficiency and guaranteeing the quality of joint-products. In this paper, the research findings and progress in recent ten years for real-time monitoring of laser welding are critically reviewed. Firstly, different sensing techniques applied for welding quality monitoring are reviewed and discussed in detail. Then, the advanced technologies based on artificial intelligence are summarized which are exploited to realize varied objectives of monitoring such as process parameter optimization, weld seam tracking, weld defects classification, and process feedback control. Finally, the potential research problems and challenges based on real-time intelligent monitoring are discussed, such as intelligent multi-sensor signal acquisition platform, data depth fusion method and adaptive control technology. This fundamental work aims to review the research progress in laser welding monitoring and provide a basis for follow-on research.

#### 1. Introduction

Laser welding is an efficient joining technique, owning to numerous advantages such as higher productivity, flexibility and effectiveness, deeper penetration of the weld seam, higher welding speeds, and power density [1–3]. That is why it obtains an increasingly widespread application in the automotive [4,5], shipbuilding [6], aerospace [7,8], micro-electronics [9,10], and other industries [11,12]. The welding quality can be disturbed by various factors during welding process, such

filters are applied to capture the images of the keyhole, molten pool, spatters and plasma. Spectrometers [25], and photodiodes [26] are utilized to collect the optical signals include visible light (VIS), infrared light (IR), and ultraviolet light (UV). The IR cameras [27], near-infrared (NIR) cameras [28], and pyrometers [29] can be exploited to gather the thermal signal. Complex monitoring system usually consists of the above-mentioned sensors and the welding signals will be more comprehensively collected. [30,31].

According to different monitoring objectives [32-34], there are







### Printed Circuit Board Defect Detection Methods Based on Image Processing, Machine Learning and Deep Learning: A Survey

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ABSTRACT Printed circuit boards (PCBs) are a nearly ubiquitous component of every kind of electronic device. With the rapid development of integrated circuit and semiconductor technology, the size of a PCB can shrink down to a very tiny dimension. Therefore, high-precision and rapid defect detection in PCBs needs to be achieved. This paper reviews various defect detection methods in PCBs by analysing more than 100 related articles from 1990 to 2022. The methodology of how to prepare this overview of the PCB defect detection methods is firstly introduced. Secondly, manual defect detection methods are reviewed briefly. Then, traditional image processing-based, machine learning-based and deep learning-based defect detection methods are discussed in detail. Their algorithms, procedures, performances, advantages and limitations are explained and compared. The additional reviews of this paper are believed to provide more insightful viewpoints, which would help researchers understand current research trends and perform future work related to defect detection.

INDEX TERMS Defect detection, PCB, image processing, machine learning, deep learning.

#### I. INTRODUCTION

The printed circuit board (PCB) is one of the most vital units in the electronic industry [11]. It plays a key role in electronic devices, mechanically holding up and electrically connecting various electronic parts together. PCBs are used in almost every kind of electronic equipment, from electronic

During the Fourth Industrial Revolution (Industry 4.0) [3], PCB manufacturing is facing new challenges and opportunities. The core premise of Industry 4.0 is automated industrial processing with high quality, precision and reliability [4]. Therefore, the producing process of tiny complex PCB boards is required to be more stable and reliable with higher speed,



### 光學量測課程期中/期末報告格式

姓名 學號 \*Email:

Report Template 建應於 2020 年 5 月 1 日 與 2020 年 6 月 30 日 以電子郵件方式將告繳交至 ang@gmail.com。所有報告之格式皆需符合本文中所規範,並於截止日前24時前寄出, 建扣10%配分比。

:報告、格式、光學量測課程(最少3個關鍵字)

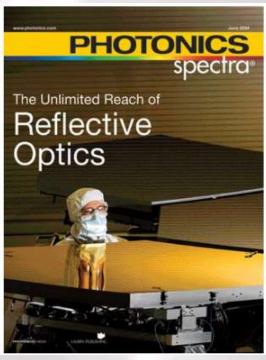
#### 1. 前言

期中報告必須包含現有技術描述,取代方法描述,取代過程所引入之時間成本與投資成本支出,並說明短中長期可能之影響;期末報告所討論之論文不得少於30篇。作者應遵守本樣板所規範之格式,並自行將文件轉換成PDF格式檔案後以E-mail方式傳送至 chihunghwang@gmail.com 完成

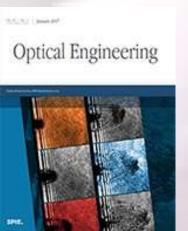
## Reference materials

### Journal

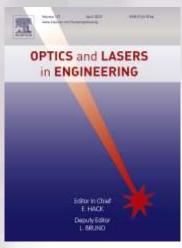
- Optics and Lasers in Engineering
- 2. Optical Engineering
- 3. Applied Optics
- 4. Vision Systems Design (Free)
- Photonics Spectra(Free)











## Societies

AOIEA 自動光學檢測設備聯盟
 https://aoiea.itri.org.tw/tc/index.aspx

中華民國光電學會(Taiwan Photonics Society)https://www.photonics.org.tw/tw/

中華民國計量工程學會(Chinese Metrology Society)
 https://www.tcms.org.tw/

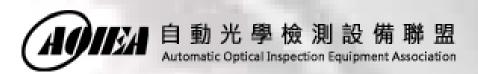
OPTICA (formerly Optical Society of America)
 https://www.optica.org/

• SPIE (The International Society for Optics and Photonics)

https://spie.org/#\_=\_

IEEE Photonics Society

https://ieeephotonics.org/





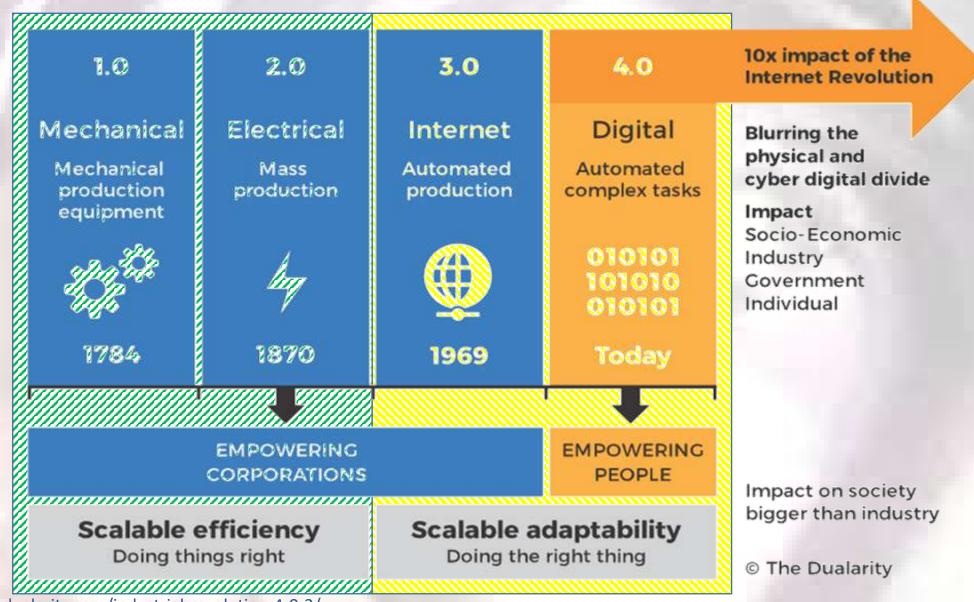








## Industrial Revolutions



https://www.thedualarity.com/industrial-revolution-4-0-2/

### The Start of the Fourth Industrial Revolution



3rd Industrial Revolution -

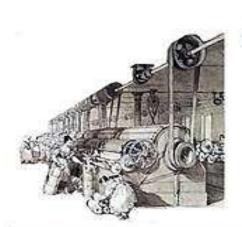
electronics, telephones,

PLCs, NC machines, PCs,

4th Industrial Revolution mobile, cloud, smart connected devices, cyber physical systems, smart factory, robots, mass customization, product as-service



Productivity, Complexity



1st Industrial Revolution – water and steam powered mechanical manufacturing facilities



............

2<sup>nd</sup> Industrial Revolution – manufacturing assembly line and infrastructure of electricity, gas, water, telegraph, roads

CAM, CIM, spreadsheets, Lean manufacturing ly of

Late 20th century

Today

End of 18th century Start of 20th century

itury To

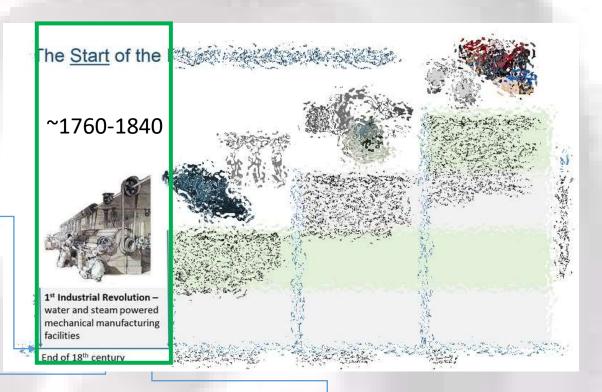
## Industrial Revolution

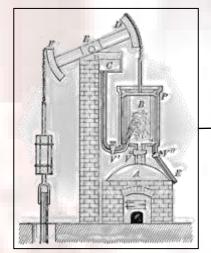
# Important Technological Developments

- Textiles
- Steam power
- Iron making
- Invention of machine tools

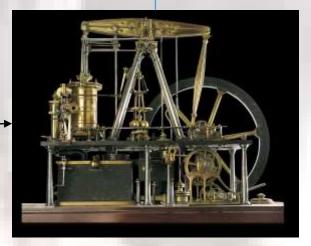


Prior to 18th-century workshop





Newcomen steam engine Invented in 1712



Watt Engine, 1763 to 1775
Milestone for Industrial Revolution

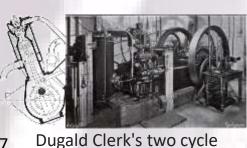


19th-century workshop

# 2<sup>nd</sup> Industrial Revolution

### **Internal combustion Engine**





engine in 1879



Rudolf Diesel, Patented 1892

#### Iron to Steel



Steel produced by Bessemer process, patented 1856





2<sup>nd</sup> Industrial Revolution manufacturing assembly line and infrastructure of electricity, gas, water, telegraph, roads

Start of 20th century



December 17, 1903,

North Carolina.

Orville Wright piloted the first powered airplane in

Assembly Line, 1901 Olds patented the assembly line concept, which he put to work in his Olds Motor Vehicle Company factory in 1901.

Transportation



Karl Benz made four-stroke engine was used for automobile, 1886



Steam Locomotive, 1830 Electric Power



Virginia & Trucke "Tahoe", 1875

#### was used for automobile, 1880

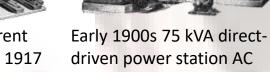
### **Electric generator**



earliest generator for industrial process, 1844



high-current dynamo, 1917 310A/ 7V

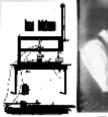


alternator



Nikola Tesla, Induction Electric Motor, 1888

#### Communication





Telegraph, 1836

Telephone, 1876

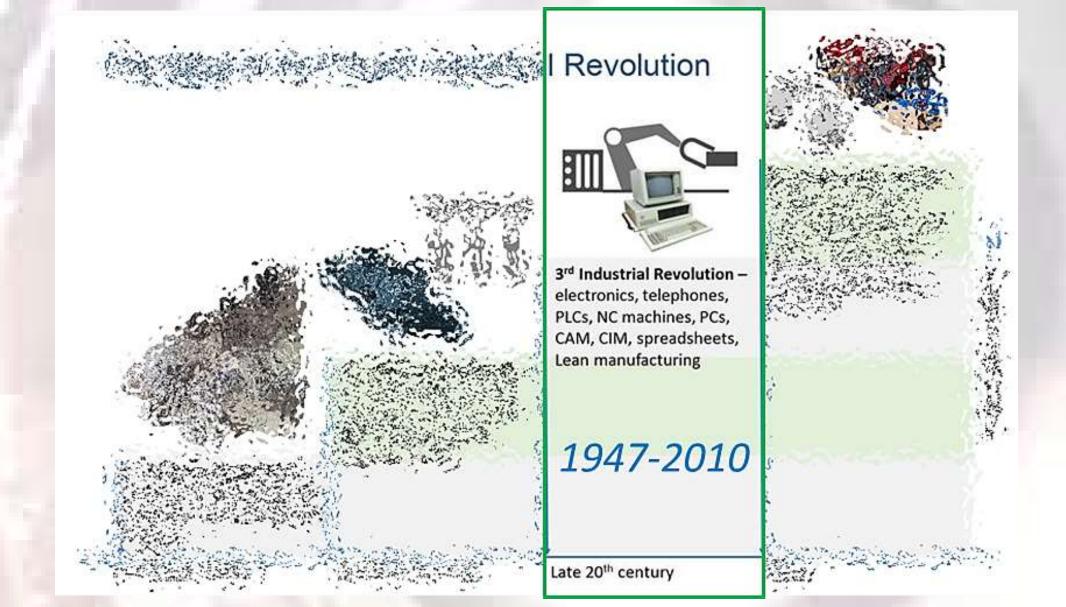




Motor Driving Machine tool

dynamo, first electrical generator power for industry

# 3<sup>rd</sup> Industrial Revolution—IT systems & digital technology overtaking analog & mechanical technology



# 3rd Industrial Revolution& Beyond—IT systems & digital technology overtaking analog & mechanical technology

26 offers text &

communications

protocols for G G III.

networks first deployed in Finland

browser for

CERN.

Mac created in

2nd generation digital cellular

2014 became global standard.

8Wastes

1988, John

Krafcik first

coined Lean

in a article

low data rate



Transistor Design Note, 1947



general-purpose electronic digital computer UNIVAC



Lab., 1947



February 15, 1946@ Univ. of Pennsylvania for the American army



programmed by punched paper tape, Harvard Mark

1948, RFID invited

by Harry Stockman

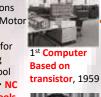
and companies

Developed in MIT





awarded "Motor Controlled apparatus for positioning machine tool control"  $\rightarrow$  NC machine tools





ASCI

for information for information to sendence

1959, APT for NC invited at

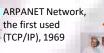




Code by MIT



and 1990 consumers 1952, first NC machine CNC with tool changer, 1959



expert system

1965, Integrated

1965: "Moore's

UNIMATE, First mass

robot at GM, 1961

Law" Predicts

the Future of

integrated

Circuits

DENDRAL@

Stanford

Circuit for

Amplifer

1968, Dick Morley introduced the first PLC, Modicon to the industrial market



to machine (M2M) devices combined telephony & computing

conceptualized by Theodore Paraskevakos

1965 Theo

Willimson patened FM



1971, world's first











1981. Univ. of Rhode Island awarded machine vision Robo





- Bluetooth Invented by Jaap Haartsen, in 1994

1994. OPC founded to secure data exchange in industi automation



Used since 1980s

1996 Compag

business plan is the

earliest known use

of the term "cloud

Grid Computing was

harness the power

of networked PCs to

solve CPU-intensive

1997, wireless M2M

became prevalent for

the specific needs of

different vertical

popularized to

problems

1999, "Internet of hings" was coined by Kevin Ashton

he term **GPU** was popularized by Nyidi 999, GeForce 256 is "the world's



2009, first-release LTE standard 4G was deployed in Oslo& Stockholm



begins to gain popularity after a papei by Geoffrey Hinton and Ruslan Salakhutdinov demonstrated that many-layered neural network could be pretrained one layer at a



appeared in 2001, allowing phones to access the internet.





2012, fog computing is



OpenFog





2016, Google DeepMind's algorithm AlphaGo beats Lee Sedol in Seoul





"Industrie 4.0" was



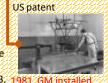


1969, first vision-based fully-automatic intelligent robot by Hitachi, Building

First industrial robot by blocks based on visual George Devol & Joseph image of assembly drawings, 1973 Engelberger, 1959



Martin Copper made the First handheld phone called, April 3,



Printing", MIT first developed in 1993 & Z Co. obtained exclusive license in

1992. Industrial

was introduced

Ethernet and TCP/I

connectively for PLCs

1992, the first SM5

"Merry Christmas"

was sent from Neil

Richard Jarvis in UK

Papworth to

markets such as automotive telematics

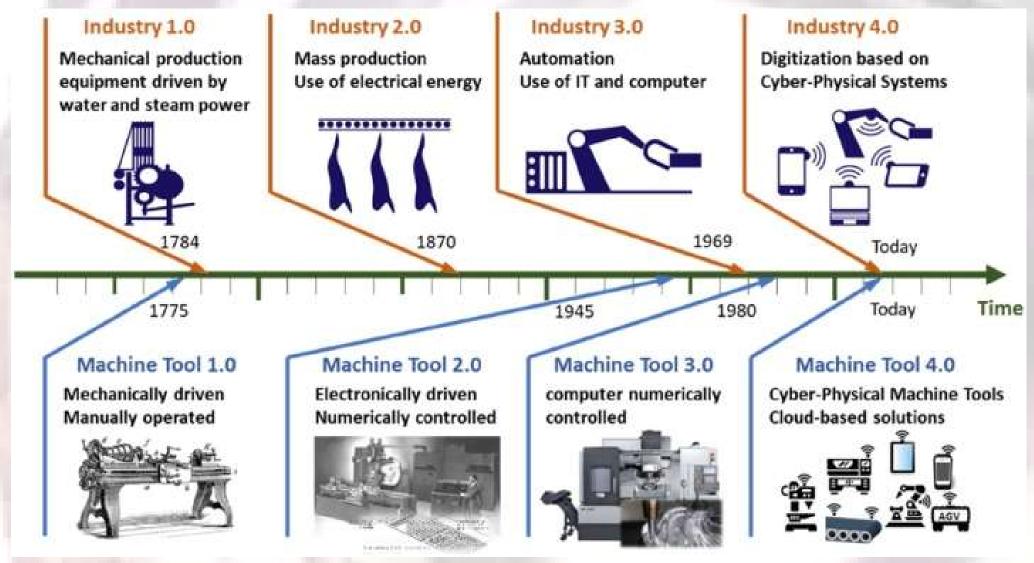
computing" is coined by Ramnath Chellappa in a

In 2000, GPRS opened as a packet-switched data service embedded to the channel-switched cellular radio network GSM

revived in 2011 at the Hannover Fair.



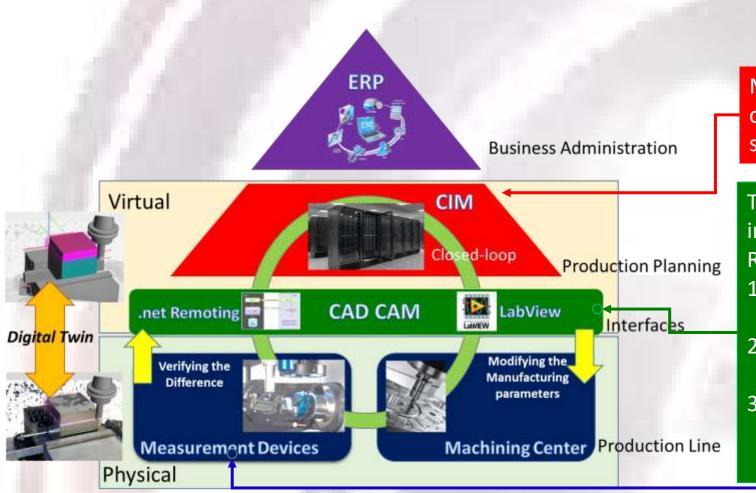
# Machine tools vs. Industry Revolutions



Chao Liu, Hrishikesh Vengayil, Ray Y. Zhong, Xun Xu, "A systematic development method for cyber-physical machine tools", Journal of Manufacturing Systems, 2018 (https://doi.org/10.1016/j.jmsy.2018.02.001)

## Industrie 4.0

"strategic initiative to establish Germany as a lead market and provider of advanced manufacturing solutions. Industrie 4.0 represents a paradigm shift from centralized to decentralized smart manufacturing and production. Smart production becomes the norm in a world where intelligent ICT-based machines, systems and networks are capable of independently exchanging and responding to information to manage industrial production processes."



Measurement processes are combined into current production planning and control systems (CIM) and ERP systems.

The measurement processes are integrated into CIM& ERP by LabView Interface, .net Remoting and CAD/CAM.

- 1. All the systems involved interact in a closed loop,
- 2. Be possible to intervene in production at any time,
- 3. Manufacturers can plan and manage production on a flexible basis based on the production measurement data

The role of optical measurement technology is to verify the Tolerances, Dimensions, Surface quality, & ....

#### Purpose/Goal:

Small batch quantities even with tight tolerances can be manufactured to a high level of precision & high quality

To make Smart Manufacturing work efficiently:

- measurement sensors shall be integrated into the production line
- The metrology equipment could be both capable of networking with current production systems and being expanded by the manufacturer

https://www.azom.com/article.aspx?ArticleID=13890

### Digital Twin for different levels Traditional modal: Rework, if the final **Digital Twin for Product** products cannot pass QC **Validation** Rework Verification Virtual Real Real Virtual **Products Productions Productions Products** Only Qualified commissioning Specification **Products allow Digital Twin** Data Evaluated from Digital Twin for manufacturing Performance Data Collected from Simulation based on

Lift time estimation/ maintain cycle operation conditions determined by data comparing

embedded sensors

# Digital Twin for different levels

Digital Twin for Product

Traditional modal:
Rework, if the final
products cannot pass QC

What we need to have—

Validation

Rework

- 1. On Machine Measurement technologies/ Metrologies
- 2. Computing Capability
- 3. Well developed (wireless) Communication Environment
- 4. Prices for Sensors must be low
- 5. Capability for managing big data

• • • •

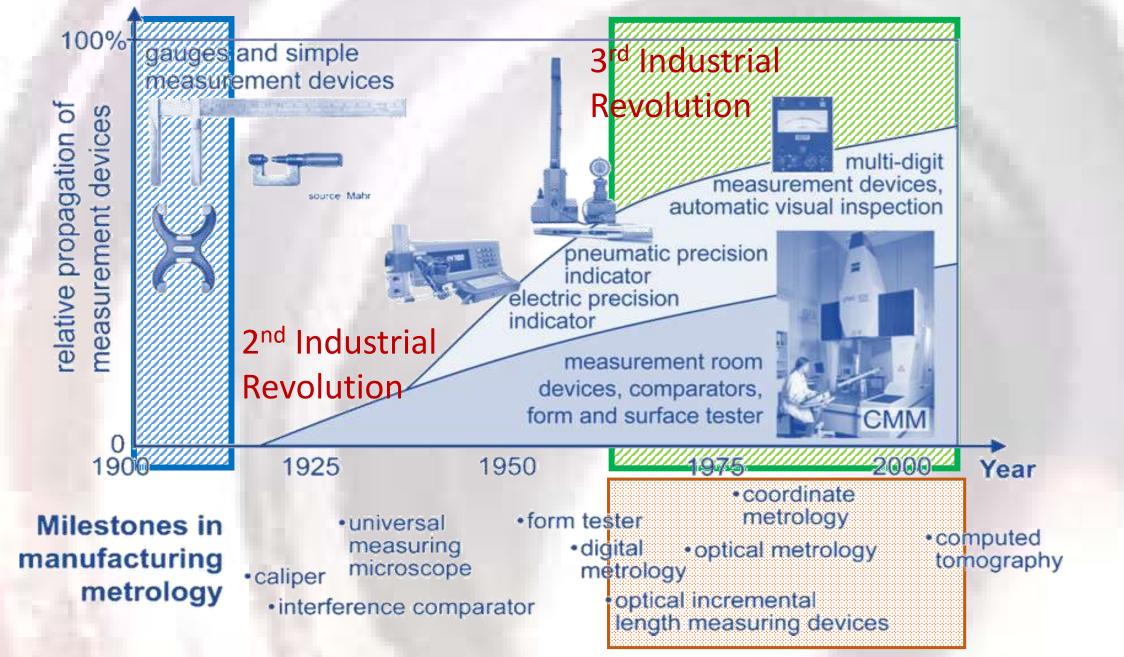
Only Qualified Products allow

Performance

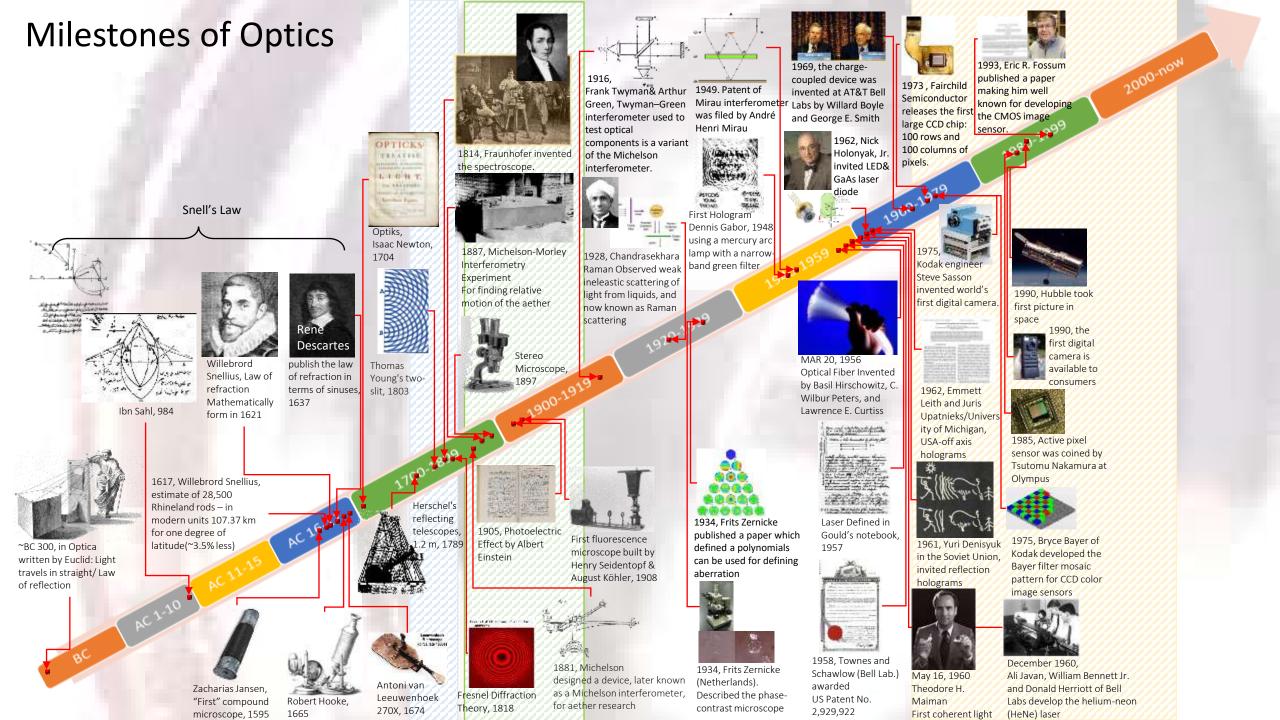
Simulation based on operation conditions

Lift time estimation/ maintain cycle determined by data comparing

embedded sensors



Weckenmann, A., & Krämer, P. (2009). Application of Computed Tomography in Manufacturing MetrologyAnwendung der Computer-Tomographie in der Fertigungsmesstechnik. Tm - Technisches Messen, 76(7-8). doi:10.1524/teme.2009.0969



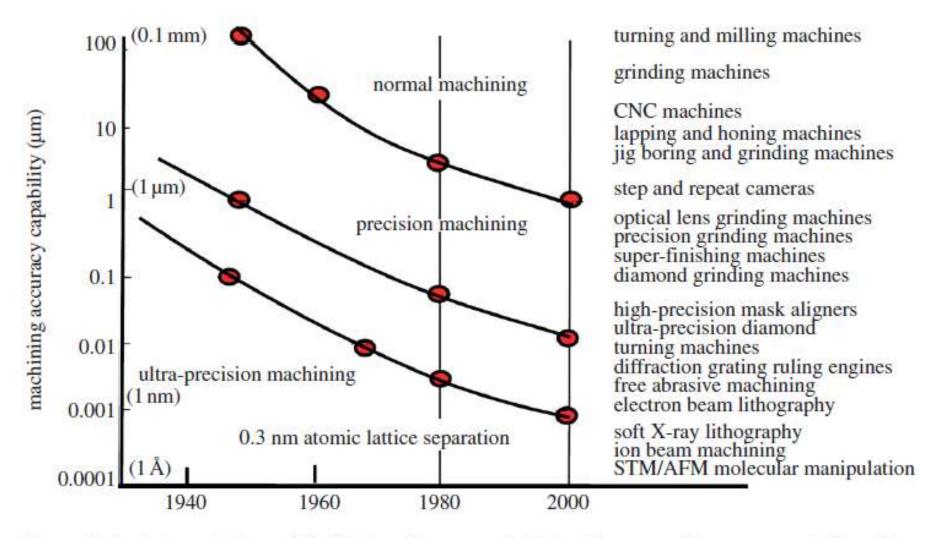
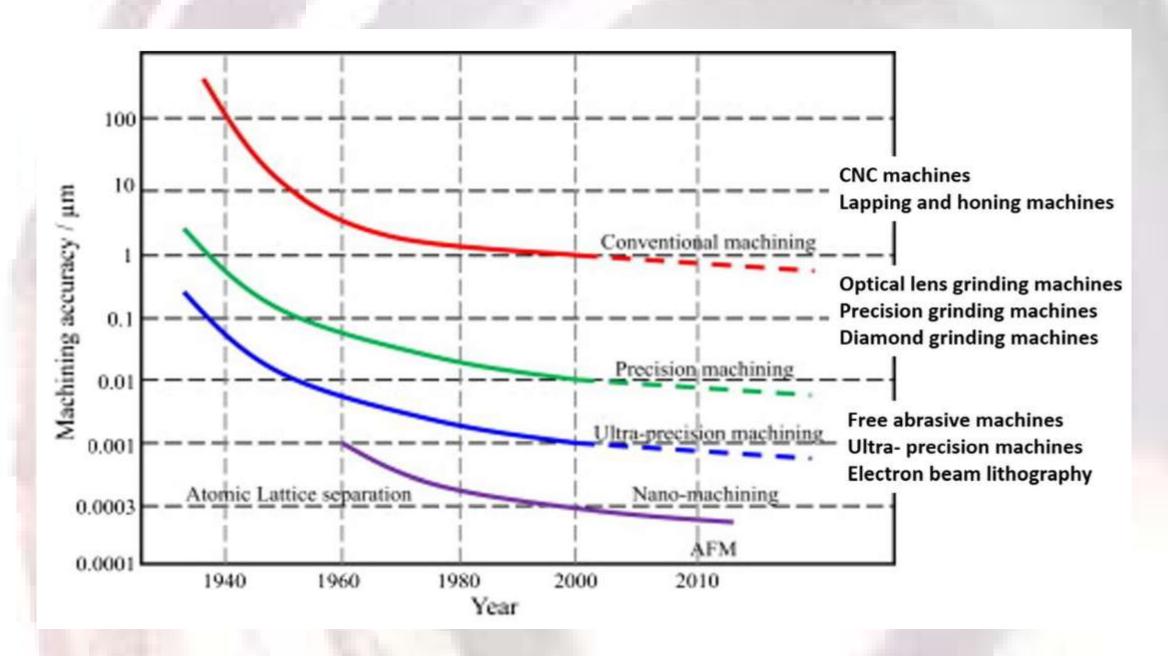


Figure 3. An interpretation of the Taniguchi curves, depicting the general improvement of machine accuracy capability with time during much of the twentieth century.



# General Descriptions of Optical Methods

- None-Contact vs. Contact
- Full Field vs. Line/ Point
- High Resolution— Spatial? Time domain?
- High Sensitivity—noise sources/ how to filter the noise

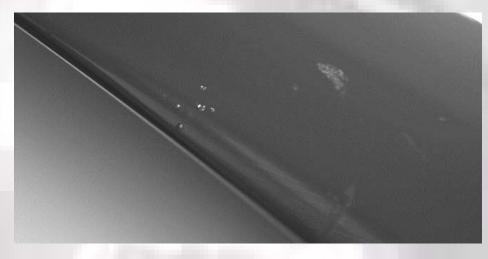
# Highly Reflective Surface Defects



Scratch



**Combination Defects** 



**Pitted**