

# ID 25: Benchmarking Reliability of Deep Learning Models for Pathological Gait Classification

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## Routine Walking

- Gait disorders increase from 10% to **more than 60% in the elderly between the ages of 60 and 80**.
- Gait abnormalities severely affect the Quality of Life (QoF).
- Altered gait could represent a lingering neurodegenerative condition.

### Geriatric Gait - At home Needs

- Regular hospital commutes difficult in remote areas.
- Early diagnosis requires detection in home settings.**
- Vision based detection is suitable for home settings.

## Proposed Method

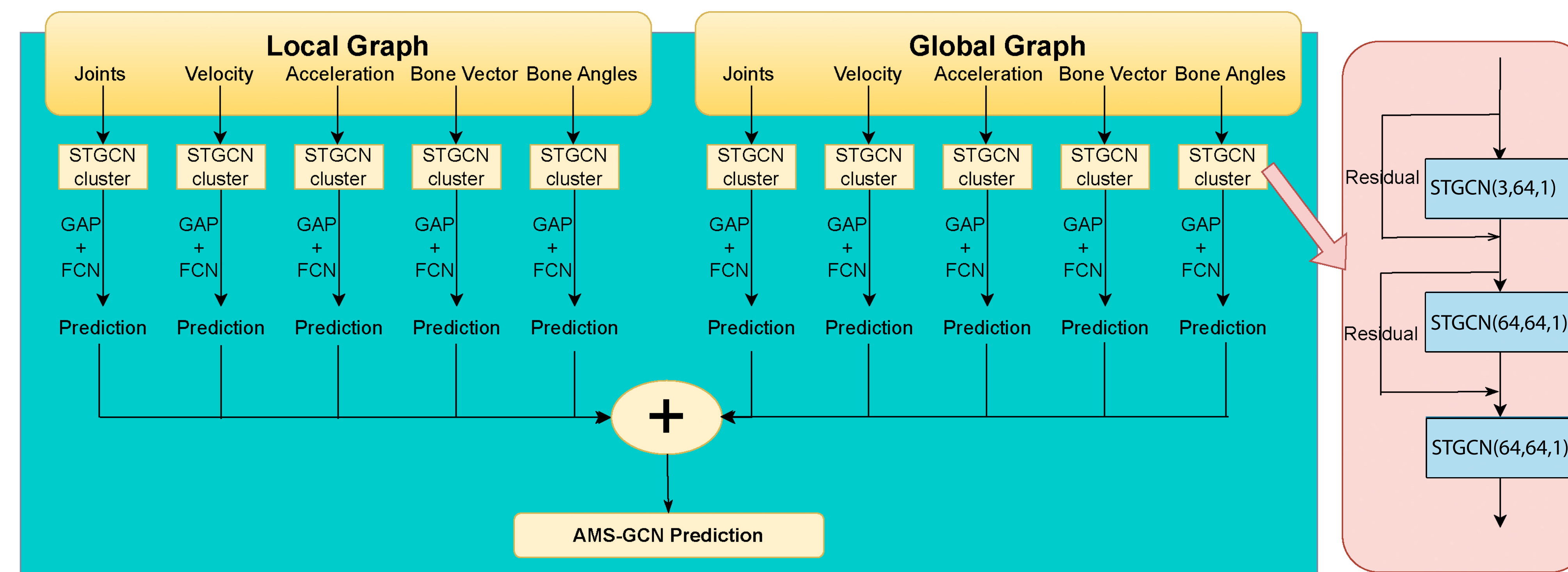


Figure 2: AMSGCN Model

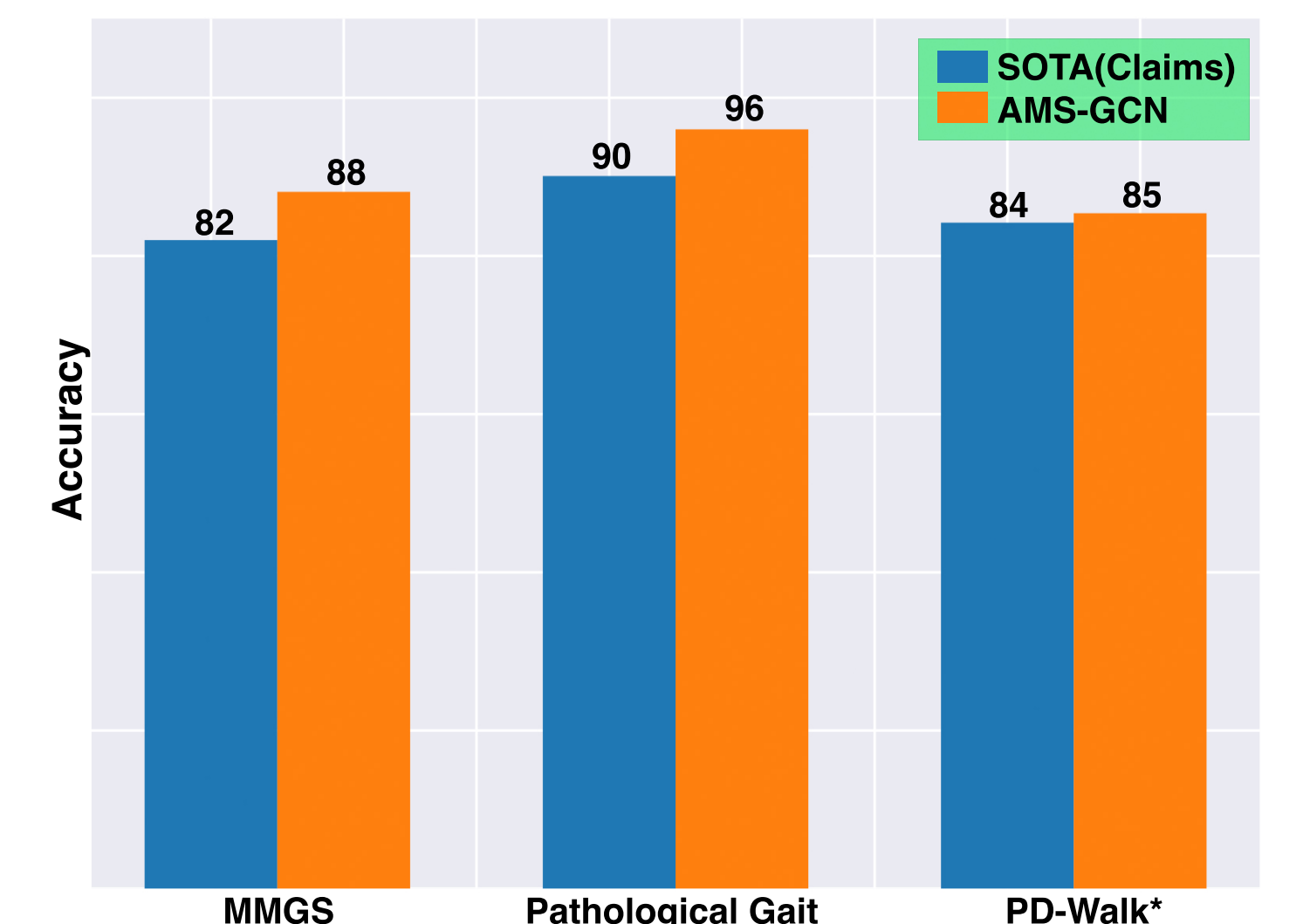


Figure 3: AMSGCN Results

## AMSGCN

We present a robust baseline to ensure stable performance on multiple datasets.

- Multi-Stream STGCN network that incorporates global and local joint connection branches
- Consistent validation and test scores across multiple dataset

Table 1: AMSGCN Input feature comparison - best and worst feature combinations and respective F1 scores.

| Method                            | Worst F1                  | best F1                            |
|-----------------------------------|---------------------------|------------------------------------|
| Multi-Modal Gait Symmetry Dataset |                           |                                    |
| 1                                 | Acc(56.0)                 | <b>boneA(86.0)</b>                 |
| 2                                 | Vel+Acc(55.0)             | boneL+boneA(87.0)                  |
| 3                                 | Joint+Vel+Acc(77.0)       | Acc+boneL+boneA(87.0)              |
| 4                                 | Joint+vel+Acc+boneL(82.0) | <b>Joint+Acc+boneL+boneA(86.0)</b> |
| PD-Walk Dataset                   |                           |                                    |
| 1                                 | Acc(17.0)                 | <b>boneL(87.0)</b>                 |
| 2                                 | Vel+Acc(17.0)             | boneL+boneA(86.0)                  |
| 3                                 | Vel+Acc+boneA(83.0)       | Acc+boneL+boneA(86.0)              |
| 4                                 | Joint+Vel+Acc+boneA(85.0) | <b>Vel+Acc+boneL+boneA(86.0)</b>   |
| Walking Gait Dataset              |                           |                                    |
| 1                                 | Joint(41.0)               | Acc(62.0)                          |
| 2                                 | Vel+Acc(54.0)             | Joint+Acc(66.0)                    |
| 3                                 | Joint+boneL+boneA(55.0)   | Vel+Acc+boneA(71.0)                |
| 4                                 | Joint+Vel+Acc+boneL(63.0) | <b>Joint+Acc+boneL+boneA(64.0)</b> |

## Performance Insights

Is any one feature more important?

- Weighted feature combination did not provide consistent improvement
- Different individual features shine for different datasets but using more features tapers this gap
- Direct combination performance improvement was visible even with reduced inputs

## What Next

Clinical viewpoint on the proposed results  
Collaborations and Postdoc?

## Contact Information

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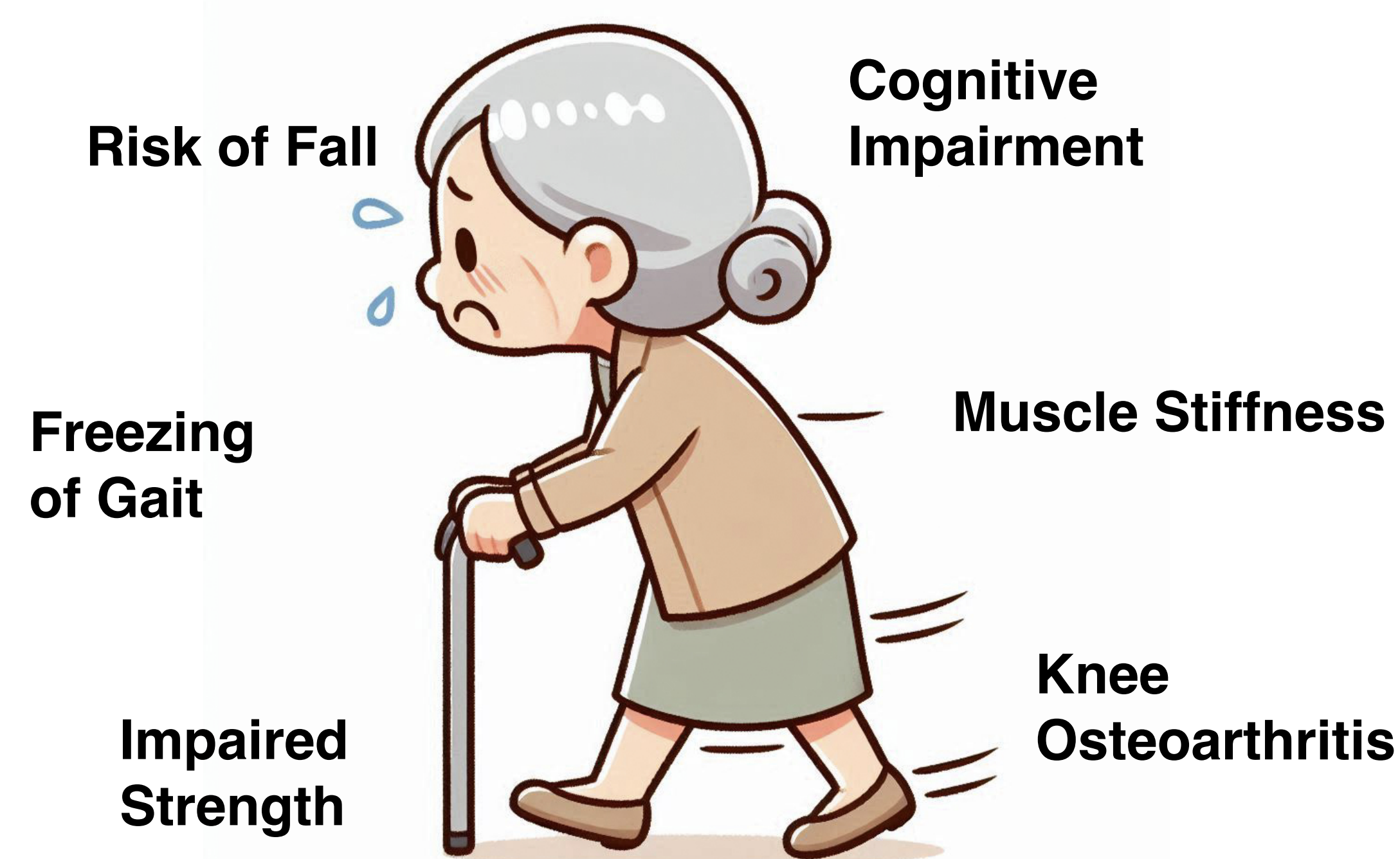
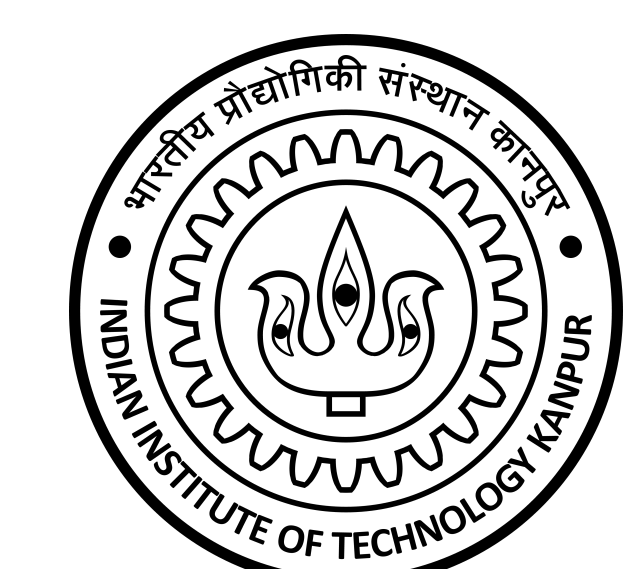
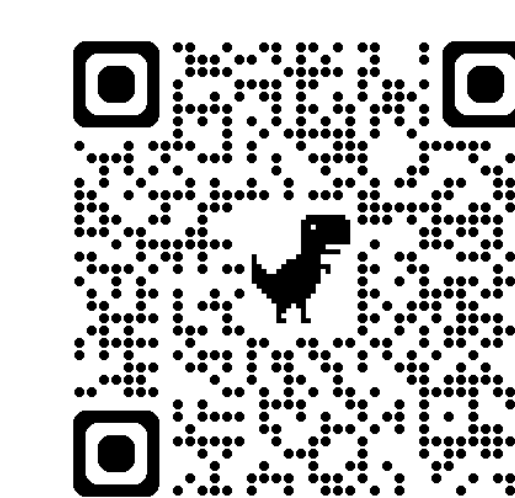


Figure 1: Elderly Gait

## Existing Space of Work

- Spatio-Temporal Graph Convolutional Networks (STGCN) are well suited to model human pose
- Many existing models are tested only on a single gait dataset

## Benchmarking Study

- Previous evaluations often lack proper validation
- Over-parameterized models like STGCN tend to overfit gait data