INTRODUCTION

Recognizing the 14 stages in the rat testis cycle is a difficult task [1] (fig. 1). We [2] have developed a deep learning (DL) based method to automate the staging which is easy to use, and provide results comparable to an expert on normal testes, and to historical data [3]. Taking this ahead, we propose a DL-based method for identification and quantification of findings in rat testes.

We developed DL models to detect those findings and we compared them to the pathologist's results (table 2).

We have concentrated on a subset of findings from the INHAND nomenclature [4]. We selected several Janssen toxicology studies on which early stages findings (spermatid retention after 1 month) and more chronic, degenerative findings were recorded (table 1).

We have trained various deep learning networks by using Hist-Net, Efficient-Net and YAMU-Net based architectures by training on 512x512 tiles from abnormal images having the abnormalities viz. tubular degeneration, tubular dilatation, degeneration/necrosis, vacuolation, multinucleated giant cell and spermatid retention [5,6] (fig. 2, 3, table 3).

MANUAL METHOD & DESIGN OF THE ALGORITHMS

A rule-based approach was followed for detecting stagespecific abnormalities viz. spermatid retention and degeneration/necrosis of spermatids. Then, we calculated the per⁻ centage of affected tubules with respect to total tubules present in the testis.

For validation, slide level findings provided by pathologists on data from two studies $(X, 1 \text{ month and } Y, 3 \text{ months})$ were compared with algorithm results, and statistical analysis was performed.

> 5. Samanta & al (2021) [CAN for Semantic Labeling in Histopathology Images; IEEE 18th ISBI] [doi:10.1109/ISBI48211.2021.9433905](http://doi.org/10.1109/ISBI48211.2021.9433905)

6. Samanta & al (2022) [Yet Another Modified U-Net Architecture for Semantic Segmentation; MIDL 2022] see openreview.net/forum?id=RPFCw3VU6R9

DL recognition. Fig. 5 (left): Retention, Spermatid in stage XI tubule. Fig. 6 (right): Degeneration, Germ Cell/Necrosis, Tubular

DL recognition. Fig. 7 (left): Degeneration, Tubular. Fig. 8 (right): Dilatation, Tubular

DL recognition. Fig. 9 (left): Multinucleated Giant Cells. Fig. 10 (right): Vacuolation, Tubular

COMPARING MANUAL & DL RESULTS

Quantitative outputs of the algorithm correlate with semiquantitative grades provided by the pathologist (table 4, 5). The proposed method can accurately detect and quantify spermatid retention, degenerative tubules, atrophic testes and other findings in normal or abnormal testes (table 6).

CONCLUSION

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> The DL method to quantify stage frequencies in H&E slides was easier and quicker than manual staging, and was in accordance with published material and with the pathologist's results. It also performed well on abnormal testes. The DL methods to detect and quantify findings like spermatid retention, vacuolation and degeneration changes also correlated well with the pathologist's grading, with a better sensitivity.

> We are preparing future developments to identify other abnormalities and generalise the method to other species (dogs, mouse, NHP).

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Fig. 1. The cycle stage [3] showing the cell associations. For example, in stage VIII, there are preleptotene and pachytene spermatocytes, round spermatid (step 8) and a mature elongated spermatid step 19 being released. In stages IX-XI, the spermatid retention is symbolized by presence of unreleased last step mature spermatid (in red)

Fig. 2. Solution approach: algorithm workflow

Fig. 3. Hist-Net Architecture used for segmentation tasks

Fig. 4. DL recognition of testis tubules and stages / group of stages. There was a perfect concordance between the historical stage frequency reported in [2], and the stage frequency in the study [3] as assessed by an expert and by the current stage-detection DL model

Deep Learning Models for Quantifying Testicular Toxicity in Rats

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* 0: no abnormality detected; 1: minimal; 4: marked; 5: severe (on a severity scale of 5)

Table 5. DL recognition of tubular degeneration/atrophy compared to the pathologist's assessment in studies X (1 month) and Y (3 months). Extract of unilateral results sorted by increasing %

Table 2. Example of descriptive/semi-quantitative results generated by the pathologist. Study Y (3 months); animals 1, 3-10, 35-37 & 40 were normal and are not reported in the table. (u: unilateral)

Table 6. Performance parameters of the AI models

* P: present, not graded; 2: mild; 3: moderate (on a severity scale of 5)

Table 4. DL AND pathologist's assessment of spermatid retention in studies X (1 month) and Y (3 months). Extract of unilateral results (except untreated group 1 animal 13, which show an incidental tubular degeneration in 1 testis) sorted by increasing %

Table 1. List of INHAND terms for the testis [4]. In bold, the findings for which the DL models were developed

Table 3. CNN selected depending on the finding

