

The background of the slide is a vibrant, abstract composition of light trails and digital motifs. On the left, there are warm orange and yellow streaks. On the right, there are cool blue and green streaks. A central green rectangular area serves as a backdrop for the text. Overlaid on these elements are various digital icons, including plus signs, arrows, and geometric shapes, suggesting a high-tech or network environment.

# SYNC in 5G and AI

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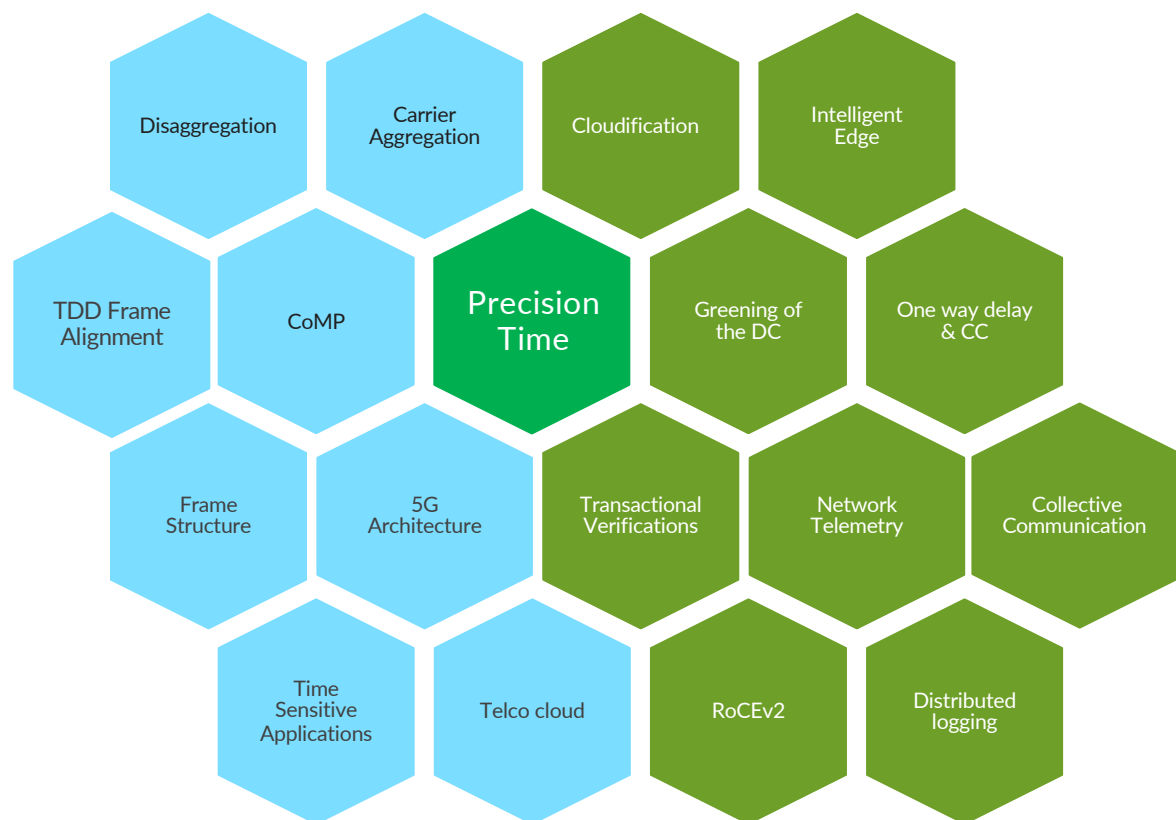
# Agenda

- Sync in 5G and AI
- 5G Timing Requirements
- 5G Timing Overview
- O-RAN Configurations and Limit
- AI for Networking and Networking for AI
- Timing in AI Cluster
- TAP DC Profile
- AI Cluster Timing Requirements
- Predictive Modelling
- Summary

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As resource becomes more efficient to use, the cost of using the resource drops, which can stimulate demand.  
-William Stalling Jevons

# SYNC in 5G and AI

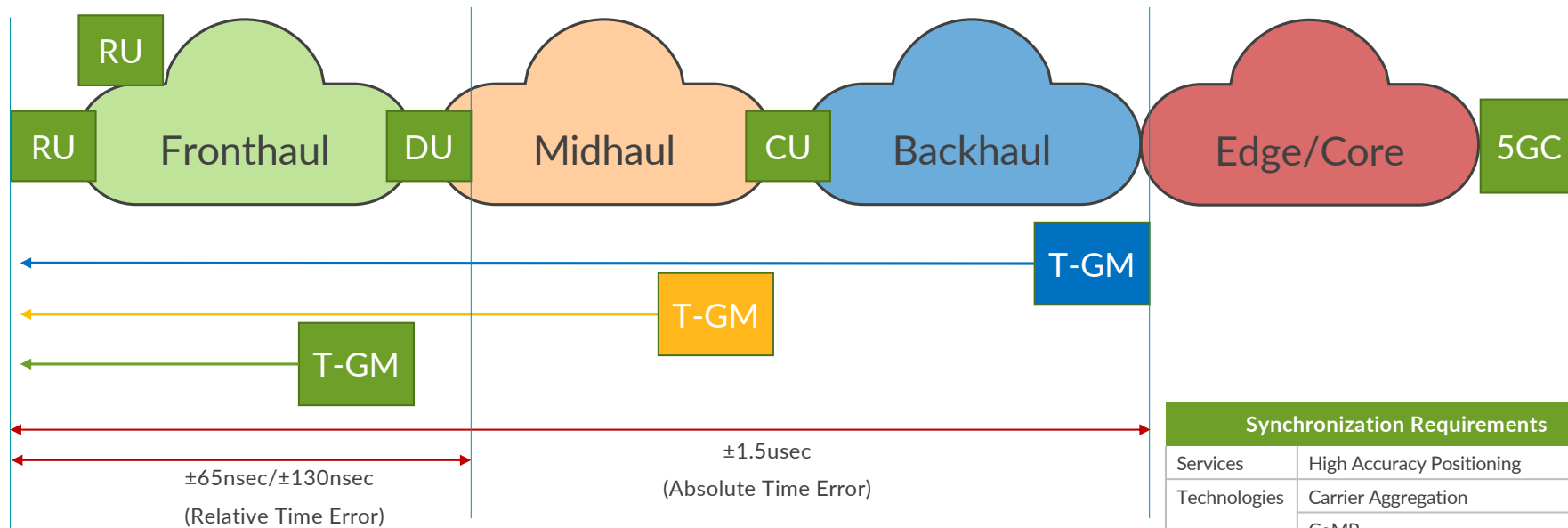


- Data centres transition from private to public cloud infrastructure
- More Intelligent applications moving to network edge. (Ex: Internet of Things, online gaming, 5G connectivity and autonomous driving)
- A single autonomous car produces **4–5 TB/day**.
- Deploy energy-efficient data centres with smaller carbon footprints
  - According to the International Energy Agency, data centres worldwide consumed about [200 terawatt-hours \(TWh\) of electricity in 2020](#), nearly 1% of global electricity demand.
- Training jobs are highly sensitive to latency and packet loss. 10usec delay can cause 20% connectivity issues and training loss.

# 5G Timing Requirements

Technology/Application	Phase accuracy	Relative/Absolute
Intra-band contiguous Carrier Aggregation (CA)	+/- 65 ns	Relative
Intra-band non-contiguous CA	+/- 130 ns	Relative
Inter-band CA	+/- 130 ns	Relative
Coordinated multi-point (CoMP) with Joint Transmission (JT)	+/- 130 ns	Relative
High accuracy positioning service (All RRU/ AAU connected to same DU)	10 ns	Relative
Self-driving/Autonomous cars	< 5 ns	Relative
Transmit diversity Category A+	+/- 32 ns	Relative
MIMO (category A+)	+/- 32 ns	Relative
eCPRI (IEEE 802.1CM)	+/- 130 ns	Relative

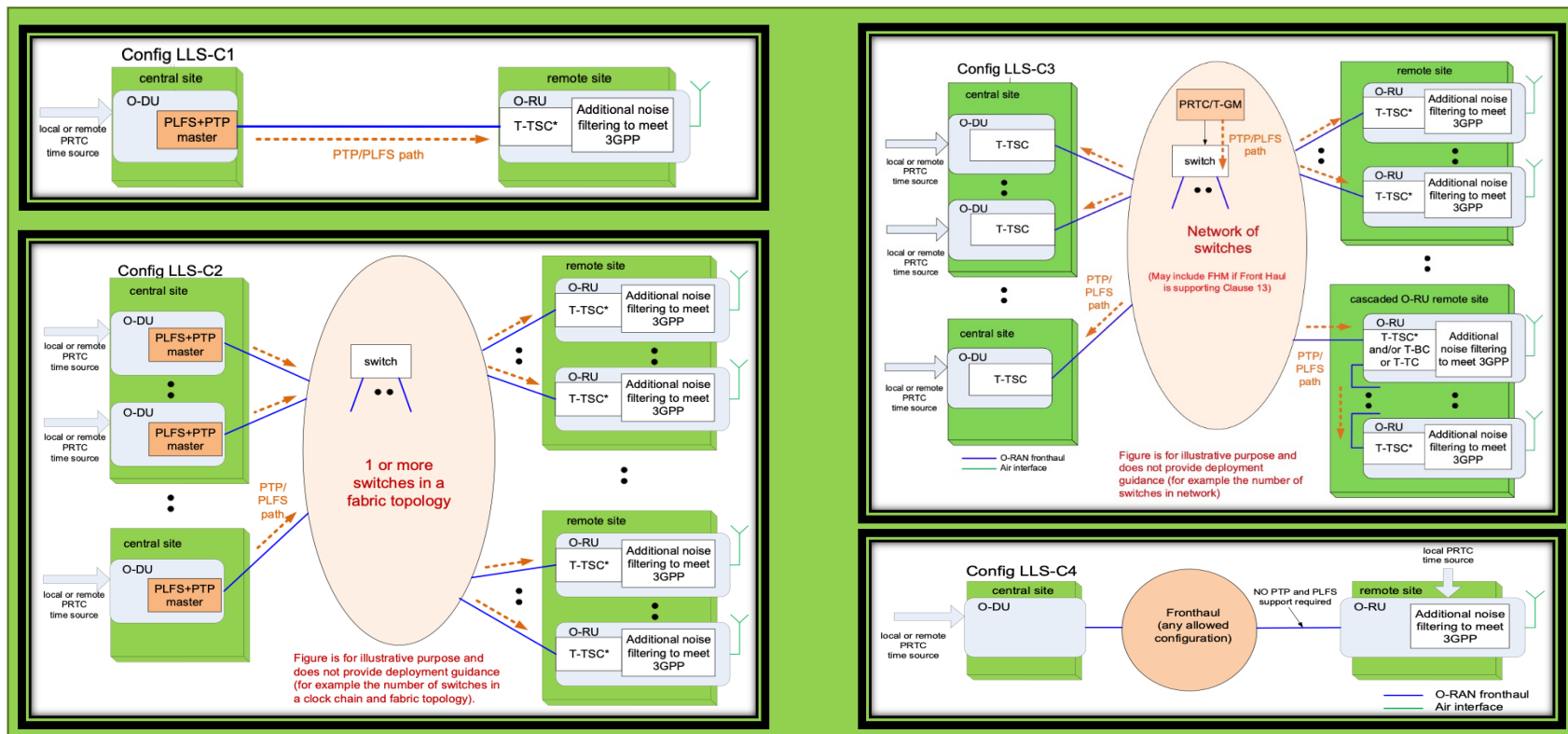
# 5G Overview



Synchronization Requirements	
Services	High Accuracy Positioning
Technologies	Carrier Aggregation
	CoMP
	5G Frame Structure
Architecture	Backhaul, Midhaul and Fronthaul



# O-RAN Deployment Configurations



Source: O-RAN-WG4.CUS.0-v08.00 specification

# O-RAN Timing Network Limit

LLS-C2 Configuration

Network Limit at O-RU UNI	Expected TE Limit	
	Enhanced RU	Regular RU
Packet max TE <sub>L</sub>	≤ 1465ns	≤ 1420ns
Packet max TE <sub>RL</sub>	≤ 60ns (FR2)	≤ 100ns
	≤ 190ns (FR1)	
1PPS max TE <sub>L</sub>	≤ 1465ns	≤ 1420ns
1PPS max TE <sub>RL</sub>	≤ 60ns (FR2)	≤ 100ns
	≤ 190ns (FR1)	
Frequency Limit (For O-DU Class-A)	≤ 36ppb	≤ 36ppb
Frequency Limit (For O-DU Class-B)	≤ 32ppb	≤ 32ppb

LLS-C3 Configuration

Network Limit at O-RU UNI	Expected TE Limit	
	Enhanced RU	Regular RU
Packet max TE <sub>L</sub>	≤ 1100ns	≤ 1100ns
Packet max TE <sub>RL</sub>	≤ 60ns (FR2)	≤ 100ns
	≤ 190ns (FR1)	
1PPS max TE <sub>L</sub>	≤ 1100ns	≤ 1100ns
1PPS max TE <sub>RL</sub>	≤ 60ns (FR2)	≤ 100ns
	≤ 190ns (FR1)	
Frequency Limit (For O-DU Class-A)	≤ 36ppb	≤ 36ppb
Frequency Limit (For O-DU Class-B)	≤ 32ppb	≤ 32ppb



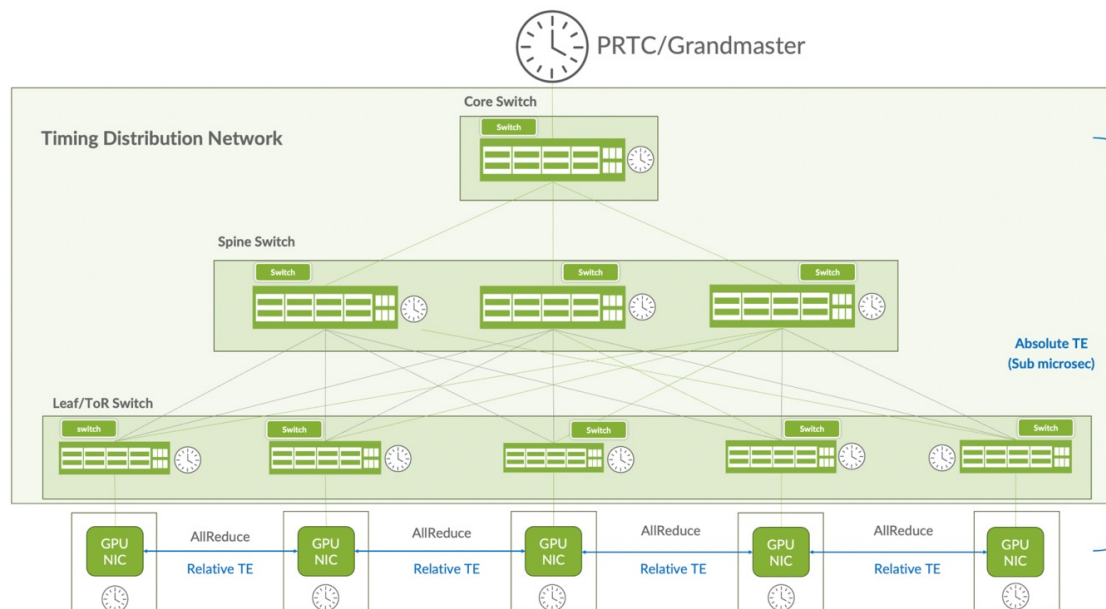
# AI for Networking vs Networking for AI

- ❑ Using AI/ML models to make the network smarter, autonomous, self-optimizing
  - ✓ 5G RAN, Core, Transport, Operators
- ❑ Building network infrastructure that supports AI workloads, AI training clusters, Fabrics.
  - ✓ AI datacenters, MEC AI clusters

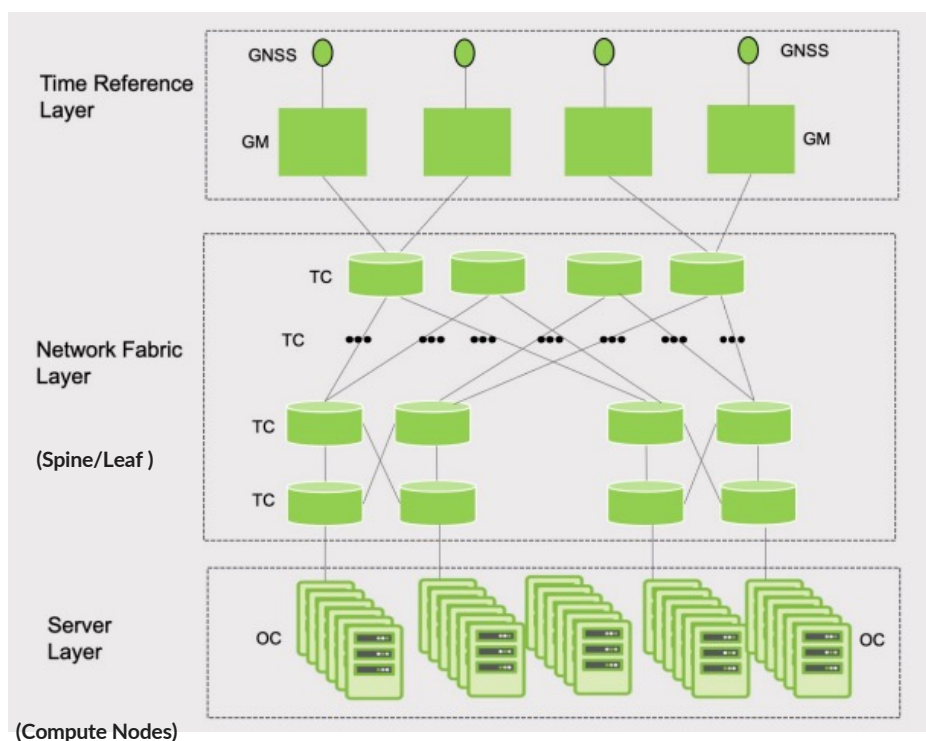
Use cases	
<b>RAN Optimization</b> <ul style="list-style-type: none"><li>✓ Beamforming</li><li>✓ PRB scheduling</li><li>✓ Dynamic TDD</li><li>✓ Mobility/Handover Prediction</li></ul>	<b>AI Training Clusters over 5G</b>
<b>Network Optimization</b> <ul style="list-style-type: none"><li>✓ Congestion prediction</li><li>✓ Path selection</li><li>✓ OWD based congestion control</li><li>✓ Drift prediction</li><li>✓ Time Error Prediction</li></ul>	<b>MEC AI inference</b>

# Timing Distribution in AI Cluster

- ❑ AI clusters interconnect thousands of GPUs and servers- **high-bandwidth, low-latency fabric** built from Ethernet, InfiniBand, or RoCEv2 switches.
- ❑ **Fabric node**, forwards compute data (model parameters, gradients) and **timing information** (PTP, SyncE, or TSN time sync).
- ❑ CLOS or Dragonfly interconnect with spine-leaf layers (**BC or TC**), ensuring deterministic multi-path routing.



# TAP-Model1



Source: OCP Profile for DC (Model1)

Time Error Requirements	Accuracy
The maximum absolute time error between any two OCs	$\leq 5 \mu s$
The maximum absolute time error between a GM and any OCs	$\leq 2.5 \mu s$
The maximum time error between any 2 GMs	$\leq 100 ns$
The maximum time error generated by a TC	$\leq 100 ns$

Source: OCP Profile for DC (Model1)

Typical Applications	Relative Time error requirements between TSCs
Distributed databases, applications profiling	$\leq 5 \mu s$
High-Frequency Telemetry, Multi-node performance analysis tools	$\leq 2.5 \mu s$
Congestion control based on one way delay, Time synchronized collective communication	200 ns

Source: ITU-T G-Sup.DCSync

## ITU-T Profiles consideration and IEEE1588.1

PTPoE link local multicast derived from G.8275.1 profile  
 PTPoIP unicast based on IEEE P1588.1 – **Client Server PTP (CSPTP)** with full timing support from the network utilizing transparent clocks.

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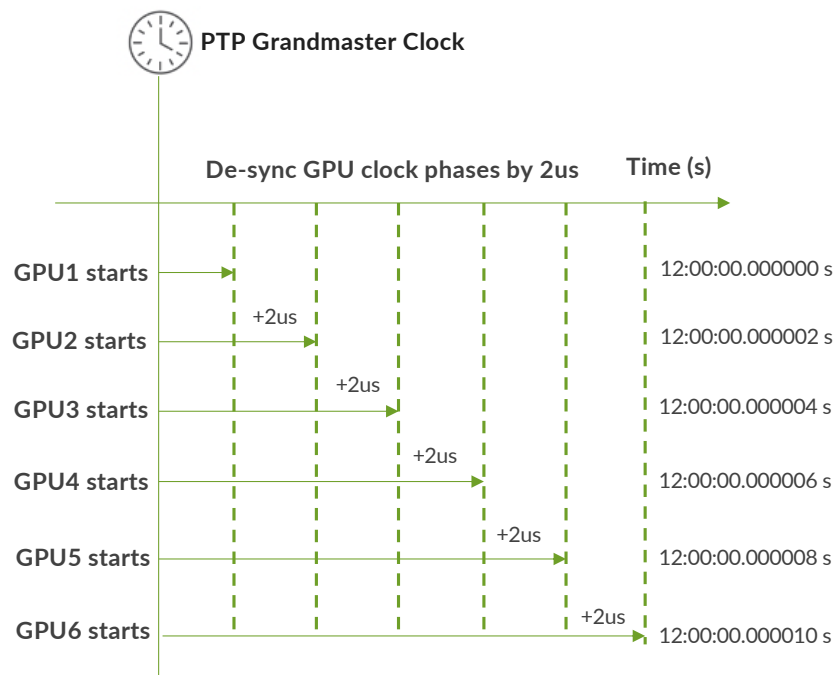
# Time aware collective communication

## Deterministic collective communication

- ❑ AI training clusters involve **tens of thousands of GPUs** exchanging gradients every few microseconds.
- ❑ Each GPU independently computes the loss and local gradients for its data slice.
- ❑ GPUs must synchronize their updates through a collective communication operation called AllReduce.
- ❑ Precision timing can ensure that these exchanges happen with sub-microsecond accuracy.

# Sync-to-desync Power Control

- ❑ Stagger the GPUs start at precise time, avoiding high current burst.
- ❑ Dynamically schedule loads to machine based on power consumption.
- ❑ Dynamically schedule loads to machine based on the computing efficiency.



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# One way Delay

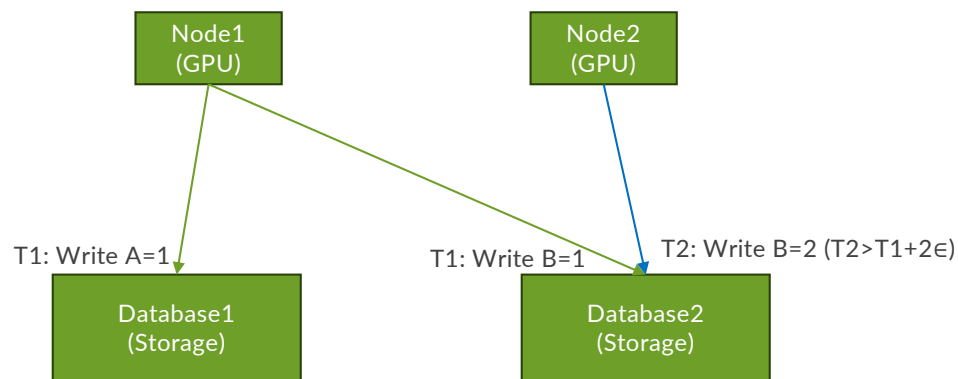
With sub-microsecond-level clock differences across devices, we can measure one-way delay, locate packet losses, and identify per-hop latency bursts.

Enables synchronized network snapshots.

Better congestion signal to delay-based congestion control to differentiate between forward and reverse path congestion

# Distributed Transactional Databases

## Time Uncertainty Bound ( $\epsilon$ )



- T1 (Node 1) is a distributed transaction that writes A=1 on Database 1 and B=1 on Database 2.
- T2 (Node 2) later wants to write B=2 on Database 2.
- ' $\epsilon$ ' is the **max clock error** (time-uncertainty) relative to real time.
- To be **certain** T2 happens after T1 in *real time*
- $T2 > T1 + 2\epsilon$
- Better clock sync  $\rightarrow$  **smaller  $\epsilon$**   $\rightarrow$  **smaller commit-waits** and shorter safe gaps
- Drive  $\epsilon$  from, say, **100 ns** with PTP/DC-grade sync, that  **$2\epsilon$  safety margin shrinks to 200 ns**, materially cutting write latency.

FaRMv2, an RDMA-based transactional system, observes the median transaction delay can drop by 25% if we improve  $\epsilon$  from  $\sim 20\mu\text{s}$  to 100ns.

CockroachDB can significantly reduce the retry rate when  $\epsilon$  drops from 1ms to 100ns.

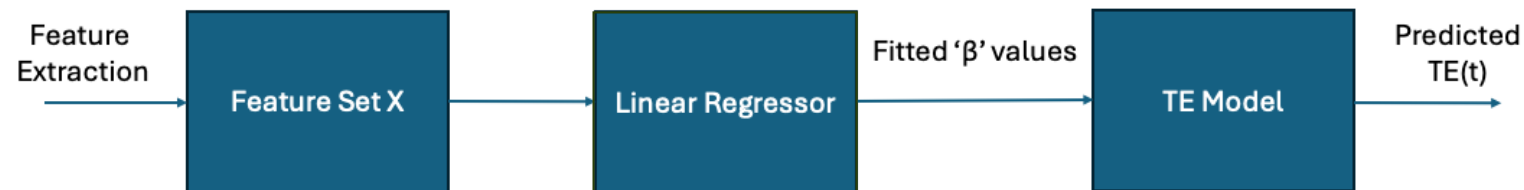


# Predictive Modelling in Sync using ML and DL

**Predictive Modelling:** Learn from the data sets and predicting.

**Adaptive Systems:** Compensate for errors due to dynamic changes.

**Optimized Control:** Optimized servo loops to minimize clock drift and deviation.



# Model Performance

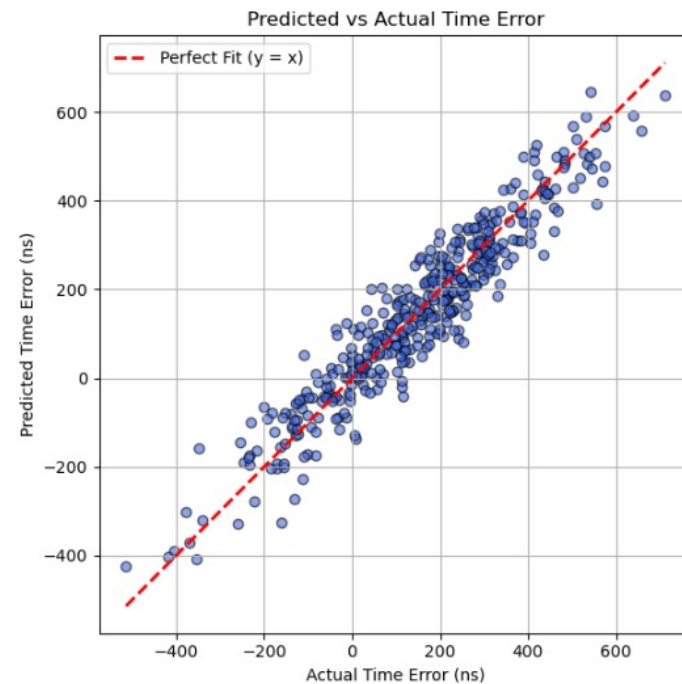
R-squared	Adj R-squared
0.899598	0.896751

## R-squared/Adjusted R-squared

- The model explains **~90 % of total variance** in time-error measurements.
- Only ~10 % remains unmodeled (random PDV, jitter spikes, or unmeasured features).
- Adjusted  $R^2 \approx 0.897$ . This indicates high explanatory power even after accounting for feature count.

## Model Quality

Very good, captures most deterministic contributors to TE.



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# Questions

The background is a vibrant, abstract composition of light trails in shades of orange, yellow, green, and blue, creating a sense of rapid motion. Overlaid on this are various network-related icons, including nodes, lines, and data flow patterns, rendered in a semi-transparent, digital style.

# THANK YOU

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