

# embedded **VISION** SUMMIT 2018

## **Data-Driven Business Models Based on 3D Technologies**



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22.5.2018

# We have provided imaging solutions for 36 years



## More than 36 years of imaging experience

- Founded in 1981
- Employees: 150
- Sales Volume ~ 50 Mio. EUR
- The only global imaging partner from sensor to system
- Development of our own embedded technologies
- Team of 40 engineers

# Technology revolutions always bring winners and losers



SOURCE: Andrew H. Walker / Stringer / Getty; Twitter

Source: [Link](#)

**NETFLIX**



SpencerStuart  
Kienbaum<sup>K</sup>



# Hypothesis: Currently, winners gain competitive edge based on data

## Def\_Business Model

A business model describes the rationale of how an organization creates, delivers, and captures value



Source: A. Osterwalder & Y. Pigneur (2010), [Link](#)

## Winners in the world of data:

- 1. Apply and combine technology**  
that enables better or new solutions
- 2. ... that generate data ...**
- 3. ... and, in turn, provide a significantly better user experience (competitive advantage)**

# Data provides the competitive edge



		
Application of technology	Streaming of audio content / cross-device / any time	Network to promote oneself professionally
That generates data	Data on which content is streamed where, by whom and on which device	Career path of person profiles with company profiles
Significantly better user experience	Optimized content suggestions for customers, market insights to artists	Prospect suggestions with much higher success rate

# 3D is a key technology to discover such business models

## 3D data

Depth Map



Point Cloud



Voxels



## Methods

Object Recognition

Obstacle Recognition

SLAM (Navigation)

Pose Estimation

Dimensioning

## Applications

AR/VR

Robotics

Drones

Logistics/Industry

Automotive

Surveillance

Gesture control



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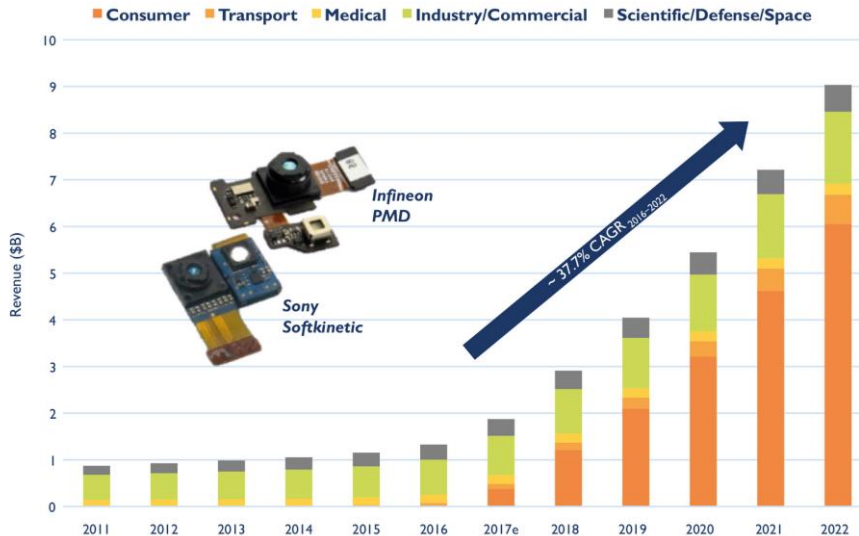


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# Consumer applications will drive 3D market growth

## 2011 - 2022 market forecast for 3D imaging & sensing devices

(Source: 3D Imaging & Sensing 2017 report, April 2017, Yole Développement)



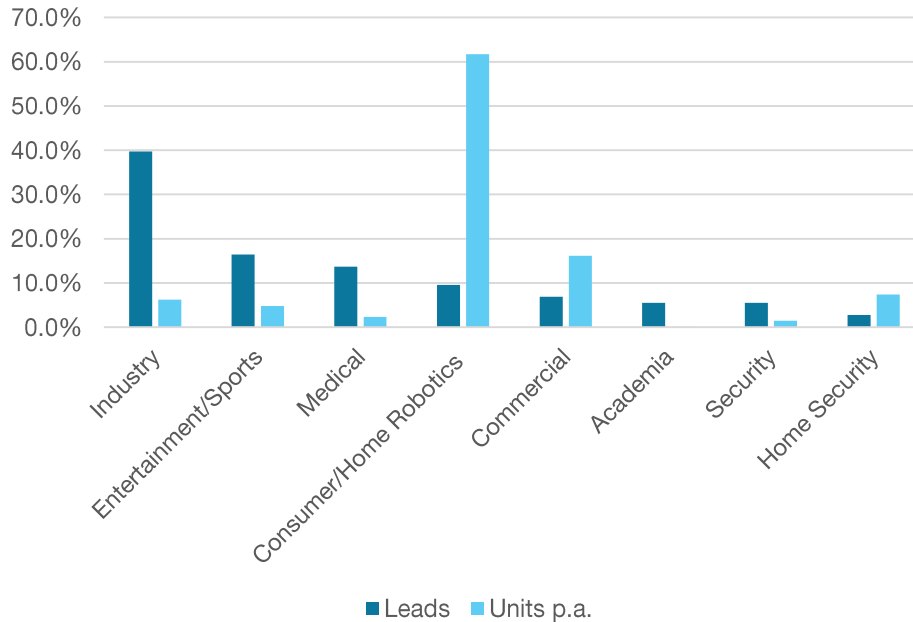
Source: [YOLE](#)

- 3D imaging and sensing devices are expected to grow at a CAGR of 38% until 2022
- Main absolute growth will come from consumer as well as industrial/commercial applications
- Rise of 3D is spurred by technology progress and declining prices



# We recognize a lot of 3D opportunity in various industries

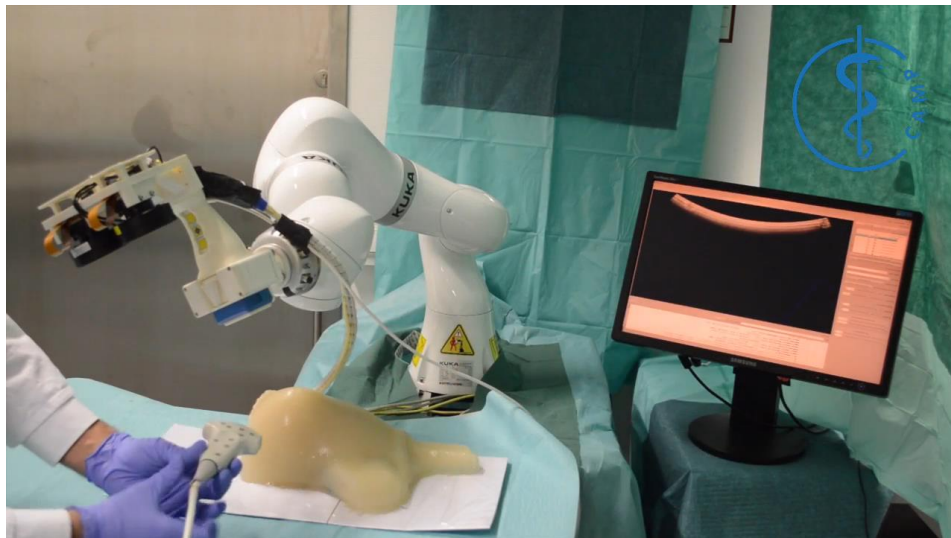
FRAMOS 3D leads and units p.a.



- Most of the leads in 3D that we receive come from the industrial field (factory, farming, logistics automation)
- However, concerning the volume measured in units p.a., the most relevant sectors are home robotics and commercial applications



# 3D enables collaborative robotics



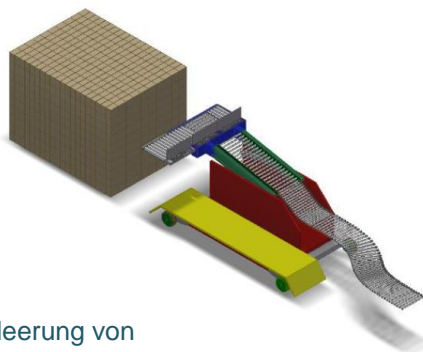
FRAMOS Installation, Munich Hospital, Rechts der Isar

**Technology:** 3D camera with marker-based tracking (pose estimation)

**Application:** Collaborative robotics in the area of medicine and industry

**Edge:** Database for pose-estimations and use of AI to improve accuracy

# 3D enables automatic container unloading



IRiS – „Interaktives Robotiksystem zur Entleerung von Seecontainern“

**Technology:** 3D camera for depth calc. and dimensioning, semantic segmentation to distinguish objects

**Application:** Unloading of trucks and containers

**Edge:** Data to optimize the system accuracy, data on flow of goods

# 3D enables autonomous navigation (I/II)



B. Busam, P. Ruhkamp et al. [under review], FRAMOS

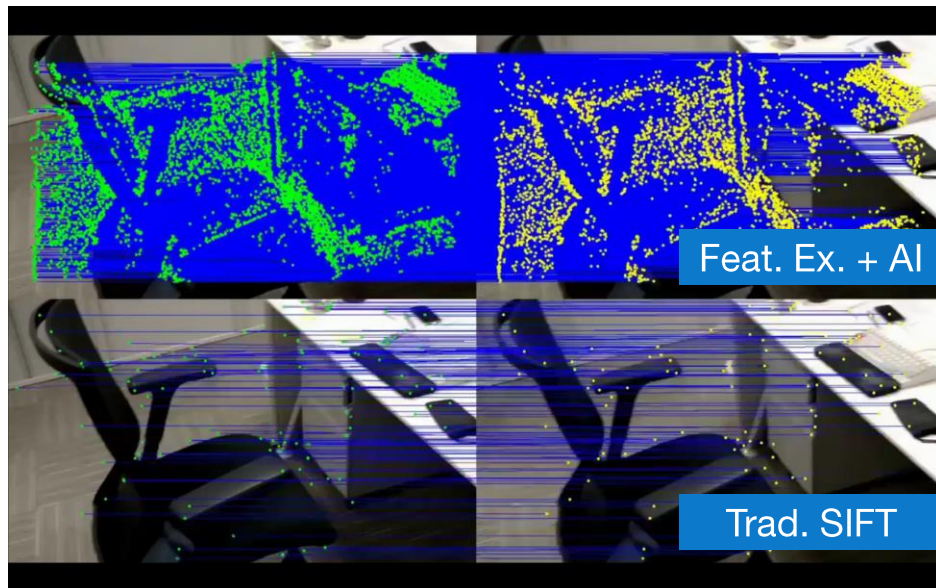
**Technology:** SLAM + AI to create 3D maps

**Application:** Navigation for UAVs, home robots, logistics robots

**Edge:** 3D maps of the world, outside and inside

# 3D enables autonomous navigation (II/II)

SLAM + AI provides  
technological edge



B. Busam, P. Ruhkamp et al. [under review], FRAMOS

**Technology:** SLAM + AI to  
create 3D maps

**Application:** Navigation for  
UAVs, home robots,  
logistics robots

**Edge:** 3D maps of the  
world, outside and inside

- 1. Apply 3D technologies to your application for collaborative robotics, navigation, dimensioning ... etc.**
- 2. Generate a 3D database (and make use of AI)**
- 3. Discover a business model that uses the database to create a significantly better user experience**

Visit the FRAMOS  
Technology Showcase

FRAMOS Imaging Center <https://www.framos.com/en/imaging-center/> (Case Studies, Specialized Articles, Whitepapers & Expert Interviews)

Stereo Vision: [Facing the Challenges and Seeing the Opportunities for ADAS Applications](#), Embedded Vision Alliance Resources

[Vision Processing Opportunities in Drones](#), Embedded Vision Alliance Resources

"[The Evolution of Depth Sensing: From Exotic to Ubiquitous](#)," a Presentation from 8tree, Embedded Vision Alliance Resources

"[How 3D Maps Will Change the World](#)," a Presentation from Augmented Pixels, Embedded Vision Alliance Resources

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