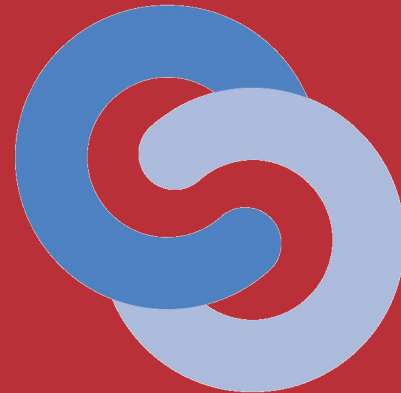


# Relational Database Similarity Detection via Network Traffic Analysis



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# Research objective

**Present and evaluate methods for measuring similarities between complex software systems where there is no access to the internal structure of the system , such as access to the source code or even compiled server binaries.**



# Similarity Detection

## Techniques

- Traditional approaches
- Software Fingerprints
- Software Birthmarks
- Code Embeddings
- LLM-based

## on top of ...

- code (in programming language)
- abstract syntax tree
- code (in binary language)
- profiling data
- sampling data
- OS data

# Similarity Detection (via Network Traffic Analysis)

## Techniques

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- Software Fingerprints
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- Code Embeddings
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## on top of ...

- ~~—code (in programming language)—~~
- ~~—abstract syntax tree—~~
- ~~—code (in binary language)—~~
- profiling data
- sampling data
- OS data (network card)

# Available monitoring tools



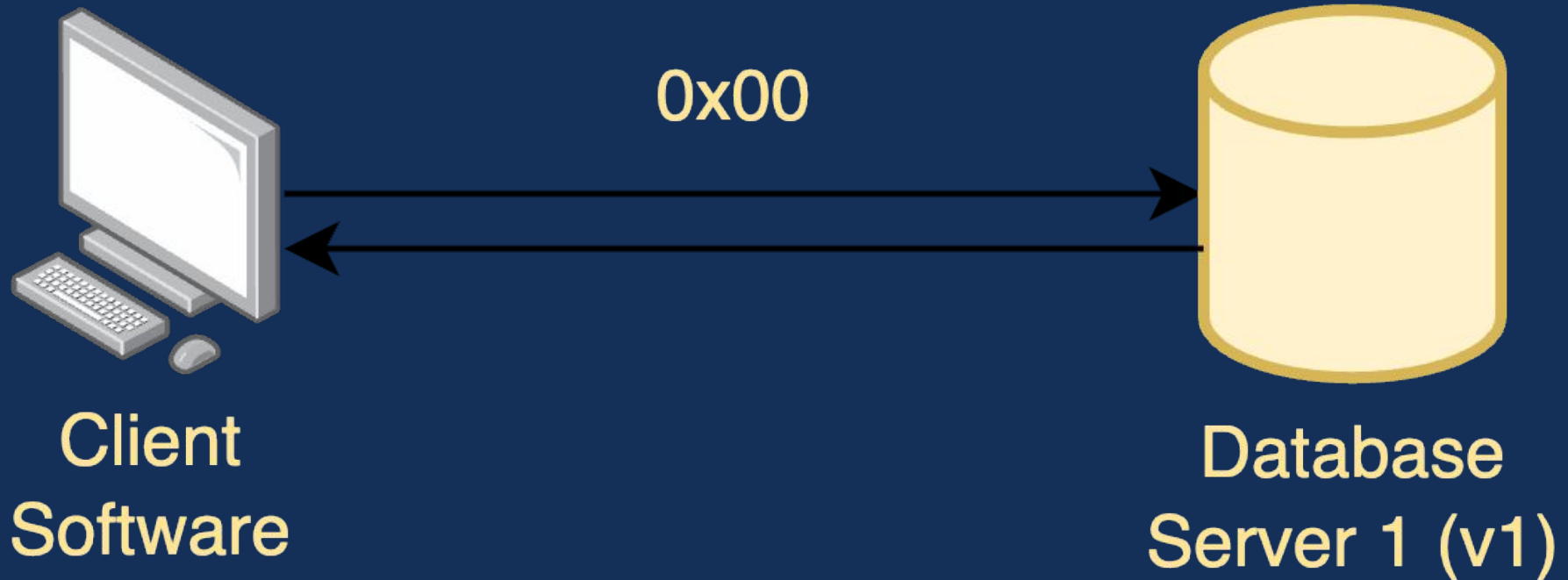
Client  
Software



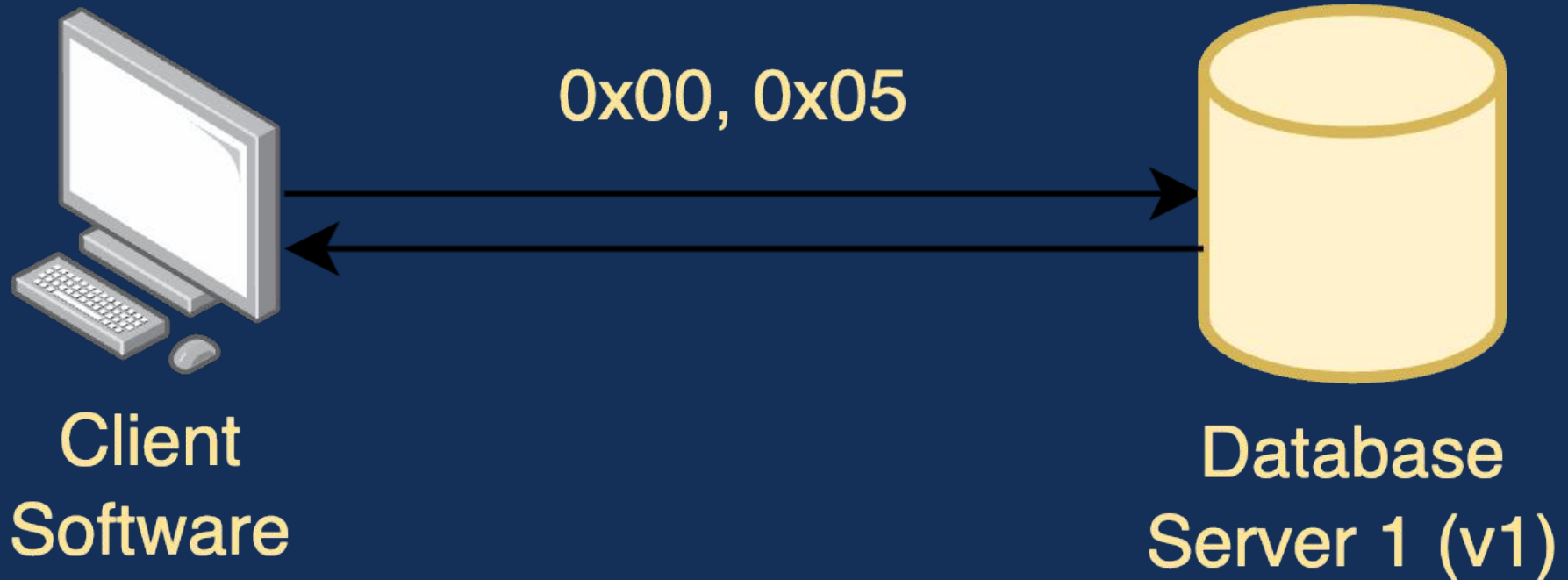
Database  
Server 1 (v1)



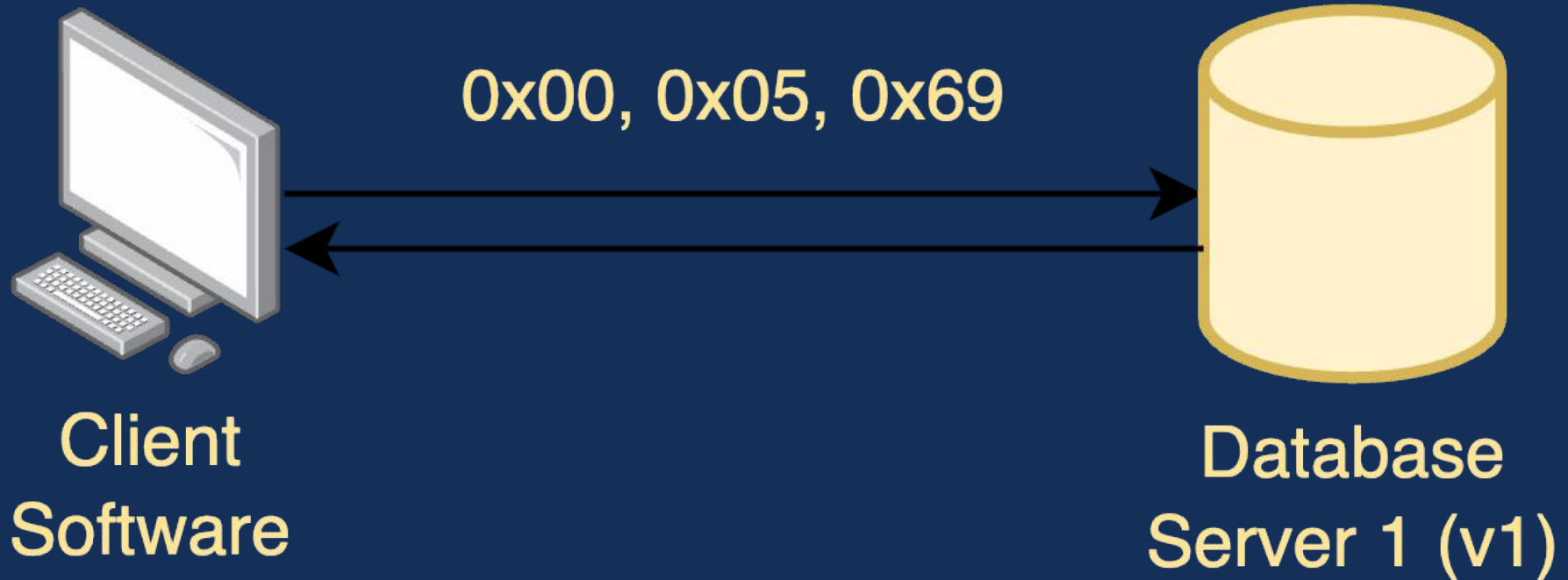
# Available monitoring tools



# Available monitoring tools

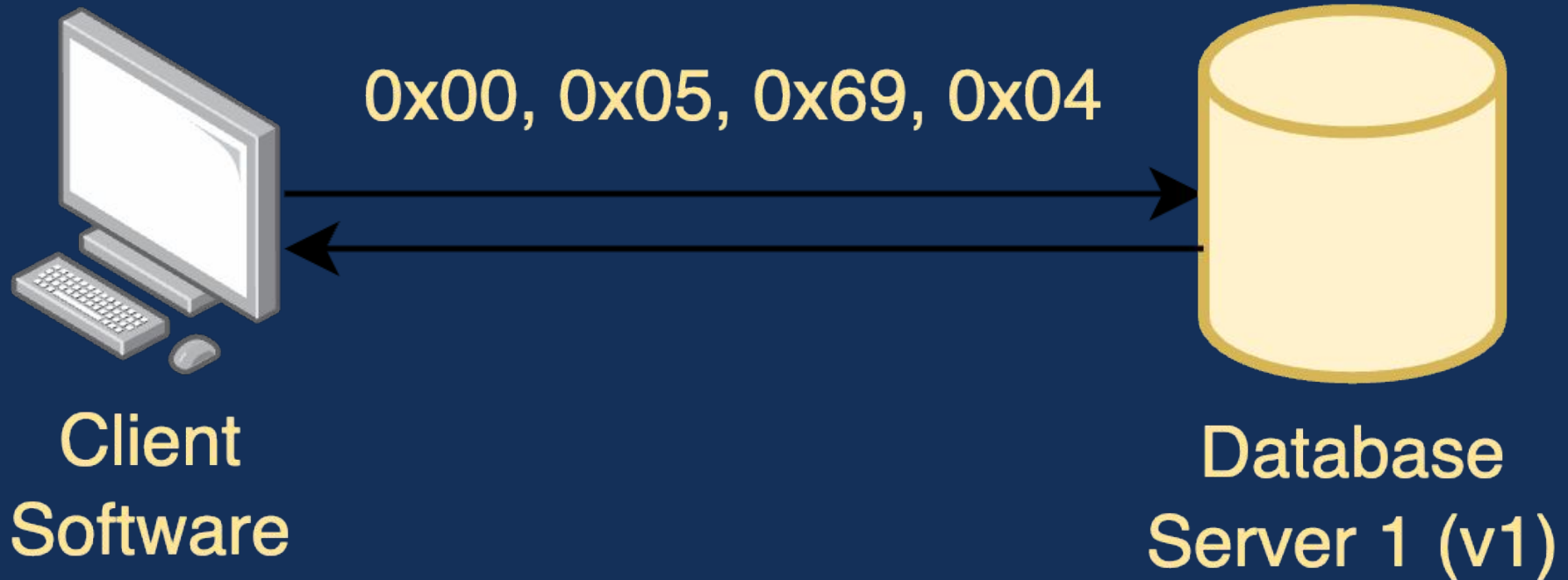


# Available monitoring tools

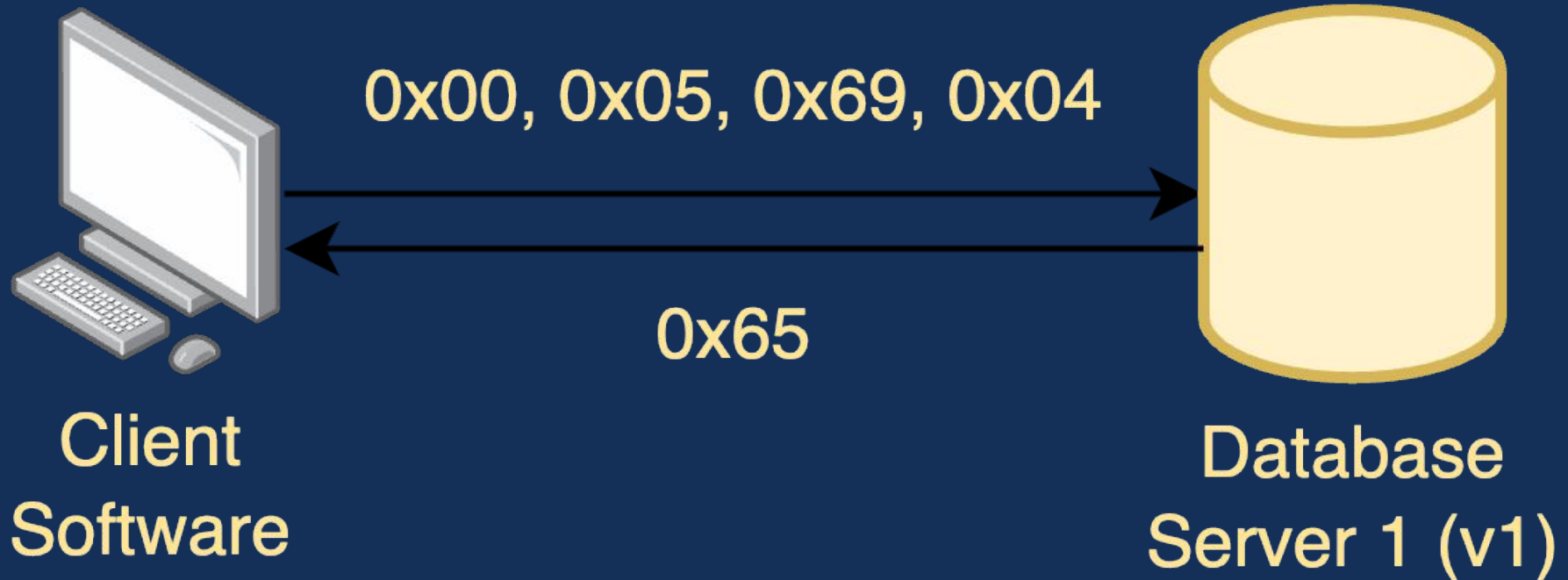




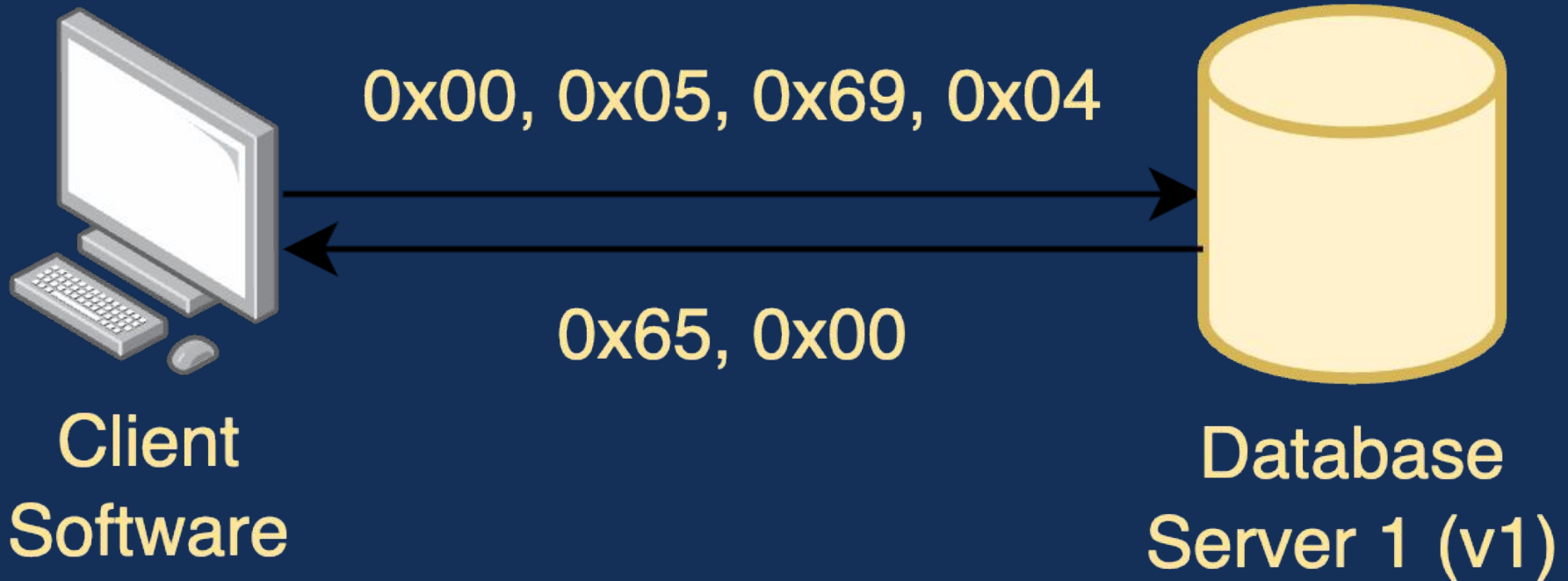
# Available monitoring tools



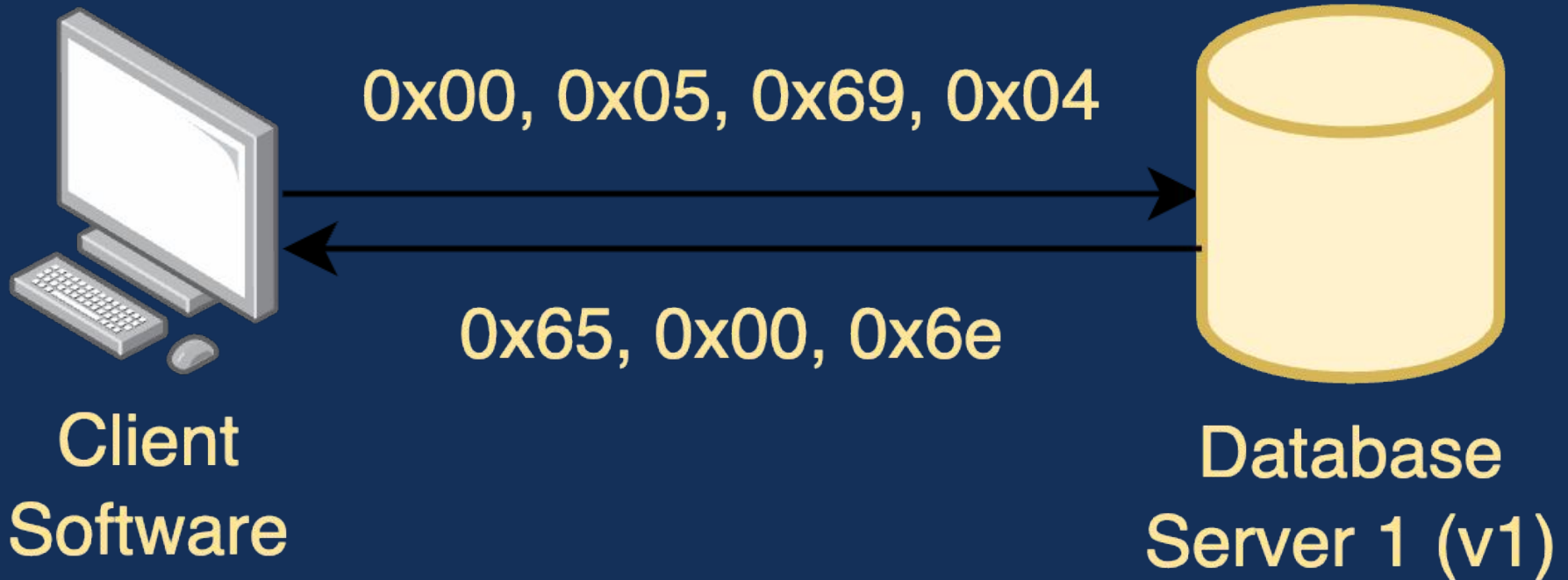
## Available monitoring tools



## Available monitoring tools

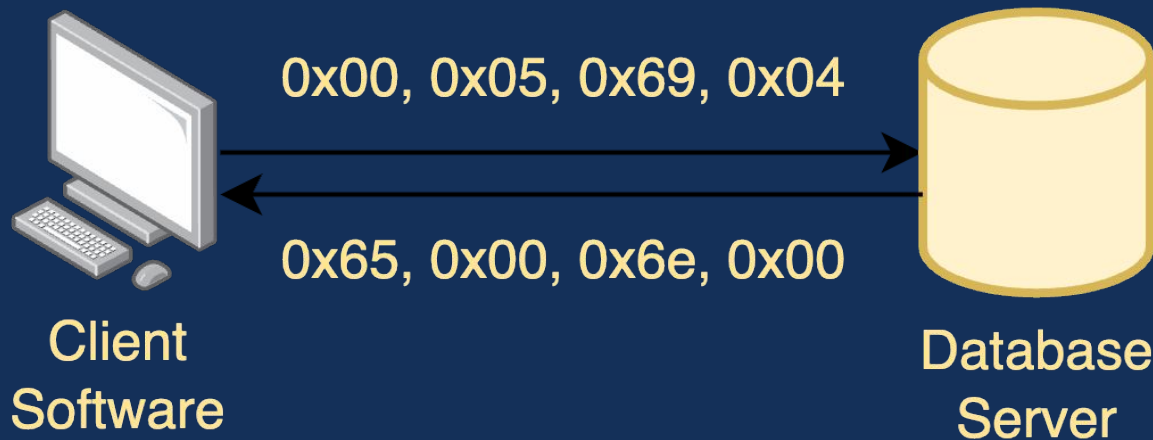


## Available monitoring tools



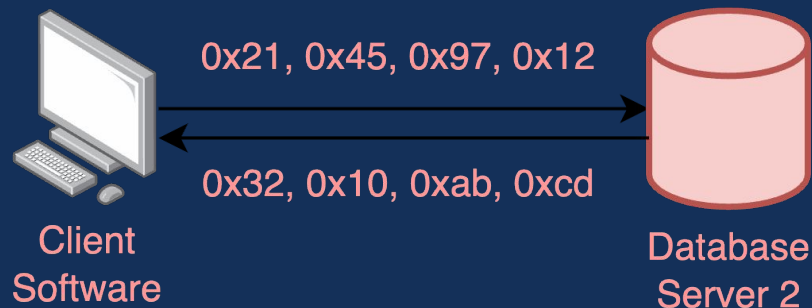
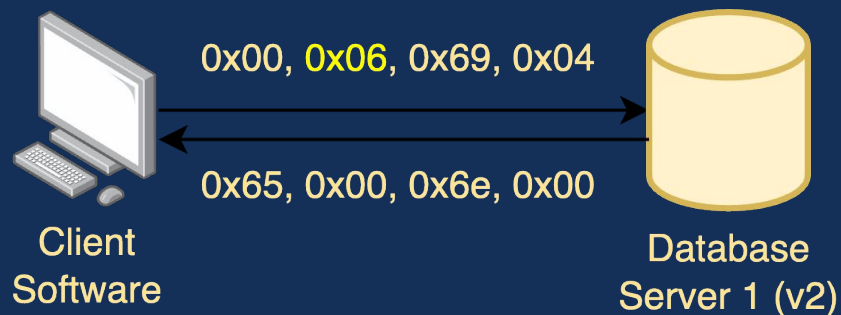
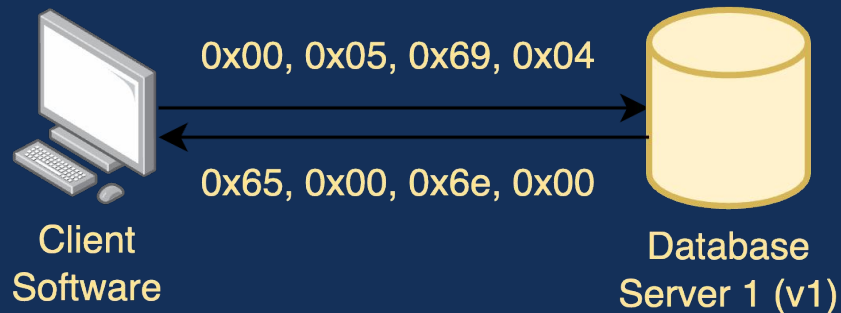
# Hypothesis

**Similar software systems will exhibit comparable communication patterns at application layer, while different systems will show distinct ones.**



# Research scope

**Similar database engines will exhibit comparable communication patterns, while different engines will show distinct patterns.**



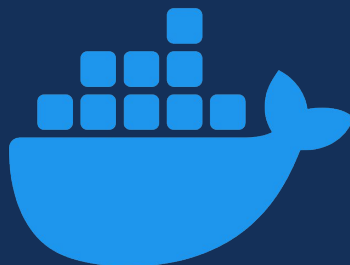
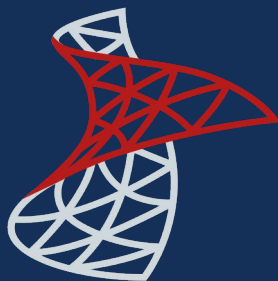
# Analysed software

**3 database types (MySQL, PostgreSQL, SQL Server)**

**20 major versions**

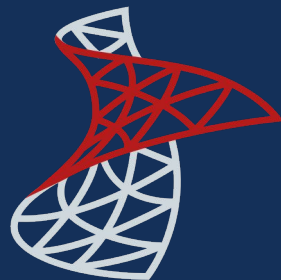
**2 deployments (official Docker images, Google Cloud SQL)**

**31 unique databases configurations**



# Corpus

Several (tens) MB of data of intercepted database traffic, capturing the interaction between **clients** and **database engines**, during regular interactions (simple interactions, sysbench, custom benchmarks).





# Similarity Score Function

**4-grams** : sequential word combinations

**TF-IDF** : word importance score

**cosine distance** : vector angle difference



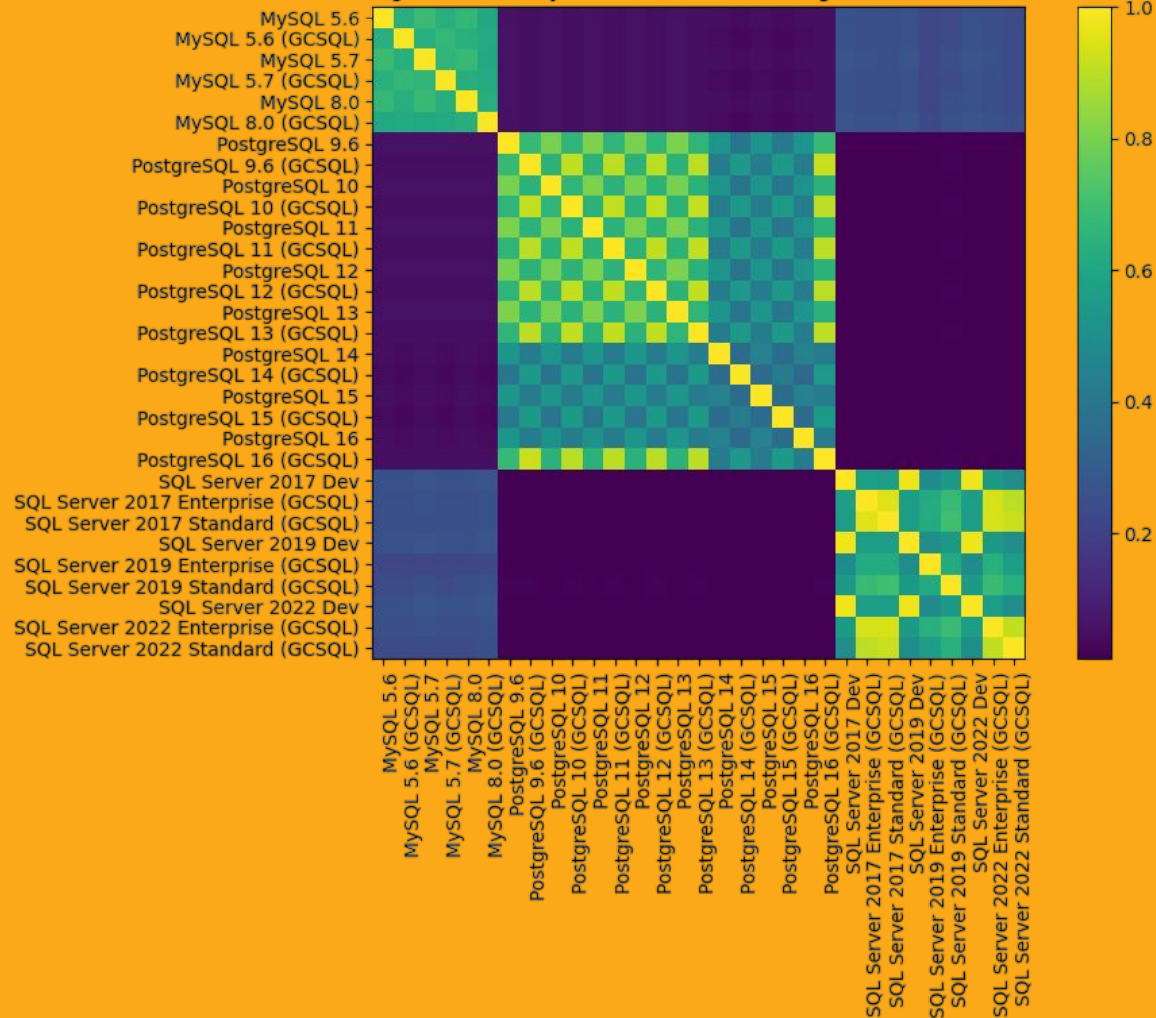
Database Engine	Deployment	MySQL 5.6						MySQL 5.7					
		Docker image			Google Cloud SQL			Docker image			Google Cloud SQL		
		2013/02			2016/07 (Cloud SQL 2nd Gen GA)			2015/10			2016/07 (Cloud SQL 2nd Gen GA)		
		Similarity scores						(2-gram)		3-gram		4-gram	
MySQL 5.6		-			.78	.68	.64	.81	.73	.69	.77	.67	.64
MySQL 5.6 (Google)		.78	.68	.64				.77	.68	.64	.79	.70	.67
MySQL 5.7		.81	.73	.69	.77	.68	.64			-	.78	.69	.66
MySQL 5.7 (Google)		.77	.67	.64	.79	.70	.67	.78	.69	.66			-
MySQL 8.0		.81	.72	.67	.77	.67	.63	.81	.73	.68	.76	.67	.63
MySQL 8.0 (Google)		.75	.64	.61	.76	.66	.62	.75	.65	.61	.76	.65	.62
PostgreSQL 9.6		.47	.25	.06	.45	.23	.06	.47	.25	.06	.44	.23	.06
PostgreSQL 9.6 (Google)		.44	.22	.05	.42	.20	.05	.44	.22	.05	.41	.20	.05
PostgreSQL 10		.47	.24	.06	.45	.23	.06	.47	.24	.06	.44	.23	.06
PostgreSQL 10 (Google)		.44	.22	.05	.41	.20	.05	.43	.21	.05	.41	.20	.05
PostgreSQL 11		.47	.24	.06	.44	.23	.06	.46	.24	.06	.43	.23	.06
PostgreSQL 11 (Google)		.44	.22	.05	.41	.20	.05	.43	.21	.05	.41	.20	.05
PostgreSQL 12		.47	.24	.06	.45	.23	.06	.47	.24	.06	.44	.23	.06
PostgreSQL 12 (Google)		.44	.22	.05	.41	.20	.05	.43	.21	.05	.41	.20	.05
PostgreSQL 13		.47	.25	.06	.45	.23	.06	.47	.24	.06	.44	.23	.06
PostgreSQL 13 (Google)		.44	.22	.05	.41	.20	.05	.43	.22	.05	.41	.20	.05
PostgreSQL 14		.41	.19	.05	.39	.18	.04	.40	.19	.05	.38	.18	.04
PostgreSQL 14 (Google)		.38	.16	.04	.36	.15	.03	.38	.16	.04	.35	.15	.03
PostgreSQL 15		.40	.19	.05	.38	.18	.04	.40	.19	.05	.38	.18	.04
PostgreSQL 15 (Google)		.38	.17	.04	.35	.16	.03	.37	.17	.04	.35	.16	.03
PostgreSQL 16		.43	.19	.05	.40	.18	.04	.42	.20	.05	.40	.18	.04
PostgreSQL 16 (Google)		.44	.22	.05	.41	.20	.05	.43	.21	.05	.41	.20	.05
SQL Server 2017 dev		.41	.30	.26	.36	.29	.26	.40	.31	.27	.36	.28	.26
SQL Server 2017 enterprise (Google)		.42	.31	.25	.37	.29	.25	.41	.31	.26	.37	.29	.25
SQL Server 2017 standard (Google)		.42	.31	.25	.37	.29	.25	.41	.31	.25	.37	.29	.25
SQL Server 2019 dev		.41	.30	.26	.36	.29	.26	.40	.31	.27	.36	.28	.26
SQL Server 2019 enterprise (Google)		.39	.28	.22	.35	.26	.22	.38	.28	.22	.34	.26	.22
SQL Server 2019 standard (Google)		.42	.30	.24	.37	.29	.24	.41	.31	.25	.37	.28	.24
SQL Server 2022 dev		.41	.30	.26	.36	.29	.26	.40	.31	.27	.36	.28	.26
SQL Server 2022 enterprise (Google)		.41	.31	.25	.37	.29	.25	.41	.31	.26	.37	.29	.25
SQL Server 2022 standard (Google)		.39	.29	.24	.35	.27	.24	.38	.29	.24	.35	.27	.24

# MySQL v. other engines

High similarity between different versions of the same database engine.

Low similarity when comparing different database engines.

4-gram Similarity Between Database Engines (full dataset)



**4-gram similarity scores between various database versions**

**High similarity between different versions of the same database engine.**

**Low similarity when comparing different database engines.**

# Conclusions

**It seems possible to distinguish between different relational database engines (e.g. MySQL, PostgreSQL, and SQL Server) and even different versions or deployment environments of the same engine, solely by examining their network traffic at application layer .**

**The research used n-gram analysis (specifically 4-grams ) of network traffic bytes and TF-IDF analysis with cosine similarity to quantify these similarities.**

# Thank you!



**Follow Project Martial development:**

**<https://github.com/raresraf/project-martial>**