Genetic Algorithm For LoRa Transmission Parameter Selection

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December 12, 2019



- 1. Introduction
- 2. Survey A
- Genetic Algorithm For LoRa
- 4. Emergency Evacuation stem
- 5. Q-Learning
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- 9. Conclusion
- 10. Survey plafforms

- 1. IoT Devices
- 2. IoT Applications
- 3. IoT Wireless Communications

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1. IoT Devices

- IoT Applications
- IoT Wireless Communications

Massive IoT devices

Higher Categories

IoT devices are useless without a good communication capability

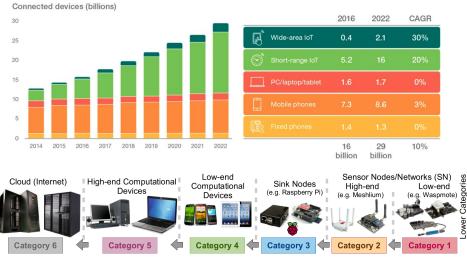


Figure 1. IoT devices [1].

1. Introduction | 1. IoT Devices 1/80

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- I. IoT Devices
- 2. IoT Applications
- IoT Wireless Communications

Applications diversification

Each application has its own communication requirements

Challenges/Applications	Grids	EHealth	Transport	Cities	Building
Resources constraints	X	/	Х	-	X
Mobility	X	-	/	✓	Х
Heterogeneity	-	-	-	✓	Х
Scalability	√	-	/	✓	-
QoS constraints	-	-	/	✓	/
Data management	-	Х	/	✓	-
Lack of Standardization	-	-	-	-	/
Amount of attacks	Х	Х	/	✓	✓
Safety	-	/	/	-	/

Table 1. Main IoT challenges [2] [3]



Figure 2. IoT Applications.

1. Introduction | 2. IoT Applications 2/80

IoT platforms

IoT platforms is a chain of communication process

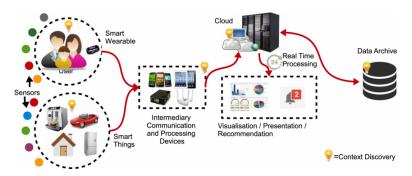


Figure 3. IoT platform.



Figure 4. IoT challenges.

Applications diversification

Requirements

Use Case	Packet rate [pkt/day]	Min success rate [Ps,min]	Payload Size [Byte]
Wearables	10	90	
Smoke Detectors	2	90	
Smart Grid	10	90	10-20
White Goods	3	90	
Waste Management	24	90	
VIP/Pet Tracking	48	90	
Smart Bicycle	192	90	
Animal Tracking	100	90	1
Environmental Monitoring	5	90	1
Asset Tracking	100	90	50
Smart Parking	60	90	
Alarms/Actuators	5	90	
Home Automation	5	90	
Machinery Control	100	90	
Water/Gas Metering	8	90	
Environmental Data Collection	24	90	
Medical Assisted Living	8	90	
Micro-generation	2	90	
Safety Monitoring	2	90	100-200
Propane Tank Monitoring	2	90	1
Stationary Monitoring	4	90	1
Urban Lighting	5	90	
Vending Machines Payment	100	90	
Vending Machines General	1	90	1K

Table 2. Application requirements for the use cases of interest [4] [3] [5]

- 1. Introduction

- 3. IoT Wireless Communications
- 1. LoRa

IoT wireless communication

Wireless communication performance need to be evaluated to match applications requirements

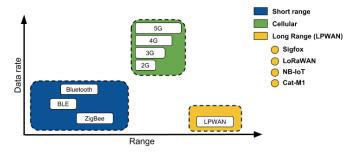


Figure 5. Short range, Cellular and Long range networks.

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Wireless communication

Exp: LPWAN in a new technology that satisfy IoT applications requirements

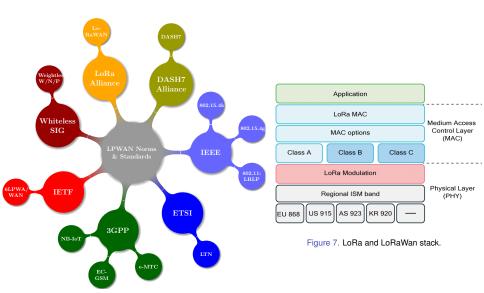


Figure 6. Wireless communication diversity.

LoRa modulation

Physical layer [6]

$$\mathbf{LoRa} = \frac{2^{SF}}{BW} \left((NP + 4.25) + \left(SW + \max \left(\left\lceil \frac{8PS - 4SF + 28 + 16CRC - 20IH}{4(SF - 2DE)} \right\rceil (CR + 4), 0 \right) \right) \right)$$
(1)

$$\mathbf{GFSK} = \frac{8}{DR}(NP + SW + PL + 2CRC) \tag{2}$$

Where:

- NP = 8, if LoRa . 5, if GFSK
- SW = 8, if LoRa . 3, if GFSK
- CRC = 0 if downlink packet. 1 if uplink packet
- IH = 0 if header. 1 if no header present
- DE = 1 if data rate optimization. 0 if not
- Pavload size (PS) = PHY Pavload bytes
- \Rightarrow Spreading Factor (*SF*) = 7. 8. 9. 10. 11. 12
- Bandwidth (BW) = 125kHz, 250kHz, where BW is the bandwidth
- Coding Rate (CR) = Indicates the Coding Rate

LoRa parameters selection

How to select the optimal configuration

- Parameters
 - → BW
 - → SF
 - → CR
 - → Transmission Power (P^{tx})

- Metrics
 - → Signal Noise Rate (SNR)
 - → Data Rate (DR)→ Air Time (AT)
 - → Payload length (*PktL*)
 - → Receiver Sensitivity (S_{rx})

Setting	Values	Rewards	Costs
BW	7.8 ⇒ 500 <i>kHz</i>	DR	S_{rx} , Range
SF	2 ⁶ → 2 ¹²	S_{rx} , Range	DR, SNR, PktL, P ^{tx}
CR	4/5 → 4/8	Resilience	PktL, P ^{tx} , AT
P ^{tx}	-4 ⇒ 20 <i>dBm</i>	SNR	P ^{tx}

Table 3. LoRa parameters selection [7]

LoRa Frame

Prea	mble	Sync msg	PHY Header	PHDR-CRC						
Modulation	length	Sync msg	PHY Header	PHDR-CRC		MAC Header				
Modulation	length	Sync msg	PHY Header	PHDR-CRC	МТуре	RFU	Major			
Modulation	length	Sync msg	PHY Header	PHDR-CRC	МТуре	RFU	Major	Dev A	Address	
Modulation	length	Sync msg	PHY Header	PHDR-CRC	МТуре	RFU	Major	NwkID	NwkAddr	ADR
0	1	2	3	4	5	6	7	8	9	10
PHY Payload							CRC			
MAC Payload					MIC	CRC Type	Polynomial			
	Frame Header FPort Frame Payload					MIC	CRC Type	Polynomial		
	FCtrl			FCnt	FOpts	FPort	Frame Payload	MIC	CRC Type	Polynomial
ADRACKReq	ACK	FPending /RFU	FOptsLen	FCnt	FOpts	FPort	Frame Payload	MIC	CRC Type	Polynomial
11	12	13	14	15	16	17	18	19	20	21

LoRa Frame

- Modulation :
 - → Lora: 8 Symbols, 0x34 (Sync Word)
 - → FSK: 5 Bytes, 0xC194C1 (Sync Word)
- → Length:
- Sync msq :
- PHY Header : It contains:
 - → The Payload length (Bytes)
 - → The Code rate
 - → Optional 16bit CRC for payload
- → Phy Header : CRC It contains CRC of Physical Layer Header
- MType: is the message type (uplink or a downlink)
 - whether or not it is a confirmed message (reqst ack)
 - → 000 Join Request
 - → 001 Join Accept
 - → 010 Unconfirmed Data Up
 - → 011 Unconfirmed Data Down
 - → 100 Confirmed Data Up
 - → 101 Confirmed Data Down
 - → 110 BFU
 - → 111 Proprietary
- RFU : Reserved for Future Use
- Major: is the LoRaWAN version; currently, only a value of zero is valid.
 - → 00 LoBaWAN R1
 - → 00 LORAWAN R
 - → 01-11 RFU
- NwkID : the short address of the device (Network ID): 31th to 25th
- NwkAddr: the short address of the device (Network Address): 24th to 0th
- ADR: Network server will change the data rate through appropriate MAC commands
 - → 1 To change the data rate
 - → 0 No change

- ADRACKReq: (Adaptive Data Rate ACK Request): if network doesn't respont in 'ADR-ACK-DELAY' time, end-device switch to next lower data rate.
 - → 1 if (ADR-ACK-CNT) >= (ADR-ACK-Limit)
 - → 0 otherwise
- ACK: (Message Acknowledgement): If end-device is the sender then gateway will send the ACK in next receive window else if gateway is the sender then end-device will send the ACK in next transmission.
 - → 1 if confirmed data message
 - → 0 otherwise
- FPending! /RFU 1: (Only in downlink), if gateway has more data pending to be send then it asks end-device to open another receive window ASAP
 - → 1 to ask for more receive windows
 - → 0 otherwise
- ▼ FOptsLen: is the length of the FOpts field in bytes ă 0000 to 1111
- FCnt: 2 type of frame counters
 - → FCntUp: counter for uplink data frame, MAX-FCNT-GAP
 - → FCntDown: counter for downlink data frame, MAX-FCNY-GAP
- FOpts: is used to piggyback MAC commands on a data message
- → FPort : a multiplexing port field
 → 0 the payload contains only MAC commands
 - the payload contains only MAC commands
 - → 1 to 223 Application Specific
 - → 224 & 225 RFU
- → FRMPayload : (Frame Payload) Encrypted (AES, 128 key length) Data
- → MIC: is a cryptographic message integrity code
 - computed over the fields MHDR, FHDR, FPort and the encrypted FRMPayload.
- CRC: (only in uplink),
 - → CCITT $x^{16} + x^{12} + x^5 + 1$
 - → IBM $x^{16} + x^{15} + x^5 + 1$

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1. A Relay and Mobility Scheme

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1. A Relay and Mobility Scheme

A Relay and Mobility Scheme for QoS Improvement in IoT

Related work¹

- Only application requirements.
 - → Environment conditions, operator rules, User preferences.
- Only one (simple) normalization function for all parameters.
 - → Use Fuzzy logic with different rules for normalization.
- Only one objective function to fits all requirements.
 - → Use Genetic algorithms with 3 objective functions.
- Only one application.
 - Use 3 applications with different requirements

¹A. A. Simiscuka and G. Muntean, A Relay and Mobility Scheme for QoS Improvement in IoT Communications, in 2018 IEEE International Conference on Communications Workshops (ICC Workshops), 00002 May 2018, pp. 1–6.

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- 1. MCDM
- 2. Genetic Algorithm
- 3. Fuzzy Logic

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- Fuzzy Logic

Multi criteria decision making

Background

- Configuration parameters:
 - → SF
 - → CR
 - $\rightarrow P^{tx}$
 - → BW
 - → PS
- Configuration metrics:
- → DR
 - → Packet delivery ratio (PDR)
 - → Round-Trip Delay (RTD)
 - → Time on Air (ToA)

		Metric 1	Metric 2		Metric M
_	Configuration 1 Configuration 2	$\begin{pmatrix} q_{11} \\ q_{21} \end{pmatrix}$	9 ₁₂ 9 ₂₂		9 _{1M})
Qn, m =	: Configuration N	; q _{N1}	: 9 _{N2}	··.	: q _{NM}

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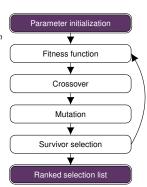
- 1 MCDM
- 2. Genetic Algorithm
- 3. Fuzzy Logic

Genetic Algorithm

Background [9]

Definition: stopping criteria, population size P, and mutation probability p_m Generate randomly the initial configurations repeat:

- ... for each configuration do
- Train a model & compute configuration's fitness
- . . end . . . for each reproduction 1 ... P/2 do
- Select: 2 configurations based on fitness Crossover: Produce 2 child configurations
- Mutate: child configurations with pm
- . . . end
- until stopping criterion is met



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Fuzzy Logic

Assign a degree of membership between 0 and 1

- We have a temperature value (16°), and we want to represent this value with 3 weighted vales.
 - → 0% hot
 - → 0.4% warm
 - → 0.6 % cold

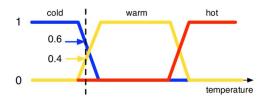


Figure 8. Temperature example.

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Selection framework

Methods

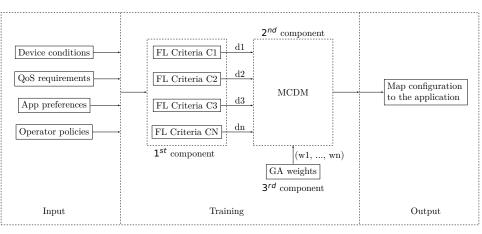


Figure 9. The proposed scheme for LoRa transmission parameters selection based on GA, FL and MCDM..

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Experimentation

Experimentation

Inputs:

- Data structure
 - * Voice, Images and Text transmission.
- → Environment conditions
 - * Rural/Urban
 - * Static/Mobile
 - * Temperature
 - * Interference/Noise
- → QoS metrics:
 - * User layer: Cost
 - * Network metrics: DR, Payload length.
 - * Radio metrics: Receiver sensitivity, SNR, DR, Air time,
- → MAC configuration (SF, CR, BW, Tx)
- Outputs:
 - \rightarrow (SF_i, CR_j, BW_k, Tx_l) optimal

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Results

Comparison

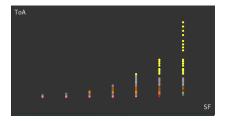


Figure 10. Impact of SF on ToA.

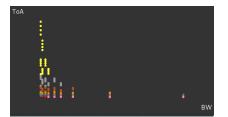


Figure 11. Impact of BW on ToA.

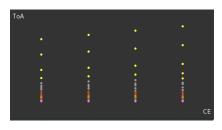


Figure 12. Impact of CR on ToA.

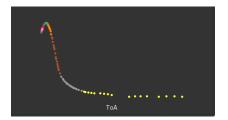


Figure 13. ToA distribution.

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Multi criteria decision making

Layer	Maximize (Reward)	Minimize (Cost)
Application	Sec security	Service Cost (SC)
Network	Range	Jitter (<i>Jit</i>)
	PDR	Traffic congestion (TC)
	PS	RTD
	DR	Packet Error Rate (PER)
		Time Complexity (Otime)
		Space Complexity (O _{space})
Radio	Mobility (<i>Mob</i>)	Bit Error Rate (BER)
	Symbol Rate (SR)	P ^{tx}
	Bit Rate (BR)	Co-channel Interference (CCI)
	Sensitivity (Sen)	Duty cycle (DC)
	Received Signal Strength Indication (RSSI)	ToA
	Signal-to-interference & noise ratio (SINR) SNR	Path loss (PL)
	Signal-to-Interference Ratio (SIR)	

Table 4. Network selection inputs and classification of parameters [10] [11]

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Problem statement

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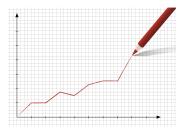


Figure 14. .

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... (step 1)
Methods

... (step 2)
Methods

... (step 3)
Methods

... (step 4)
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Results

Comparison



Table 5

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Experimentation

Experimentation

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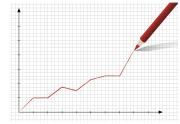


Figure 16. .

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1. Problem statement

Problem statement

Introduction



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Figure 18. .

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Game theory

Related work

```
Players: K = \{1, ..., K\}
```

- Strategies: $S = S_1 \times ... \times S_K$
 - \rightarrow S_k is the strategy set of the k^{th} player.
- Rewards: $u_k: S \longrightarrow R_+$ and is denoted by $r_k(s_k, s_{-k})$
 - $\Rightarrow s_{-k} = (s_1, \dots, s_{k-1}, s_{k+1}, \dots, s_K) \in S_1 \times \dots \times S_{k-1} \times S_{k+1} \times \dots \times S_K$

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Game theory

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Multi-Armed-Bandit Algorithm

Related work

```
⇒ Arms: K = 1, ..., K

⇒ Decision: T = 1, ..., T

⇒ Reward: X_t^k with \mu_t^k = \mathbb{E}[X_t^k]

⇒ Best reward: X_t^* with \mu_t^* = \max \mu_t^k, k∈K
```

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Q Learning Related work

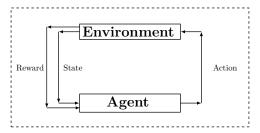


Figure 19. qlearning.

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Marcov chain

Related work

$$V(s,\pi) = \mathbb{E}_{s}^{\pi} \left(\sum_{k=0}^{\inf} \gamma^{k} \cdot r(s_{k}, a_{k}) \right), s \in \mathbb{S}$$
(3)

$$r(s_k, a_k) = G_k \cdot PRR(a_k) \tag{4}$$

$$\pi^* = \arg\max_{\pi} V(s,\pi)$$

$$PRR = (1 - BER)^{L} \tag{6}$$

$$BER = 10^{\alpha e^{\beta SNR}} \tag{7}$$

(5)

Marcov chain

Related work

Learning iterative steps:

- **Choose** action $a_k(t) \sim \pi_k(t)$
- Observe game outcome
 - $\rightarrow a_k(t)$
 - $\rightarrow u_k(a_k(t), a_k(t))$
- **Improve** $\pi_k(t+1)$

Thus, we can expect that $\forall k \in K$

$$\pi_{k(t)} \xrightarrow{t \longrightarrow \infty} \pi_k^*$$

$$u_k(\pi_k(t),\pi_{_k}(t)) \xrightarrow{t \longrightarrow \infty} u_k(\pi_k^*,\pi_{_k}^*)$$

Where:

 $\pi^* = (\pi_1^*, ..., \pi_k^*)$ is the NE strategy profile

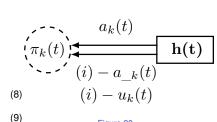


Figure 20. .

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- io. nesults exploitation
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... (step 2)
Methods

... (step 3)
Methods

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Methods

Results

Comparison



Table 6

- 5. Q-Learning

- 4. Experimentation

Experimentation

Experimentation



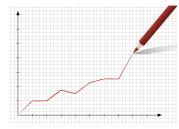


Figure 21. .

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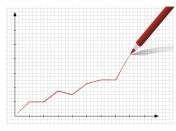


Figure 22. .

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Figure 23. .

5. Q-Learning | 6. Discussion 42/80

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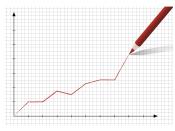


Figure 24. .

- - 5. Q-Learning

- 8. Contagion process

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Methods

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Figure 25. .

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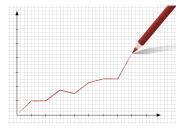


Figure 26. .

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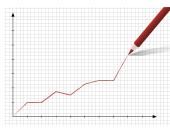


Figure 27. .

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1. Problem statement

Problem statement

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Figure 28. .

- 7. ...

- 2. Contagion process

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Methods

... (step 3)
Methods

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Table 8

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- 3. Experimentation

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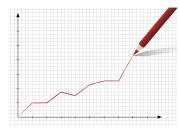


Figure 29. .

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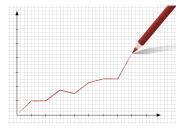


Figure 30. .

- - 7. ...

- 5. Discussion

Discussion



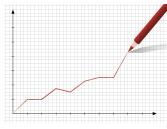


Figure 31. .

7. ... | 5. Discussion 60/80

- 7. ...

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Problem statement

Introduction





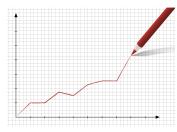


Figure 32. .

- 7. ...

- 7. Contagion process

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Methods

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Methods

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Table 9

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Figure 33. .

- 7. ...

- 9. Results exploitation

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Figure 34. .

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Figure 35. .

7. ... | 10. Discussion 69/80

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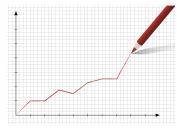


Figure 36. .

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Figure 37. .

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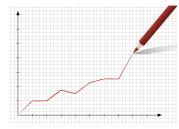


Figure 38. .

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Figure 39. .

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Conclusion

Our main goal was

- .
- .

Our main contribution was

- -
-

Our main results was

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- .

9. Conclusion 79 / 80

Future Challenges

Conclusion

Our future goal was



.

9. Conclusion 80 / 80

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