

# User-Centered Visual Analysis of PCAP Data

Dr. Alex Ulmer

#### Introduction





Dr. Alex Ulmer

Research: User-Centered Data Science and Progressive Visual Analytics

**Application Area**: Cybersecurity

Visual Analysis of PCAP Data

#### Agenda



- 1. Introduction & Agenda
- 2. Fraunhofer Applied Research
- 3. Data Science and Visual Analytics
- 4. Visual Analysis of PCAP Data -Live
- **5.** Application Classification of Packets
- 6. Explainable Al
- 7. Visualization Plugin for Wireshark



6 min /

1 min

- 15 min
- 7 min
- 9 min
- <sub>3</sub>5 min



# 2. Fraunhofer – Applied Research

#### **Fraunhofer Mission**

1. Introduction & Agenda





Explore research ideas and get them market ready

Collaborate with businesses Create an advantage in the market

#### Fraunhofer Institute



#### 76 Institutes

@ 192 locations in Germany

#### Topics:

- Digitalization
- Energy
- Health and Medicine
- Mobility and Transport
- Work and Management
- Mechanics and Production
- Society and Security





2. Fraunhofer - Applied Research

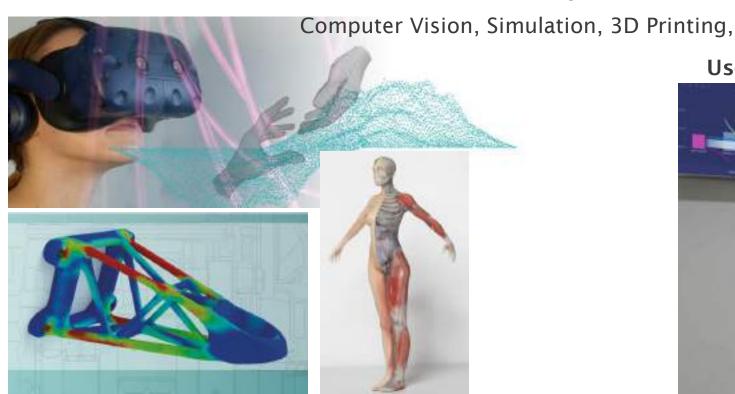
# Fraunhofer Institute for Computer Graphics Research





# Fraunhofer

40 Years of Computer Graphics Research



1. Introduction & Agenda

#### **User Centered Data Science and Visual Analytics**





# 3. Data Science and Visual Analytics

## Why Visualization?



10

A is connected to B

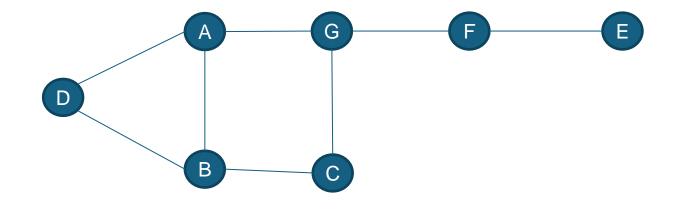
B is connected to C

D is connected to A and B

E is connected to F

G is connected to A and C

F is connected to G



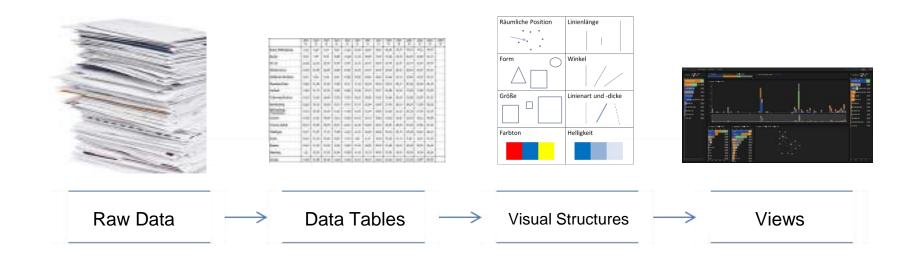
#### Data Science and Visual Analytics





#### From Raw Data to Visualizations

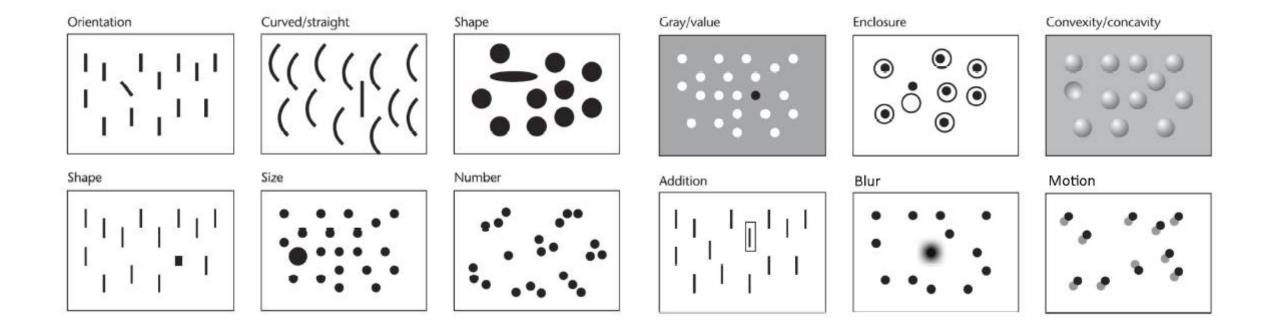




adapted from Card et al. 1999

#### **Human Perception for Visualization**

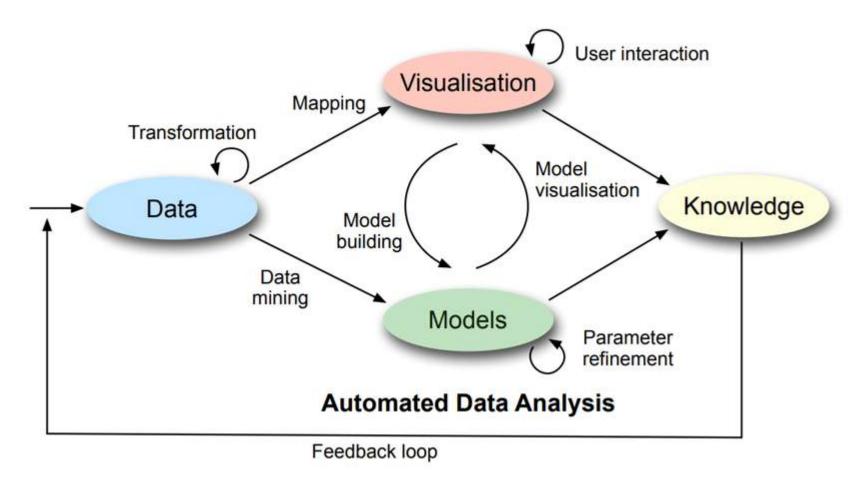




# **Data Science and Visual Analytics**



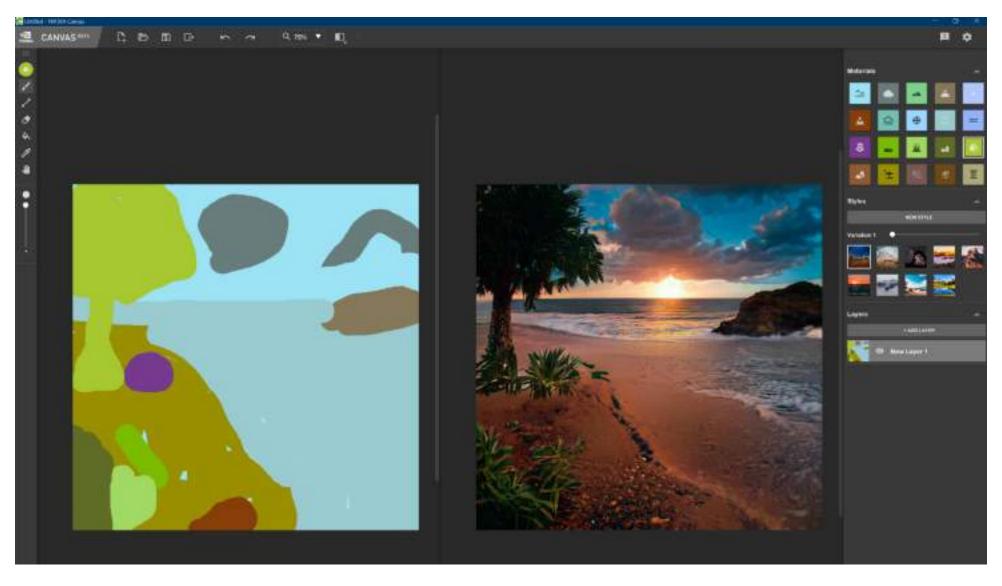
#### Bridging Data, Humans, and Artificial Intelligence



Visual Data Exploration, adapted from Keim et al. 2008

#### **Interactive AI - NVIDIA Canvas**

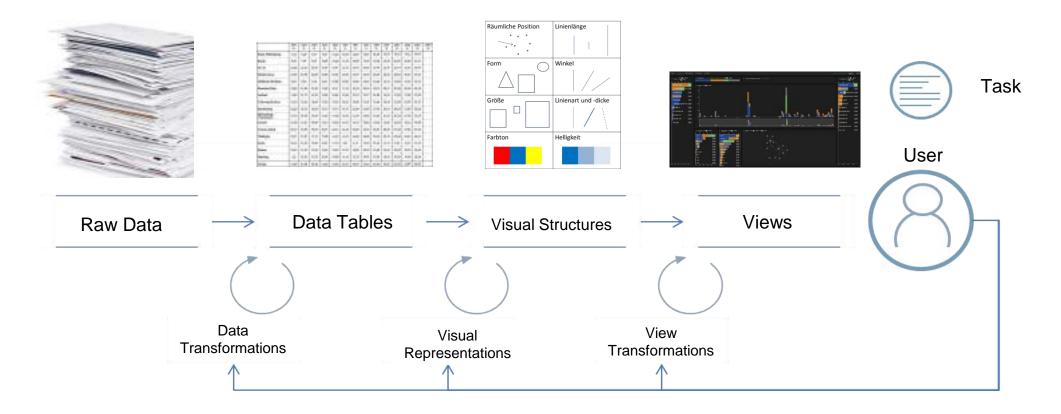




#### **User Centered Design**



## Change of direction - Which task needs to be completed?



adapted from Card et al. 1999



# 4. Visual Analysis of PCAP Data - Live

## **NetCapVis**



Why did we start NetCapVis?



Wireshark is fantastic for experts



But what about beginners? Can we make it easier to use?

3. Data Science and Visual Analytics



#### NetCapVis - Live



#### https://netcapvis.igd.fraunhofer.de



# NetCapVis



#### Wireshark filter export



## NetCapVis - Parsing and Storage



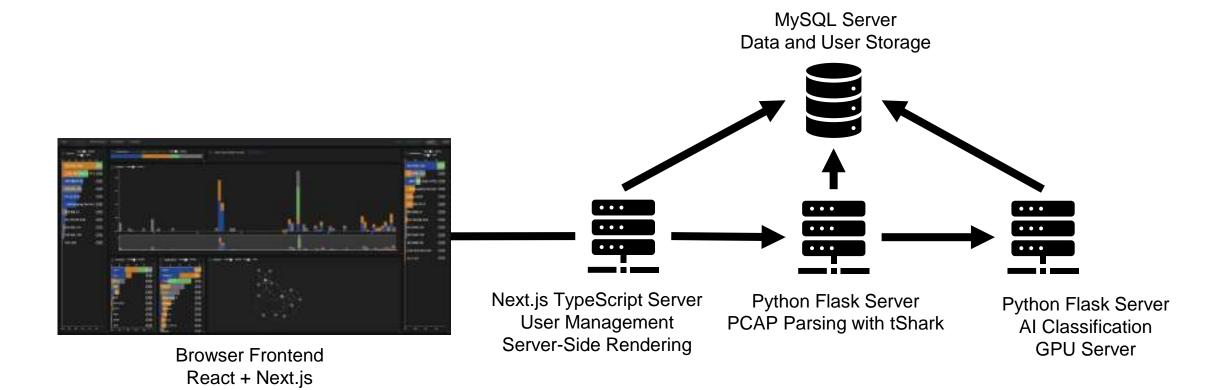
#### Parsing

- From Java Backend to Python with tShark
- DNS resolving included
- Working on GeoIP
- tShark optimizations needed

MySQL database to access uploaded data

#### **NetCapVis - Architecture**





#### **NetCapVis – Future Work**



Progressive Data Analysis Approach -

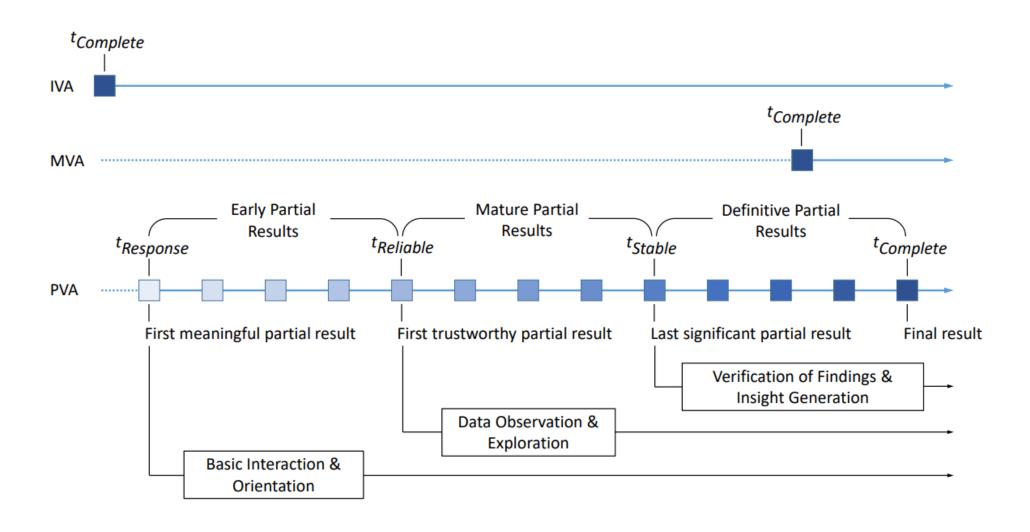
Load large captures

Load streams of data

with interactive visualizations at all times

### Progressive Visual Analytics





Angelini, Marco, et al. "A review and characterization of progressive visual analytics" 2018.

# **Progressive Visual Analytics Example**





## Progressive Visual Analytics Challenges



Chunking a PCAP file is difficult cutting and merging leads to corrupted data

Priority sampling for the filter that the user is setting



# 5. Application Classification of Packets

#sf25eu

#### **Machine Learning Basics**



#### Machine Learning (ML)

uncover hidden relationships in multidimensional data

#### Goals

- extract meaningful patterns from data
- make decisions based on those discoveries
- make predictions with AI systems

5. Application Classification of Packets

#### **Machine Learning Basics**



#### **Unsupervised Learning**

- Find structure in unlabeled data
- For PCAPs: Distinguish between normal traffic vs anomaly

## Supervised Learning

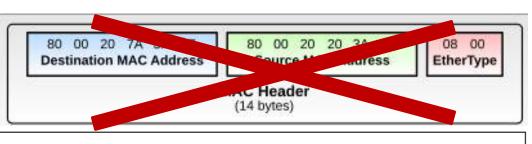
- Learn from labeled data (prior knowledge)
- For PCAPs: Define specific classes, gather training data and label it

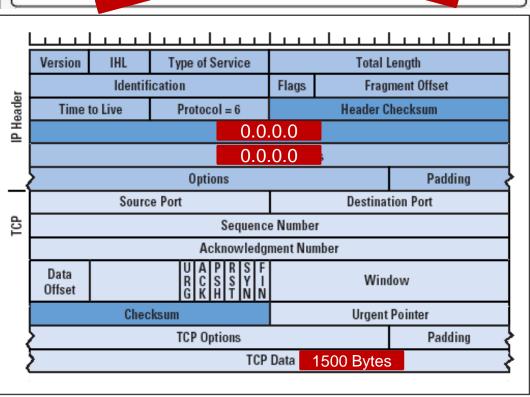
4. Visual Analysis of PCAP Data - Live

#### **PCAP Data Preparation**



- 1. Remove small packets (SYN/ACK, ...)
- 2. Remove ethernet header
- 3. Set src and dst address to 0.0.0.0
- 4. Cut or pad payload to 1500 Bytes
- 5. Similar approach for other protocols





M. Lotfollahi, et al. Deep packet: A novel approach for encrypted traffic classification using deep learning 2017. Dataset: ISCXVPN2016

### **Training Data and Classes**



#### Goal:

Predict which application caused a packet

Training data for the AI system

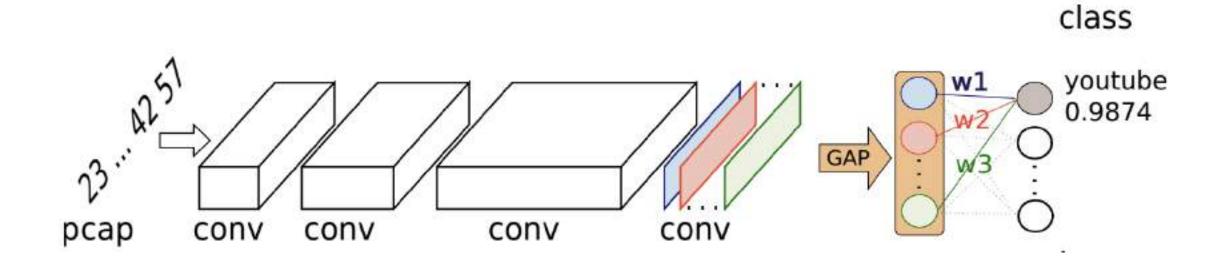


Application	Size (IPv4)	Size (IPv6
Big Blue Button	122672	150474
Email	66564	52041
Facebook Video	111950	62030
FTPS+SFTP	103701	50589
Google Meet	90214	55059
Amazon Prime Video	123329	174930
Reddit	119221	157737
Telegram Files	42252	41187
TikTok	128739	138929
Twitch	105531	138029
Vimeo	129652	177734
Youtube	139248	179525
Zoom	53029	114624
Instagram	69872	153085
Facebook Feeds	117894	93891

#### **Al Architecture**



#### Convolutional Neural Network (CNN)



# **NetCapVis AI Predictions**

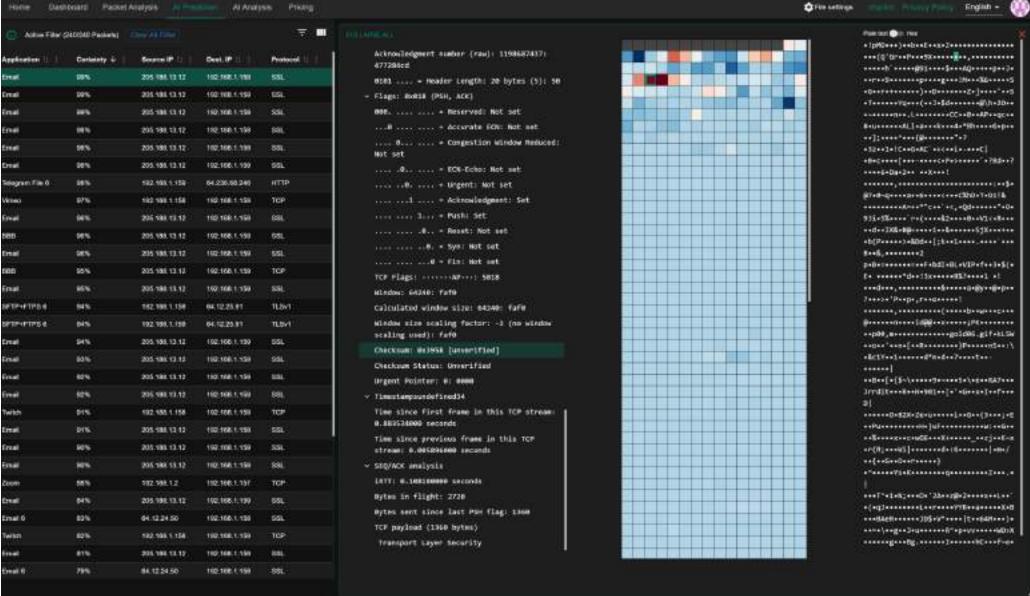


Adm Film (M0000 Periods)				
Application 1. 1	Certainty 4	Source P 1	Dest. P L	Protocol II I
treat.	50%	205 188 11 12	102 168 1 199	SSL
Creat	20%	205.188.13.12	102:100.1-159	SSL
Front	SOL.	205 188 11.12	192,165 1,159	551
Emel	88%	205.189.13.12	192,166,1-199	88L
tinell	56%	205 588 11 12	102166.1 119	88L
Ernal.	56%	205.160.13.12	102,166,1,199	SSL
lelegram File 6	10%	192,189,1-159	64.256.69.246	илте
Veneo	57%	192 100 1 158	100 160 1 159	TCP
Ereal	10%	205.588.13.12	100,168,1,169	881.
188	56%	205.188.13.12	192,168.1.198	85L
tmal	tes.	205 188 13.12	1021681.199	SSL
100	95%	205 188 13 12	192-160-1-150	TCP
Enal	65%	205.588.53.12	192,168,1-159	SSL
SETEMPTED 6	14%	192,168,1,158	84.12.25.81	TLBVT
orte-error e	54%	192,168,1,116	64.12.25.91	TLBv1
Ernal	94%	205.188.13.12	100:168.1.159	55L
Ernell .	80%	305.189.13.12	192,160.1.159	55L
Creat	10%	205 188 13 12	190,160,1-159	MIL
Email	60%	205.183.11.12	150,168.1.198	881.
Twitch	51%	192,188,1.158	150,166 ( 159	TOP
Imal	01%	305 188 13 12	152 168 1 150	SSL
(real	90%	205.188.13.12	192,100.1.159	551
Email	10%	205 188 11 12	190 166 1 156	580.
Zoom	88%	792.189.1.2	110.166.1.197	TOP
Crest	84%	205,188,13.12	100,168,1,189	SSL
Email 6	80%	64.12.24.00	192,160,1,158	SSL
Teiton	50%	192 193 1 156	192,160,1.158	TOP
treat	81%	205 188 13.12	100 166 1 159	100
Const II	76%	84 12 24 50	102 168 5 158	884

35

#### **NetCapVis AI Predictions**



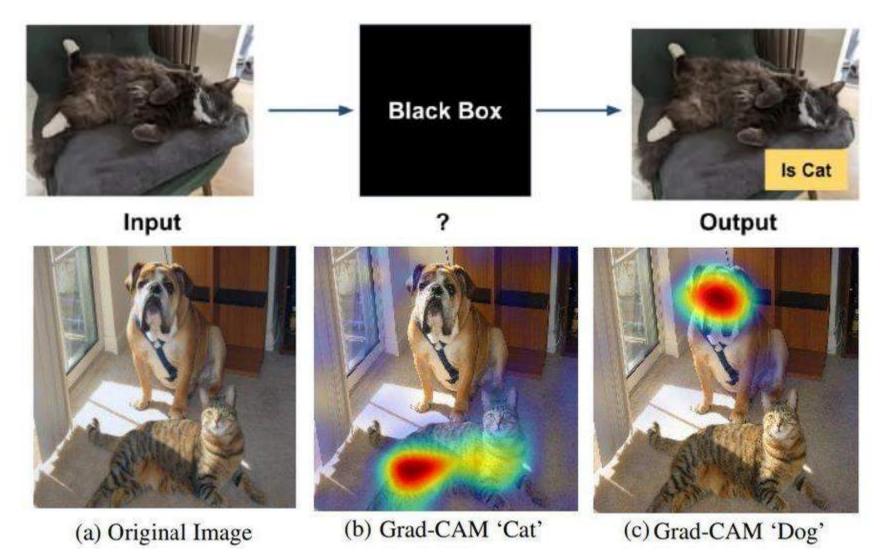




# 6. Explainable Al

#### What is XAI?

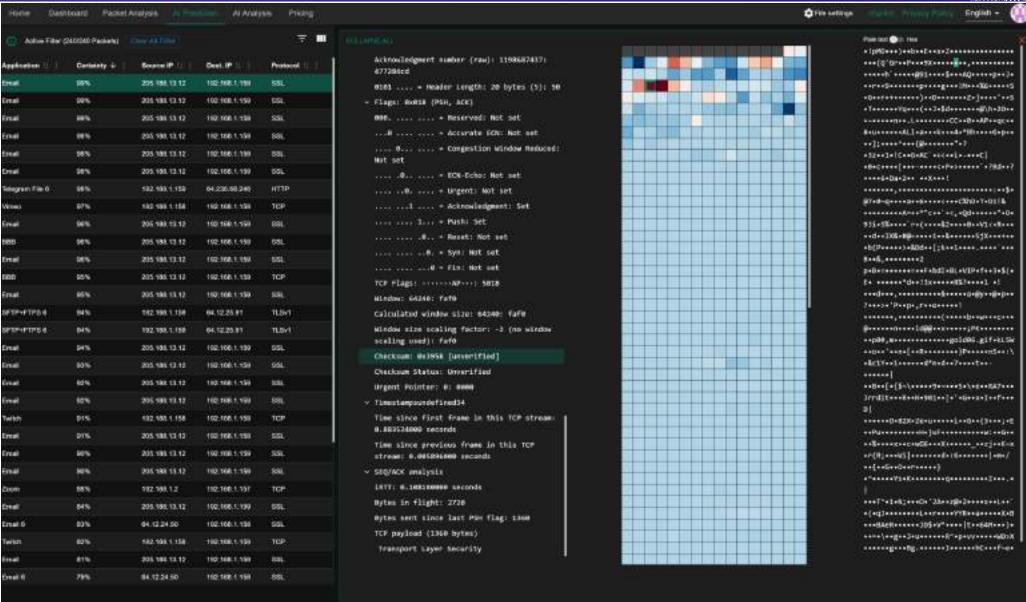




Trustworthy versus Explainable AI in Autonomous Vessels, Glomsrud et al. (2019)

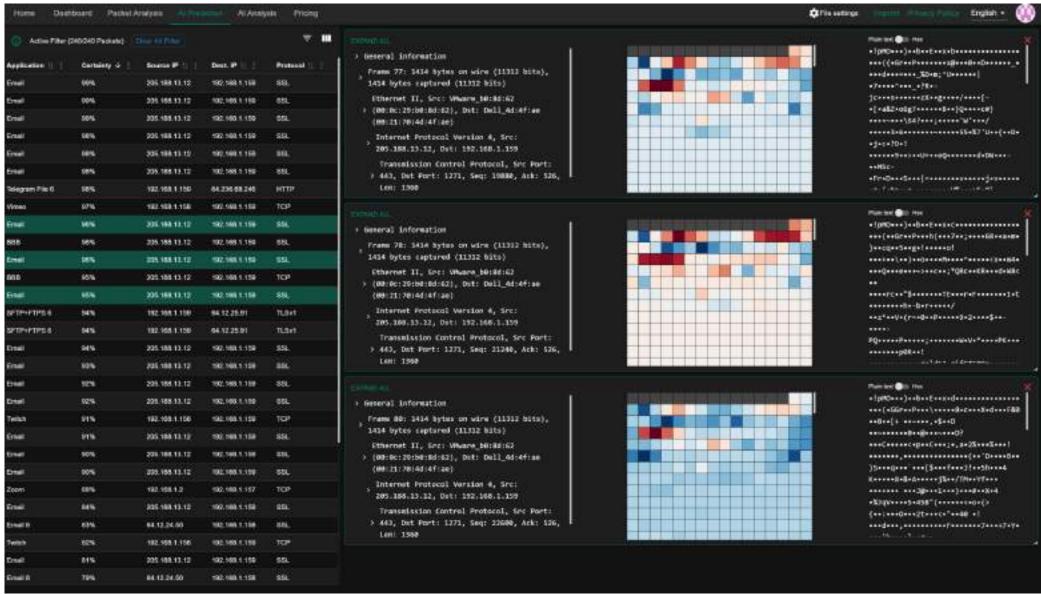
#### **Local Class Activation Maps (CAMs)**





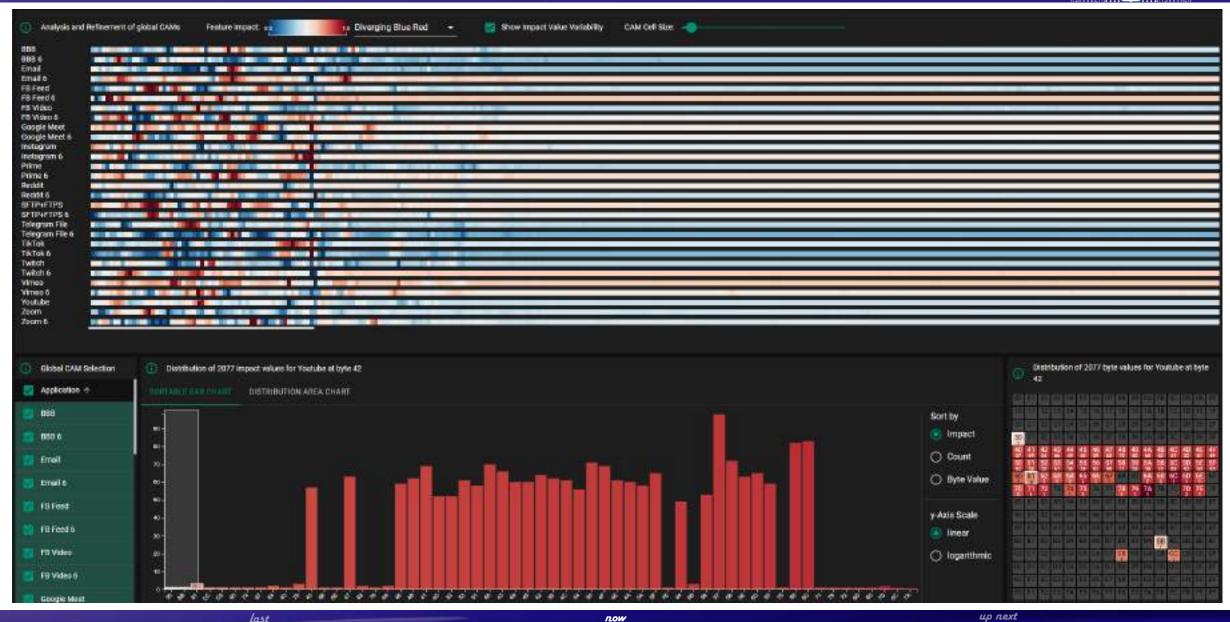
# **Local Class Activation Maps (CAMs)**





# **Global Class Activation Maps (CAMs)**





# **Data Drift and Active Learning**





# **Insights and Future Work**



### **Insights**

- Enrcypted payload has almost no contribution to classification
- · Training data preparation requires more expertise
- Sequence and ack number need to be cut
- Variable length of TCP options

#### **Future Work**

- · Feedback system: network expert corrects the AI system
- Aggregate bytes to acutal fields of the protocol
- · Active Learning: split or combine application classes

# **Bridging Expertise between Communities**



# Bringing visualization, machine learning and network experts together

#### Come talk to us!



Alex



Igor



David



Felix

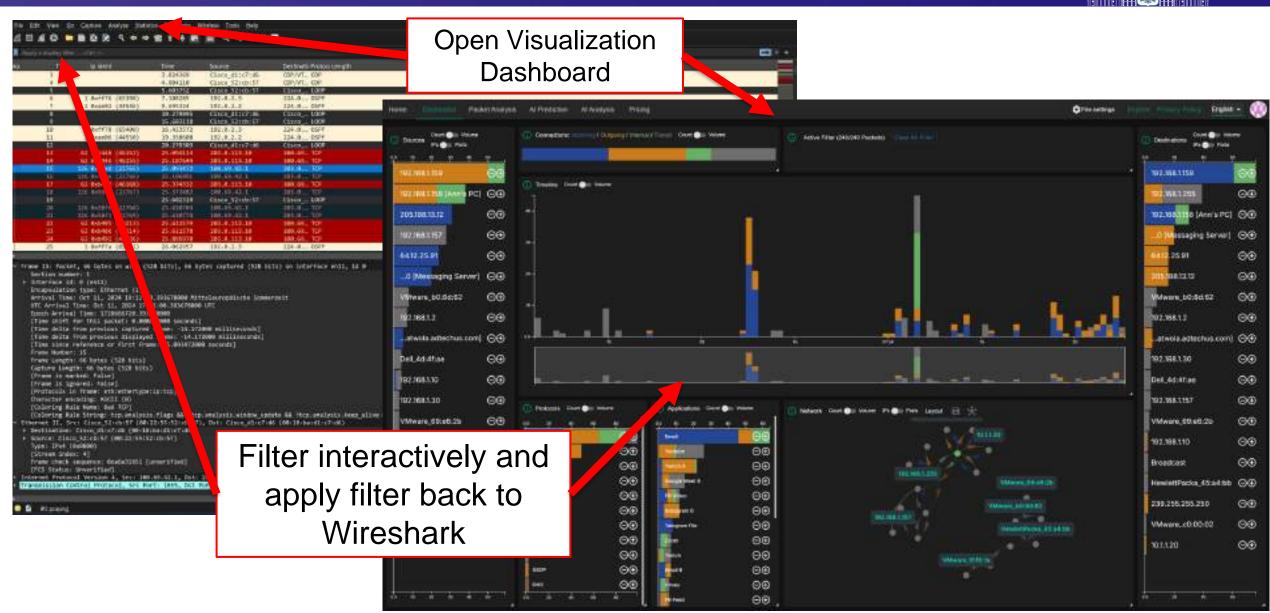


# 7. Visualization Plugin for Wireshark

#### Idea: Vizualization Dashboard

6. Explainable Al







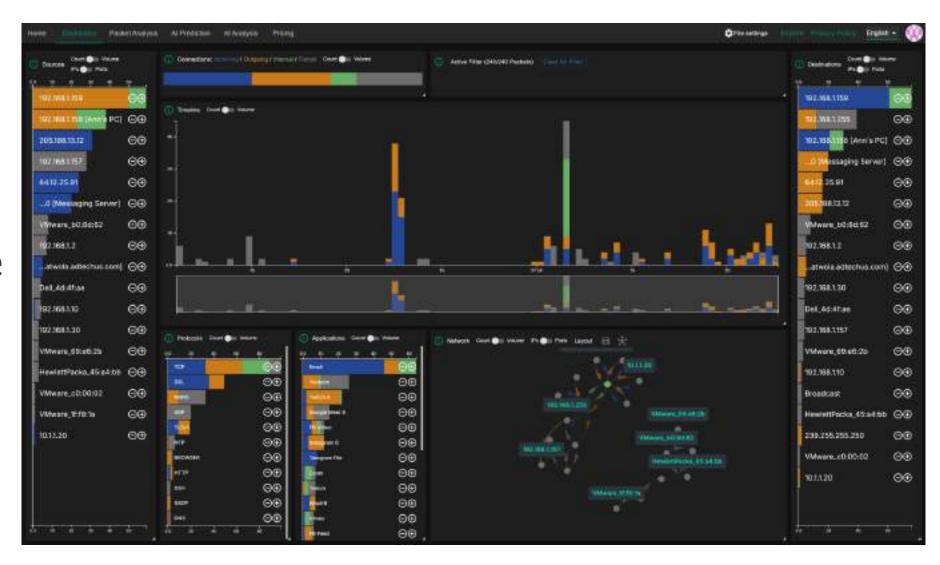
#### Idea: Vizualization Dashboard



Interacitve Dashboard only

No Packet Table
No Al Functions

For free



6. Explainable Al

### **Option A: Cloud-hosted Dashboard**



#### Reduced Packet Header Data







filter query

(frame.time >= "Aug 13, 2009 07:57:35") && (frame.time <= "Aug 13, 2009 07:58:06") && (ip.dst == 192.168.1.159) && (tcp || udp)



### **Option B: Local Dashboard**



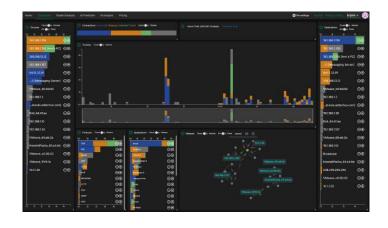
#### Reduced Packet Header Data





filter query

(frame.time >= "Aug 13, 2009 07:57:35") && (frame.time <= "Aug 13, 2009 07:58:06") && (ip.dst == 192.168.1.159) && (tcp || udp)



# **Options: Vizualization Dashboard**



# Option A: Cloud-hosted Dashboard

- easier integration
- no extra install
- internet connection required
- data upload

# Option B: Local Dashboard

- data stays on the device
- no service dependency
- more install requirements
- start own localhost service

# What do you think?



Would you use this plugin?

Which option would you perfer?

Are there more options?

6. Explainable Al

#### **Contact and Feedback**



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#### LinkedIn:





#### SharkFest Feedback Form

