

Bio-inspired Neuro-Fuzzy Based Dynamic Route Selection to Avoid Traffic Congestion

Sagheer Abbas, M. Saleem Khan, Khalil Ahmed, M. Abdullah and Umer Farooq

Abstract— this paper presents the bio-inspired neuro-fuzzy based route selection system to avoid traffic congestion. The proposed neuro-fuzzy system selects the best multi-parameters direction between two desired nodes: source and the endpoint. This research practices a mixture of neuro-fuzzy logic and ant colony system (ACS) algorithm for the principal routing to fulfill all the preferred requirements of the user using operational traffic data directly carried by the traffic control room and predicted by artificial neural networks for forthcoming minutes of the traffic.

Index Terms— Bio-inspired, Traffic congestion, Dynamic route selection, artificial neural networks, Bio-inspired Fuzzy based traffic system.

1 INTRODUCTION

Traffic congestion is becoming more alarming with every day as the number of vehicles on the roads is increasing rapidly. Route selection grounded on specific desires is a foremost problem for city travelers. In route selection systems, the source and the target are given. The objective of our system is to find out a route with the most favorable conditions satisfying all the needs and desires of a traveller. During the last decades, many researchers tried different approaches to find an optimized route, considering the important parameters that were the most advantageous for the city travellers. In these systems, the distance and travel time were tried to minimize. Now-a-days many drivers are becoming concerned with not only these desires but also with rising fuel costs, to avoid traffic jamming and to adopt the route that satisfies their all desires like if users are facing road-blocking on the way then it should be possible to adopt the next favorable route. It is a non-polynomial NP hard problem to find out the route that fulfills all prerequisites that force to enumerate all the possible routes. Previous researchers have not spoken a fast, self-motivated and a applied except Salehinejad et al who tried a fuzzy based dynamic route selection approach [1].

In 2007, Barth et al have mechanized some pleasant navigation based environmental systems. It consists of the fuzzy logic for pervasive and authoritative device for the

ideal routing [1].

In 2011, Raghavendra et al used a new mechanism for pheromones disappearance through a function based on the convergence of the time. It shows a better performance as compared to all others traditional methodologies [4].

In this paper, Diogo et al attempted ADR algorithm for the routing owing to enhance the efficiency of the system. A new mechanism was used for the cost calculation based on the incentives [6].

In this paper, Saliba et al proposed the quality of service for internet traffic routing based on ants' colony optimization algorithm. The system's performance was compared with the other traditional approaches and it was observed that the presentation of this model was much better as compared to all other methods [7]. In it three parameters were used for the optimizations that are time, distance and the no of vans. This work has much importance regarding time-based planning. Regarding digital image and tracks detection was proposed on ants based algorithms. [8]. Salehinejad et al. have proposed much for the route selection based on the ACS [9].

In this paper, Attia et al suggested two algorithms for routing the network traffic. First got the inspiration by the multi-ant parameters, based on the reactive and proactive approaches to adopt the path and second was the mix-up with the QoS constraints. Its performance was excellent for the real time scenario [12].

In this paper, Xiaomeng et al attempted a quantum mechanism for the travelling salesman problem, initializing the ants with the superposition of qubit. This system shows the performance much better as compared to all other classical techniques [13]. Salehinejad et al introduced a fuzzy based route selection and ant colony system (ACS) based on "Distance", "Incident Risk" and "Traffic" variables [1].

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This research work proposes a bio-inspired neuro-fuzzy based route selection system with six adjustable constraints. This system uses a combination of ANNs, neuro-fuzzy logic and ant colony optimization algorithm to find out an optimum path based on the costs calculated with time, aimed at all possible directions for the user between two nodes origin and the destination, valuing the user's point of interests. An optimum route refers to a route that satisfies all the preferred parameters. In this paper, these parameters are "Distance", "Traffic-flow", "Environment monitoring", "Width", "Road condition", "Traffic lights". Due to the computational complexity six important parameters are taken else some more variables like "Incident risk" "Road condition" and the "View" (Home, building, park, river or any other scenic view the user wants to pass through). In this proposed system the present data is provided by the traffic control center and the coming traffic proceedings data is anticipated by artificial neural networks.

The arrangement of this research paper follows as: section II describes the overview of our suggested system, Artificial Neural Network (ANN) and Ant colony optimization (ACO) algorithm. System and Parameters of the proposed system are discussed in section III, while in section IV the simulation results are discussed. Section V provides the conclusion and future work. Acknowledgement of this work is declared in Section VI.

2 OVERVIEW OF PROPOSED SYSTEM

The proposed system is a combination of ANNs and neuro-fuzzy inference engine. All six inputs to the proposed system are provided from the traffic control room TCR and the average speed of the vehicle is given from the user. The system comprises of the previous data, TCR indications that are being received continuously, the present time and the typical speed of the vehicle. Artificial neural networks are casted for the forseen minutes of the traffic. Time delay estimation is used because time is required to move from base to destination. This estimation is done due to different loads of traffic among varied hours of day and night. This estimation is done as the system should be able to estimate the arrival time of the vehicle to the destination. This estimation is done on the basis of the average speed of the vehicle.

The system has the capability to avoid up coming traffic congestion as it is attentive to the up-to-date vehicle position system using GPS. Therefore, if congestion ensues the system recommends the user next direction according to the customized parametres. It is done by keep notifying the quantity of pheromones on the paths using ACS. The system has the flexibility for the customization of the parametres that are being used in this system.

In the Fig. 1 main block diagram of our proposed scheme is shown, how the TCR, present time, Avg. speed, previous data and the ANN collectively work to select the most favorable route using neuro-fuzzy system.

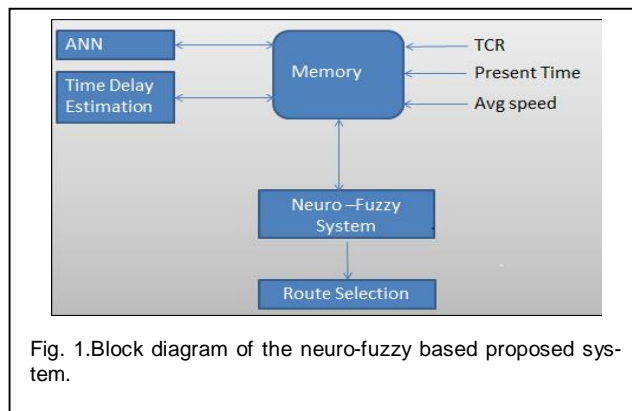


Fig. 1. Block diagram of the neuro-fuzzy based proposed system.

2.1 Neuro Fuzzy Inference System.

It is hybrid approach where neural and fuzzy advantages are combined for the parallel execution. The arrangement contains the structure of fuzzification and de-fuzzification. They both work with combined method.. The conforming grid has enhanced results as compared with the earlier work. The process of transforming crisp values into grades of membership for linguistic terms of fuzzy sets

Fuzzification: In it, the crisp values of the fuzzy system are converted into grades of member functions.

Defuzzification: It is the reverse process of above to create a defuzzified value for the system.

2.2 Artificial Neural Networks (ANNs)

Scheming and executing intellectual systems is a complex factor for innovation and developments of enhanced products. Artificial neural networks are the approach that works highly parallel computations as compared to our current machines. For a dynamic system, these networks are highly attentionable and can resolve problems that are not possible in linear computation. A very attractive attribute of these networks is, they are adaptive in nature. If a dynamic application is required, then these networks are considered most suitable due to adaptive nature and decidedly parallelism.

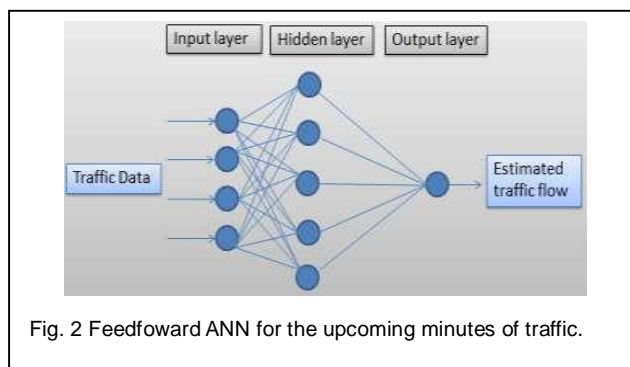


Fig. 2 Feedforward ANN for the upcoming minutes of traffic.

Neural networks are being engaged in various applications including handwritten character recognition, human mind simulation and patterns appreciation [14]. ANNs

have different types and these networks are trained, grounded on the training data to perform a specific task.

2.3 Ant Colony Optimization Algorithm

Ant colony optimization (ACO) is a probabilistic technique that disentangles the cost for most favorable path selection. In ACO algorithm, artificial ants simulate intelligent behavior for discovering the optimized cost of the route selection. These ants are capable of solving complex problems like to find food using a chemical substance called pheromones. As a single ant moves between its nest and the source, it deposits some pheromones quantity on that route. This pheromones quantity contains the path intensity to be adopted for other ants. As the pheromones, quantity is being added frequently, becoming more noticeable for the ensuing ants.

An ant naturally adopts the way where the pheromones quantity is highly concentrated. With the passage of the time, if ants are not following the highly concentrated path, the pheromones quantity start decreasing, causing the path to be less predicted. Those trails that are navigated (to food and back) most quickly soon become the trails with the highest intensity of pheromones, thus boosting the next ants to make choice of these routes.

It is not necessary that the ants are bound to follow the route with the highest intensity of pheromones. Ants can disregard the previous route and search food on all other possible routes. If the ants are able to find a shortest path towards food, then the current solution is swapped with the previous one with highest pheromones quantity. As a new shortest path is defined and updated, the pheromones quantity will start evaporation with time, causing the path is less anticipated for the next ants and resulting the longer path to be skipped.

Route selection based on the above mechanism is the prime factor of ACO. The detailed description of ant behavior as it narrates to ACO is provided by Dorigo et al [15].

3 DESIGN ALGORITHM OF PROPOSED SYSTEM

On the user defined parameters, for every single vehicle the proposed system is executed. In this system, the traffic indication is delivered by the TCR that contains present traffic data and is updated dynamically. This system consists of average speed of vehicle, previous vehicles data, present time and the environment monitoring, no of lanes etc. ANNs are used for the future traffic minutes. The ANNs are trained as traffic predictor are used in this system causing helpful to the system with use of TCR and all other inputs to the system. The next section defines in details about the strategy algorithm of the system.

Initialize: it comprised of the initial values of the parameters such as number of ants and evaporation co-efficient etc.

LocateAnts: Ants are located on the origin. an ant which is on the way informs other ants, either it is congested or still on the way to destination. It is must that each ant can cross each junction only once. If an ant is blocked some where due to congestion or road blockage, no chance to continue further then will move back to source.

Paradigm 'p':

During this phase, cost based probability for each route is calculated for all active ants. This probability is calculated from the route source to. For each ant, here it is elaborated with k.

$$p_{ij}^k = \begin{cases} \frac{\tau_{ij}^\alpha \prod_{l \in \text{parameters}} \xi_{ijl}^{-\alpha_l}}{\sum_{k \in \text{tabu}_k} \tau_{ik}^\alpha \prod_{l \in \text{parameters}} \xi_{ikl}^{-\alpha_l}} = j \in \text{tabu}_k \\ 0 & \text{else} \end{cases}$$

Where, τ_{ij} denotes the pheromone intensity from source to destination 'i' and 'j' respectively. Constraint, α is showing the the route [15, 16], tabu_k is showing the all routes that are not suitable for all ants to be adopted in future. ξ_{ijl} and ' l ' are showing the cost of constraint l for route ij and the worth level of constraint l respectively. In this process list is consisted of all the jammed routes and the constraints are the parameters for the user to adopt the route according to his customized way. Currently the six most important parameters "Distance", "Trafficflow", "Environment monitoring", "Width", "Roadcondition", and "Trafficlights" are used in this system. However, it can be enhanced by incorporating other parameters that are desirable such as "Road Quality" and "other desired facilities like hospital" etc. [8-10]. All these parameters have different significance impact from i to j ; distance ij , road-condition ij , Env-monitoring ij , Traffic lights ij , Traffic flow ij , width ij .

Distance: It is a distance between two junctions from the source to target at time t .

Traffic flow: It is load of traffic on the routes from the source to target at a time t .

Environment Monitoring: It is the environment condition on the different routes from the source to target at a time t .

Width: It is the total no of lanes on the all-possible routes at a time t from the source to target.

Traffic Lights: It is the total no of the lanes on all possible routes from the source to target at a time t .

Road Condition: it is the scale of the road condition that is available at a specific time t on all possible routes from the source to target

Select Route: a random parameter $0 \leq p \leq 1$ with uniform probability is matched with the parameter ' P ' where $0 \leq P \leq 1$ and mostly fixed to 0.90. The comparison result

between 'P' and 'p' choices one from these.

- If 'p' is larger than 'P' the ant adopts the route with highest pheromones intensity.
- If 'p' is smaller than 'P' or equal, then ant opts the route with possibilities.

Update Tabu List: In this process, the route is updated and added to the list, causing other ants not to select that path. If an ant is on the way to source and is blocked due to traffic congestion or any other reason, that path is deactivated from the route for the following ants in order to provide the route reliability. More over its probability is not considered any more.

Update Pheromone: Ant colony system consist of two main rules: first is updation of local pheromone and the other rule is updation of the global pheromone. This rule is applied after all the traffic is travelled to the destination. The pheromone amount is added to the route that is successful completed by an ant as,

$$\tau_{ij}^{new} = \tau_{ij}^{old} + (10 \times \Delta \tau)$$

Where, ' $\Delta \tau$ ' is the amount of local pheromone updating. $\Delta \tau$ is the amount of pheromones of our neuro-fuzzy system. Six parameters of this system use three member functions that are taken by our systems and the pheromone intensity was used between '0' to '5'. If all parameters are in the most ideal situation then the pheromone intensity is maximum else varying between '0' to '5'.

It is the final step of the algorithm to keep posted the global pheromone.

$$\tau_{ij}^{new} = p \tau_{ij}^{old}$$

Where p is the evaporation constant and is usually set to 0.9.

In Table 1, some rules for the system are discussed. If all the parameters are found at their ideal values then the intensity of the pheromones is extraordinary. If three parameters comprised of maximum value and the other three contained minimum value then the pheromones intensity will be 3. If all the parameters have minimum degree then the pheromones intensity is minimum and the lowest probability to adopt that path. If the pheromone intensity is high then it is decidedly observed to adopt that route.

Select best route: after m loops, route with the minimum

TABLE 1
PARAMETERS RELATIONAL TABLE OF THE SYSTEM

Rules	If						Then
Rule no	Distance	Traffic	R.C.	E.M	Lights	Width	Pheromns
1	low	high	high	Good	low	max	6
2	low	high	low	Good	medium	max	4
3	low	high	medium	Good	medium	max	5
4	medium	medium	medium	Avg	medium	medium	3
5	medium	medium	high	low	low	low	2
6	high	low	low	low	high	low	0
7	high	low	high	Low	low	low	1

cost is suggested by the system.

4 SIMULATION RESULTS AND DISCUSSION

According to the scheme of the neuro-fuzzy system for the route selection is designed and simulated on Matlab. All the parameters have significant effect on the system's output. A neuro-fuzzy system is designed for the optimum route selection between two nodes, called the origin and the destination.

In the fig.3, six parameters are used as inputs to the Neuro-fuzzy system. The system makes three member functions of each parameter itself. The rules for the system are denser due to six inputs and three-member function against each input as shown in the figure 3. The output member functions are also denser and the output is generated after all the computation is done.

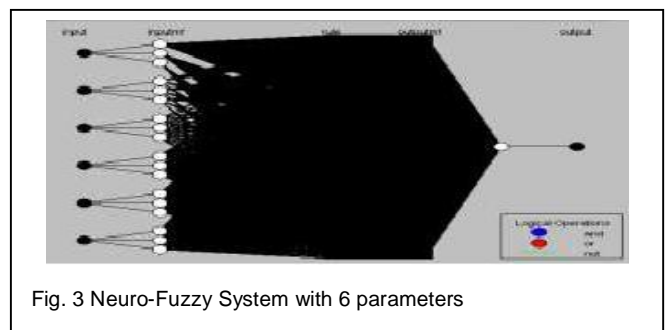


Fig. 3 Neuro-Fuzzy System with 6 parameters

In the following table, we showed the relational parametric relational among all constraints of the system.

In the fig.4, Some data is given to the system for the working out of the proposed neuro-fuzzy system. Total no of nodes 1503, participated during the training. In which there are 54 nonlinear parameters, 5103 linear parameters and 729 fuzzy rules are used.

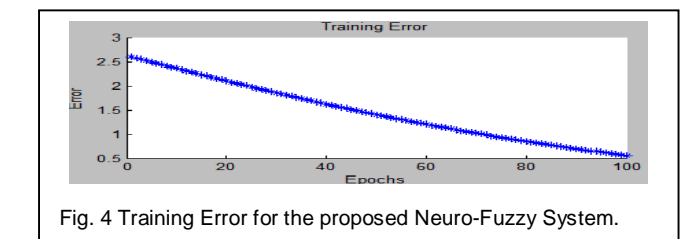


Fig. 4 Training Error for the proposed Neuro-Fuzzy System.

ANFIS training using back propagation technique completed at epoch 130 and the error is 0.56108, is too small.

Fig 5 shows blue circles for the data, offered for the testing purposes to the system and red stars are the outputs against the given inputs to the system. In the testing data fifteen pairs are randomly selected and presented to the system. Red stars are the neuro-fuzzy outputs of the testing data. The average testing error is 2.4827, remains too low and acceptable.

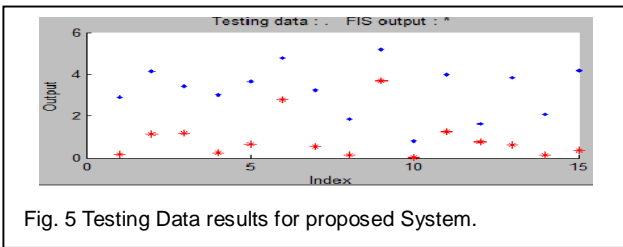


Fig. 5 Testing Data results for proposed System.

In Fig 6 the checking data was presented to the system. Total no of pairs that are presented to the system, are 22 and the FIS output is shown against those pairs. Results against the checking data pairs are satisfied and the error is 0.54811.

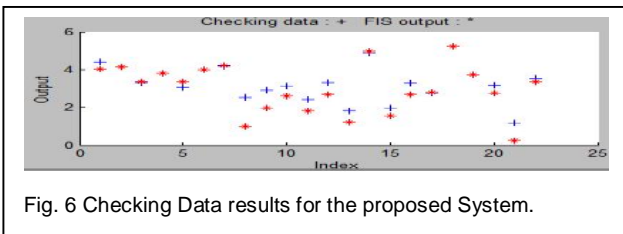


Fig. 6 Checking Data results for the proposed System.

In fig 7 surface graph among inputs road condition and traffic flow against output is shown. When the road condition is at the ideal value the pheromones amount will be max against these inputs that will enhance the probability to adopt that route against two parameters,

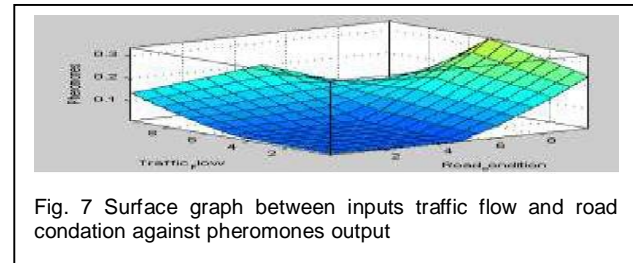


Fig. 7 Surface graph between inputs traffic flow and road condition against pheromones output

In figure 8,9,10 and 11 different parameters combination impact on the output is shown. These parameters impact on the pheromones are customizable.

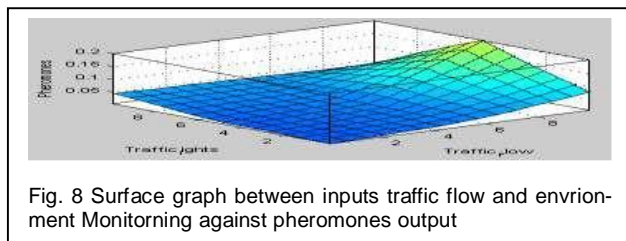


Fig. 8 Surface graph between inputs traffic flow and environment monitoring against pheromones output

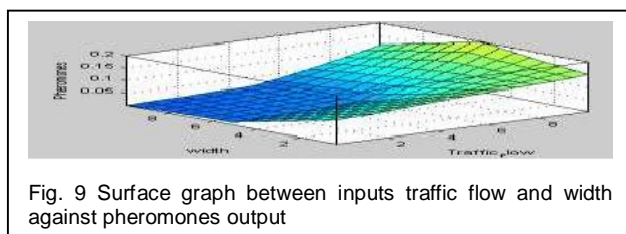


Fig. 9 Surface graph between inputs traffic flow and width against pheromones output

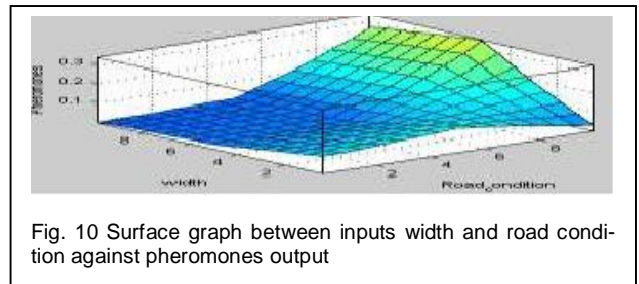


Fig. 10 Surface graph between inputs width and road condition against pheromones output

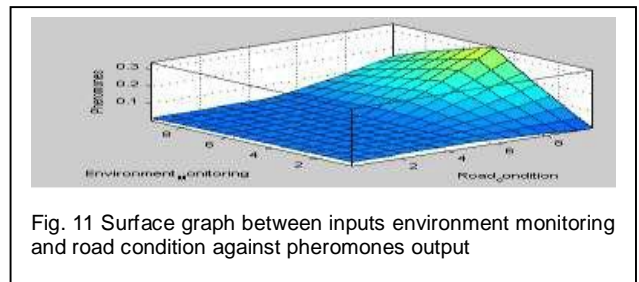


Fig. 11 Surface graph between inputs environment monitoring and road condition against pheromones output

Fig. 12 shows that Input variable traffic flow impact is directly relational to the pheromones intensity. Greater the intensity of the environment monitoring, the more impact will be on the pheromones intensity. Like this all others variables impact can be seen on the output variable.

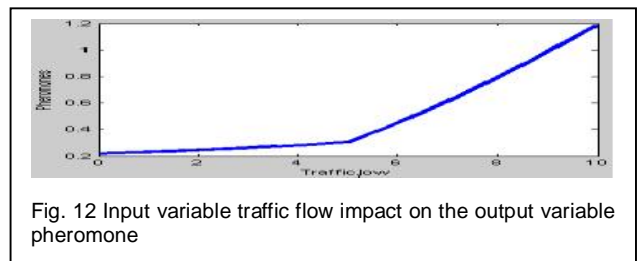


Fig. 12 Input variable traffic flow impact on the output variable pheromone

5 CONCLUSION AND FUTURE WORK

In this proposed system, a neuro-fuzzy system is carried out to represent the characteristics of the constraints owing to select the most encouraging route according to the customized preferences of the driver who is utilizing this system. The proposed system is aware of changing the route dynamically when a congestion or obstructive traffic is perceived on the selected route. A route ranking can be provided to the user forecasting the decision making more adaptive and naturalist. This work will reduce the cost handsomely in order to select the route according to the user's customized preferences.

In this proposed system, costs of all the possible routes are studied to accept the favorable one. In conclusion, the route with the lowest cost is adopted using our system. This research can be continued for day-to-day life practices, by employing vehicle to vehicle (V2V) Communication. In future microelectronics, state of the art technology can be used to develop the hardware of this system.

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