

Aggregate Trends of Apolipoprotein E on Cognition in Transgenic Alzheimer's Disease Mice

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Abstract.

Background: Apolipoprotein E (*APOE*) genotypes typically increase risk of amyloid- β deposition and onset of clinical Alzheimer's disease (AD). However, cognitive assessments in *APOE* transgenic AD mice have resulted in discord.

Objective: Analysis of 31 peer-reviewed AD *APOE* mouse publications ($n = 3,045$ mice) uncovered aggregate trends between age, *APOE* genotype, gender, modulatory treatments, and cognition.

Methods: *T*-tests with Bonferroni correction (significance = $p < 0.002$) compared age-normalized Morris water maze (MWM) escape latencies in wild type (WT), *APOE2* knock-in (KI2), *APOE3* knock-in (KI3), *APOE4* knock-in (KI4), and *APOE* knock-out (KO) mice. Positive treatments (t+) to favorably modulate *APOE* to improve cognition, negative treatments (t-) to perturb etiology and diminish cognition, and untreated (t0) mice were compared. Machine learning with random forest modeling predicted MWM escape latency performance based on 12 features: mouse genotype (WT, KI2, KI3, KI4, KO), modulatory treatment (t+, t-, t0), mouse age, and mouse gender (male = g_m; female = g_f, mixed gender = g_mi).

Results: KI3 mice performed significantly better in MWM, but KI4 and KO performed significantly worse than WT. KI2 performed similarly to WT. KI4 performed significantly worse compared to every other genotype. Positive treatments significantly improved cognition in WT, KI4, and KO compared to untreated. Interestingly, negative treatments in KI4 also significantly improved mean MWM escape latency. Random forest modeling resulted in the following feature importance for predicting superior MWM performance: [KI3, age, g_m, KI4, t0, t+, KO, WT, g_mi, t-, g_f, KI2] = [0.270, 0.094, 0.092, 0.088, 0.077, 0.074, 0.069, 0.061, 0.058, 0.054, 0.038, 0.023].

Conclusion: *APOE3*, age, and male gender was most important for predicting superior mouse cognitive performance.

Keywords: Age, Alzheimer's disease, *APOE*, cognitive function, gender, Morris water maze, treatment

INTRODUCTION

Alzheimer's disease and the Apolipoprotein E gene

Alzheimer's disease (AD) is a neurodegenerative disorder that gradually impairs cognition in patients suffering from the condition [1] and is considered to be a polygenic disease [2]. The Apolipoprotein E

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(*APOE*) gene synthesizes various isoforms of apoE, a biomarker strongly hypothesized to be implicated in AD, as initially observed by researchers at the Duke Alzheimer's Disease Research Center in 1993 [3–5]. The biochemical exhibits three common isoforms in humans: apoE2, apoE3, and apoE4, of which apoE2 [6] and apoE3 [7] have been shown to confer protective effects, while apoE4 has been documented to increase cognitive impairment [7–10]. The protein apoE primarily functions in brain lipid transport, glucose metabolism, and neuronal signaling, among other roles [11]. Its biochemical pathway involves binding to surface apoE receptors on cells throughout the body to initiate further metabolism, transport, and signaling, but the single amino acid differences between apoE2, E3, and E4 result in a diversity of protein structures that, therefore, affect receptor binding in different ways respective to each isoform [12]. For instance, apoE3 and E4 display 50-fold greater binding affinity when compared to apoE2 [12].

Moreover, through yet-uncovered mechanisms, *APOE* isoforms are also believed to accentuate or attenuate pathways involving other well-established AD biomarkers, such as amyloid- β (A β) [13] and p-tau [14]. The amyloid cascade hypothesis posits that the deposition of A β plaques from improper cleavage of amyloid- β protein precursor is the most important molecule in AD pathogenesis, but heterogeneity of results in the literature has challenged this notion [15, 16]. For example, Foley et al. showed that mice with increased A β levels do not perform significantly worse on cognitive assessments than mice with lower levels of A β , suggesting that the plaques could instead be a side effect of AD pathogenesis [13]. The interaction between *APOE* isoforms and amyloid metabolism was examined to show that *APOE4* was associated with elevated A β secretion and defective A β uptake compared to isogenic *APOE3* human brain cells [8]. This interplay of *APOE* isoforms with the amyloid cascade further strengthens the literature trend that *APOE4* confers negative effects on cognition since conversion of apoE4 to E3 repeatedly mitigated various pathologies related to AD [8]. Lastly, meta-analysis work revealed that p-tauopathy is a greater predictor of cognitive decline than A β in mouse models [14].

AD APOE mouse model

Endogenous mouse apoE is structurally different from human isoforms, so the most prominent *APOE* mouse models use human isoforms for study

[13, 17]. The selection of appropriate promoters and other regulatory elements is important to note since *APOE4* expressed in neurons have demonstrated more detrimental effects compared to its expression in astrocytes [18]. Furthermore, mouse *APOE4* knock-in (KI) groups display neural plasticity defects that appear to be absent in *APOE3* KI groups [19–21].

Genotype effect on AD

Extensive research has been conducted on transgenic mouse models to investigate the effects of various features, such as genotype, treatment, age, and gender on AD pathogenesis. Because of the high-dimensional nature of AD diagnosis, there is heterogeneity within the literature about the importance of such features [13]. For example, Shi et al. demonstrated that *APOE4* KI mice exert a gain of toxic function while *APOE* knock-out (KO) mice instead displayed protective effects [22]. However, others have found conflicting results among the same genotypic groups. Zerbi et al. found that both *APOE4* KI and *APOE* KO mice demonstrate a comparable magnitude of brain functional connectivity deficiencies in progressing from adulthood (12 months) to old age (18 months) [23]. Thus, uncovering aggregate trends is pivotal to AD *APOE* research. But genotype is not the only feature whose effect on AD is still under debate.

Treatment effect on AD

The effects of various forms of treatment have shown similarly unclear findings. While prior studies have tested treatment regimens that may be promising avenues for remediating AD progression in transgenic mouse models [24], the overall efficacy of treatment is still undetermined. For instance, pharmacological targeting of the apoE/A β interaction pathway significantly reduced aggregation of both apoE and A β in *APOE2* and *APOE4* KI mice, preventing memory decline in both groups [24]. However, Cramer et al. demonstrated that oral administration of bexarotene, a retinoid X receptor agonist, reversed various cognitive deficiencies observed in WT mice but without significant effect on *APOE* KO groups [25]. The complicated interplay between *APOE* and other AD biomarkers makes it difficult to discern whether *APOE* modulatory therapy could successfully lessen cognitive decline in AD. Furthermore, treatments that have been investigated in transgenic AD mice have not translated well to human

trials. Such difficulty in translating *APOE* therapies from mice to humans is likely due to inherent differences in *APOE* between species. Humans express three major *APOE* isoforms, while only one is naturally observed in wild type mice [11].

Gender and age effect on AD

Female gender and old age are frequently cited in human studies as probable AD risk factors. However, it is difficult to assess whether gender has a significant effect on cognition due to confounding variables such as life expectancy, innate differences in performance on cognitive tasks, and difference in hormone levels [26]. Although many studies have shown that AD may occur more often and more severely in women [27–29], the underlying mechanism by which women are more predisposed than men remains undiscovered [26, 30–33]. Meanwhile, incidence of AD increasing with age is perhaps the most substantiated pattern documented in the literature [34–37]. However, even this trend is still under scrutiny due to a lack of understanding of how the effects of natural aging can be compounded by other non-age-related features to result in increased AD risk for certain populations [37].

Scope of present study

A complex and difficult to interpret relationship exists between age, gender, treatment, and genotype on cognition in AD, making it challenging to develop broadly effective treatments through diet, vitamin supplementation, exercise, or other specific apoE modulatory drugs. Though age has most clearly and consistently been shown to be the biggest risk factor for developing AD [37], the relationships between both age and cognition [34–37] as well as gender and cognition [26, 30–33] are less well known. The effects of various *APOE* treatments are also similarly disputed [24, 25, 38]. Likewise, categorizing genotypes by the effect they have on cognition has yielded inconsistent results [19–23]. Therefore, there is a need for a deeper understanding of the relationship between various features that contribute to AD pathogenesis across a large, aggregate set of data from multiple published *APOE* transgenic mouse experimental studies. The objective of the present study was to ascertain aggregate trends in AD *APOE* mouse literature by performing statistical analysis machine learning feature importance modeling to ascertain the extent that cognition in AD *APOE* mice is affected

by *APOE* genotype, *APOE* modulatory treatment, mouse gender, and mouse age.

METHODS

The goal of this work was to perform an aggregate analysis of cognition in *APOE* transgenic AD mice to determine the importance specific features, namely mouse genotype, type of *APOE* modulatory treatment, mouse age, and mouse gender, in predicting cognition as measured via Morris water maze (MWM) escape latency. PubMed database (<http://www.ncbi.nlm.nih.gov/pmc/>) searches identified peer-reviewed publications that were manually reviewed to determine if their data met criteria for inclusion into the present study. All quantifiable experimental data from tables and figures were transcribed into a relational database with resultant transcription accuracy in excess of 98.8% under consistent oversight by meticulous quality control using a published protocol [39].

Inclusion criteria

PubMed was queried using the key search terms “Alzheimer’s Disease”, “*APOE*”, and “Morris Water Maze” in addition to various synonymous combinations and abbreviations. To be included in the present study, each publication must have included MWM escape latency results at baseline and after 4 or 5 days of training; inclusion of a control and treatment group; wild type genotype, *APOE* KI genotype (namely *APOE2*, *APOE3*, or *APOE4* knock-in), or *APOE* KO genotype; demarcation of mouse gender; demarcation of mouse age at time of MWM testing. A total of 31 peer-reviewed papers met all requirements for inclusion. Figure 1 displays a PRISMA Flow Diagram for the present study.

Morris water maze

The MWM is used to test spatial memory as a measure of cognitive functioning and was first developed by Richard Morris in the early 1980s [40]. MWM is a standard metric of cognitive performance in mouse models, and due to its widespread use in pre-clinical AD studies [40, 41], MWM escape latency was chosen as the outcome metric for mouse cognition in the present study. In the beginning of the MWM, a mouse is placed into a pool of cold water (13°C) that contains a visibly discernible platform [42, 43]. On the first day of training, the mouse learns to find the

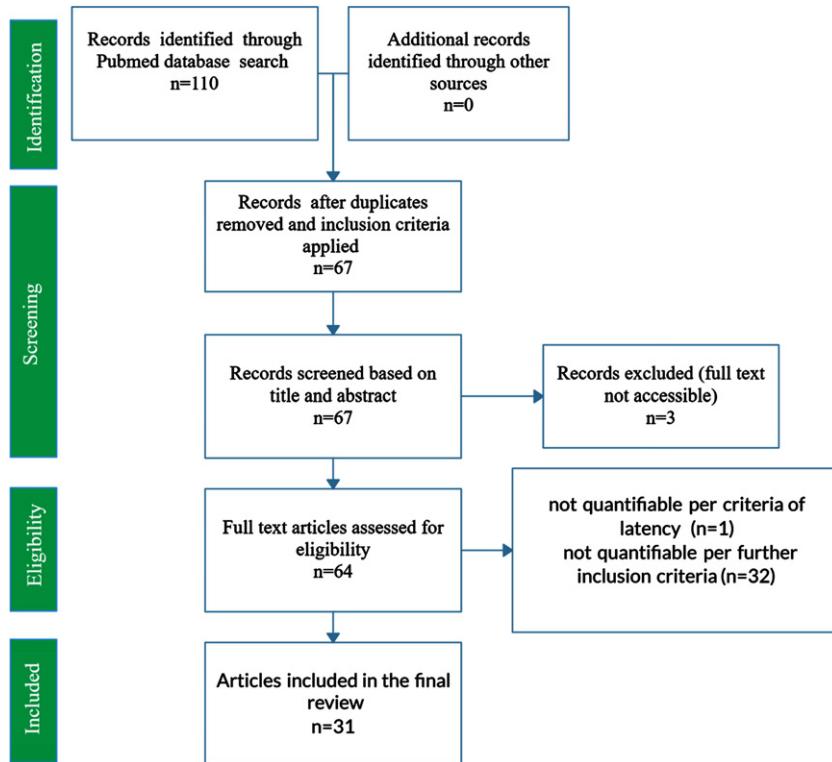


Fig. 1. PRISMA Flow Diagram for the systematic review of PubMed articles related to AD APOE. The PRISMA flow diagram represents the systematic review of PubMed articles and the compilation and curation of journal article data into a manually constructed relational database. The final set of included studies comprised 31 journal articles and 3,045 mice. Features assessed include mouse genotype (wild type, *APOE2* knock-in, *APOE3* knock-in, *APOE4* knock-in, *APOE* knock-out); type of external *APOE* modulatory treatment and its intended impact on the underlying etiology and corresponding cognition (positive treatment, negative treatment, untreated); mouse age (in days); and mouse gender (male, female, or mixed/unknown).

visible platform and, thus, escape the cold-water bath. Subsequently, the mouse is reintroduced to the maze with the platform hidden under the water, instead. As a result, the mouse must rely on visible cues outside of the maze to locate the platform. Training with the hidden platform usually takes place approximately four times daily over the course of four to seven days. The amount of time it takes for the mouse to find the hidden platform is measured as escape latency [41]. To prevent skewed results from over-training in the MWM, only the first 5 days of training trials were included for analysis. Overtraining due to either too many days of training or too many trials per day can cause all of the mice in the experiment, whether wild type or transgenic, to display similar escape latency values that inhibit quantitative assessment of learning differences [44]. Trials on the first day of training are used as a control group between mice to account for individual swimming performance [42, 43]. Supplementary Table 1 includes the specific description of

the MWM used in each data source included in the present study.

Data sources

The C57BL/6 mouse model is the most widely utilized genetic background of altered mice for study of human pathology due to its ability to maximally express most mutations [45]. In addition to being relatively easy to breed, the mouse model is highly sensitive to cold temperatures, making it particularly useful for examining cognition through the MWM [46]. The C57BL/6 mouse model subline was originally isolated at the Bussey Institute and most extensively distributed by the Jackson Laboratory; however, as a result of its ubiquity in research, many groups have bred colonies in isolation from one another for many generations, potentially exacerbating the effects of genetic drift, the random change in allele frequency in a population over subsequent

Table 1

Mean normalized escape latency and standard deviation is the age-normalized Morris water maze escape latency in seconds. Mice n is the sample size of the group, whereas study n is the number of studies utilized for the group. Bracketed numbers in the sources column correspond to full data source references. There were 1,430 female mice, 1,181 male mice, and 258 mixed/unknown gender mice

Group	Mean Norm. Latency (s)	Std. Dev. Norm. Latency (s)	Mice (n)	Data Sources (n)	Data Source References
KI3_t-	37.6	—	8	1	[48]
KI4_t-	30.9	11.0	66	3	[48–50]
KO_t-	47.7	6.2	62	4	[49, 51–53]
WT_t-	32.8	9.5	203	5	[49–53]
all_t-	35.3	11.0	339	6	[48–53]
KI2_t0	38.2	5.4	354	3	[34, 54, 55]
KI3_t0	30.6	9.0	677	16	[34, 48, 54–67]
KI4_t0	48.6	15.3	738	19	[34, 49, 54–57, 59–71]
KO_t0	40.9	15.3	337	10	[34, 49, 53, 58, 63, 72–76]
WT_t0	35.8	17.2	326	11	[49, 50, 53, 63, 64, 68–70, 72, 74, 75]
all_t0	39.3	14.9	2432	28	[34, 48–50, 53–76]
KI3_t+	29.2	5.2	61	5	[56, 61, 64–66]
KI4_t+	33.5	8.9	115	9	[56, 61, 64–66, 68, 70, 71, 77]
KO_t+	28.3	6.0	72	3	[58, 74, 75]
WT_t+	25.8	1.6	26	2	[68, 74]
all_t+	30.4	7.9	274	12	[56, 58, 61, 64–66, 68, 70, 71, 74, 75, 77]

generations [47]. The studies included in the database range from years 1997 to 2020, with the majority from 2013 to 2020. Table 1 details collected data segregated into the subpopulations of interest used in the present aggregate analysis. Supplementary Table 1 displays all included studies, their respective publication years, available source information for mouse backcrossing, and detailed treatment descriptions.

Assessed features

Twelve possible categorical or continuous features of each study were curated in the present work as follows: transgenic *APOE* knockout (KO), transgenic *APOE2* knock-in (KI2), transgenic *APOE3* knock-in (KI3), transgenic *APOE4* knock-in (KI4), wild type (WT), treatments meant to improve cognition or improve the AD etiology (t+), treatments meant to decrease cognition or worsen the AD etiology (t-), untreated mice (t0), female gender (g_f), male gender (g_m), mixed or unstated gender (g_mi), mouse age in days (age), and MWM escape latency in seconds (latency). Notably, mouse age was normalized and resampled such that treated and untreated groups had the same mean age within 0.05%. Unless stated otherwise, reported MWM escape latency is the normalized MWM escape latency in seconds.

Varied treatment methods were employed to directly or indirectly change the underlying neurodegenerative etiology and corresponding mouse cognition. To appropriately aggregate treatment effects, treatments were categorized based on their intended effect

on cognition (positive or negative) using descriptions of the Methods in the original published works. Examples of positive treatments include bexarotene administration [56], medical ganglionic eminence transplantation [64], and acetyl-L-carnitine and dex-lipotam dietary supplementation [70]. Examples of negative treatments include repetitive traumatic brain injury [50], intraperitoneal injection of EtOH and ACh [53], and a high-fat diet [49]. Detailed treatment descriptions and treatment category labels for each included data source are listed in Supplementary Table 1.

Statistical analysis

Age-normalized mean MWM escape latency and standard deviation was used to conduct all statistical analysis in the present study as noted under *Assessed Features* section above.

The reported aggregate mean and standard deviation for each group in Table 1 was calculated using the frequency distribution, which weights the average and standard deviation for each group based on the sample size associated with each observation within the group. As such, the contributions of individual original data sources to the reported aggregate mean and standard deviation are a function of the original source's reported mouse sample size for each observed group. Equal weighting of all studies, which would disregard mouse sample size for each reported study when calculating the aggregate group mean and standard deviation, was not possible due to lack of statistical power.

Data distributions were assessed for normality using Shapiro-Wilk test and found to exhibit a sufficiently Gaussian distribution for assessment via ANOVA and *t*-tests. An ANOVA was performed to assess which groups should be further examined using *post-hoc* testing to identify pairwise significant differences. To explore pairwise differences in cognitive performance as measured by mean normalized MWM escape latency, two-tailed *t*-tests were performed at an overall alpha of 0.05. Bonferroni correction was utilized to correct the *p*-value threshold for significance for multiple comparisons, resulting in a *p*-value threshold for significance of *p* < 0.002. All statistical analysis was performed in Microsoft Excel.

Random forest modeling

The MWM escape latency data in the present study was used to produce a supervised random forest machine learning model in Python version 3.8.3 (with Pandas version 1.0.4, NumPy version 1.18.0, and Scikit-learn version 0.23.2 python packages) to predict the importance of twelve binary features: transgenic KO, KI2, KI3, KI4, WT, treatments meant to improve cognition or improve the AD etiology (t+), treatments meant to decrease cognition or worsen the AD etiology (t-), untreated mice (t0), female gender (g_f), male gender (g_m), mixed or unstated gender (g_mi), and mouse age (age). The model was set to classify either superior or inferior normalized MWM escape latency performance using a threshold obtained via data standardization. Curated data was standardized and converted into binomial features (feature present = 1, feature not present = 0). The continuous features of normalized escape latency and age were also standardized using a z-score. For example, normalized escape latency is standardized using the following procedure: [(normalized escape latency for each mouse – mean normalized escape latency for all mice)/ standard deviation of normalized mean escape latency for all mice]. Standardized age was calculated using the same z-score method. A standardized normalized escape latency less than or equal to the mean was binomially classified as “superior” performance, whereas inferior performance was binomially classified as “inferior” performance. Likewise, standardized mouse age less than the mean was binomially classified as “young”, whereas standardized mouse age greater than the mean was classified as “old”.

Random forest models have been widely used in AD literature to assess the high dimensional nature of diagnosis [78]. A random forest is composed of a set of decision trees that each consist of split nodes and leaf nodes. The decision tree is fed a predictor, or target variable, as well as corresponding features used to predict the target variable [79]. MWM escape latency was used as the labelled predictor in this study. Each sample passed to the model is assessed by the split node before being passed to its left or right child, depending on the sample’s features. The decision trees are fed random samples (with replacement) from a subset of the training data or testing data. Beginning with the top split node, or root node, each subsequent node in the tree is trained to continuously split until it achieves stop criteria such as maximal tree length or minimal number of samples. The mean and variance of the target group in the data subset of each decision tree is stored in leaf nodes for future forecasting [79].

While in the training stage, the random forest model can see the labels on the target variable so that it can recognize when a mistake is made by incorrectly predicting the target variable based on the corresponding features for a particular observation. The process of training is how the model learns. During the independent testing stage, the model is exposed to data that was not seen during the training phase, so it does not know the true value of the target variable [79].

RESULTS

This systematic review and meta-analysis with adjunctive machine learning compares cognitive function as empirically measured through escape latency from the MWM test in various WT, *APOE* KI, and *APOE* KO mice. A total of 3,045 mice from 31 peer-reviewed scientific papers [34, 48–77] were included in the analysis. The included peer-reviewed studies consisted of MWM escape latency results at baseline and after 4 or 5 days of training; inclusion of a control and treatment group; WT or *APOE* transgenic mouse type; mouse age in days; and mouse gender (female, male, mixed/unknown). There were 1,430 female mice, 1,181 male mice, and 258 mixed/unknown mice.

Table 1 displays the study citation references for each aggregated data pool utilized for standard statistical analysis to compare the normalized escape latency between groups. Groups are categorized by

transgenic mouse type and treatment with an underscore “_” in between. Mouse genotypes include: KO, KI2, KI3, KI4, WT, and all mice in the study (all). Treatment types include those meant to directly or indirectly improve the AD etiology and thus improve cognition (t+), treatments meant to perturb or assess the worsening of AD etiology and thus diminish cognition (t-), or untreated mice (t0) for which no additional treatments or procedures were performed. For standard statistical analysis, the age was normalized such that there was less than a 0.05% difference between groups (see Methods). To maintain statistical power, gender was not considered in the statistical analysis comparing normalized escape latency between groups, but gender was considered in the supervised machine learning model. Unless stated otherwise, reported MWM escape latency is the normalized MWM escape latency in seconds.

Comparison of cognition in untreated mice

Standard statistical analysis with two-tailed *t*-tests and overall alpha of 0.05 was utilized to compare the impact of mouse genotype on cognition as measured via the normalized MWM escape latency. To prevent skew of data results from over-training in the MWM, only the first 5 days of training trials were included for analysis. The threshold for significance was adjusted using a Bonferroni correction for multiple comparisons (with $p < 0.002$ required for significance), which greatly reduced the likelihood of a false positive.

First, mean normalized MWM escape latencies were examined for all included mouse genotypes without any additional modulatory treatments. Figure 2 illustrates the mean escape latency in seconds for each untreated group (t0) with the error bars denoting the corresponding positive standard deviation. Untreated transgenic *APOE3* KI mice (KI3_t0) was the only group to have a significantly lower normalized mean escape latency compared to WT. Untreated KI2 mean normalized MWM escape latency was not significantly different than WT. However, KI4 and KO had significantly longer MWM escape latencies compared to WT. Moreover, untreated KI4 mice had the highest normalized mean escape latency, meaning they had significantly worse cognitive performance compared to the other untreated groups. The colored * above each bar in Fig. 2 represents a significant pairwise comparison between the group corresponding to the labeled bar and the group corresponding to the color of the asterisk. For example, untreated

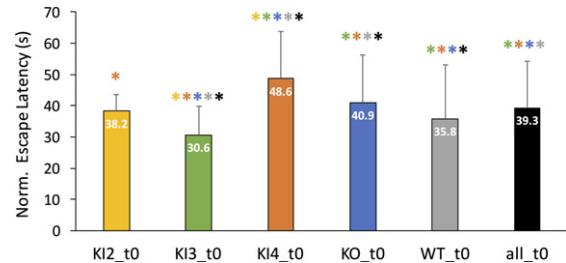


Fig. 2. Comparison of mean normalized Morris Water Maze (MWM) mean escape latency (in seconds) between untreated (t0) wild type and transgenic *APOE* genotypes. MWM escape latency is normalized for age but not gender. Mean normalized escape latency comprises MWM escape latency measurements for baseline through day 4 or 5 of training. The error bar corresponds to the positive standard deviation for the corresponding group. Groups are as follows: transgenic *APOE* knockout (KO), transgenic *APOE2* knock-in (KI2), transgenic *APOE3* knock-in (KI3), transgenic *APOE4* knock-in (KI4), wild type (WT), all mice in the study (all). The color-coded asterisk (*) indicates Bonferroni-corrected pairwise statistical significance between groups ($p < 0.002$).

transgenic *APOE3* KI (KI3_t0) had a significantly different normalized mean escape latency than every other untreated group (KI2, KI4, KO, WT, and all).

Treatment effect on cognition in *APOE* mice

The normalized mean MWM escape latency was analyzed for mice that were treated with *APOE* modulatory treatments. Such treatments could include either direct or indirect modulation of underlying etiology beyond the original *APOE* genotype modifications listed in Table 1. Treatments were separated based on their intended effect. Positive treatments (t+) were meant to lessen the AD etiology and thus improve cognition, whereas negative treatments (t-) were meant to perturb or worsen the AD etiology, and thus, further impair cognition. A full description of treatments for each study is given in Supplementary Table 1. The untreated group (t0) for each mouse genotype is shown in each panel of Fig. 3 for visual comparison. Again, to maintain statistical power, the statistical analysis was not segregated by mouse gender. However, all groups were normalized for age to better isolate the impact of the treatment and mouse genotype.

Figure 3 illustrates the impact of positive treatment (t+), negative treatment (t-), or no treatment (t0) for each mouse genotype. A couple of groups had extremely limited data. Namely, only untreated data was available for KI2 and only one study was available for negatively treated transgenic *APOE3*

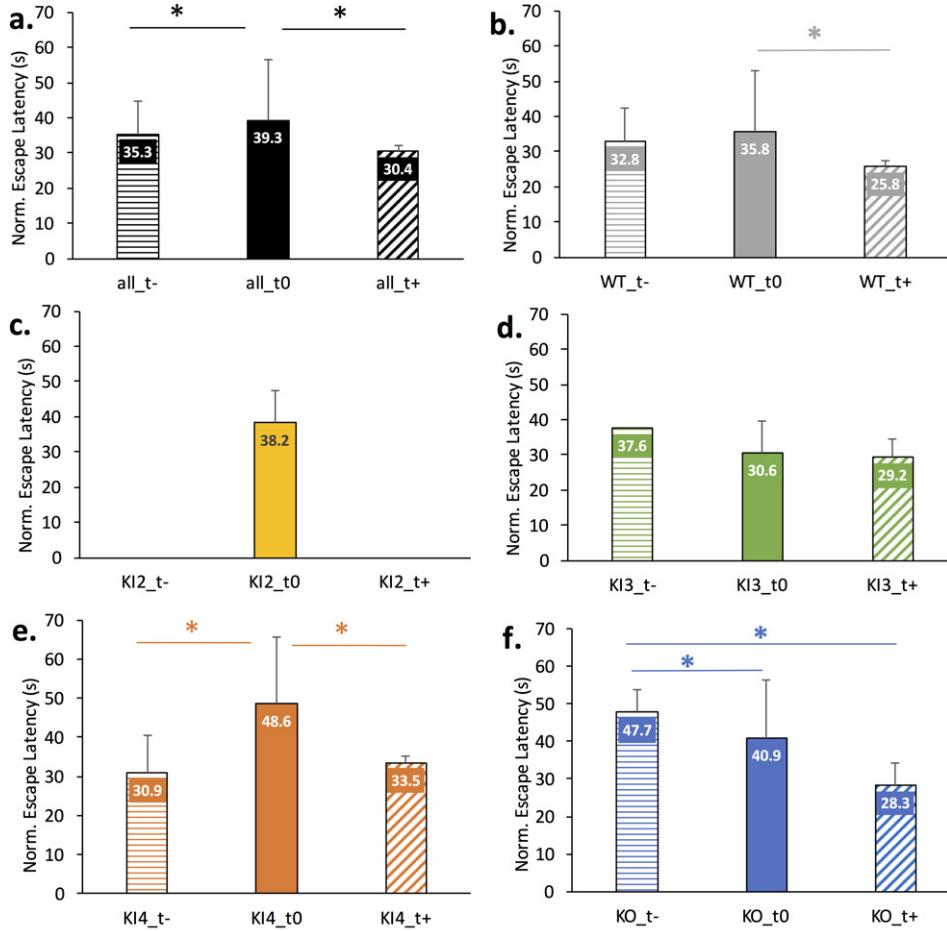


Fig. 3. Comparison of modulatory treatments on normalized Morris Water Maze (MWM) escape latency across wild type (WT) or *APOE* mouse genotypes. Data is normalized for age but not gender. Mean normalized escape latency comprises MWM escape latency measurements for baseline through day 4 or 5 of training. The error bar corresponds to the positive standard deviation for the corresponding group. Positive treatments (t+) were meant to lessen the AD etiology and/or improve cognition; negative treatments (t-) were meant to worsen the AD etiology and/or worsen cognition; untreated mice (t0) of the same genotype are shown for comparison. A full description of the treatments for each study is given in Supplementary Table 1. The asterisk (*) indicates Bonferroni-corrected pairwise statistical significance between groups ($p < 0.002$). a.) All mice genotypes; b.) Wild type mice; c.) Transgenic *APOE2* knock-in (KI2) mice; d.) Transgenic *APOE3* knock-in (KI3) mice; e.) Transgenic *APOE4* knock-in (KI4) mice; f.) Transgenic *APOE* knockout (KO) mice.

KI (KI3_t-). For the remaining groups, the mean normalized escape latency is shown with the error bar visually denoting standard deviation. Pairwise significance is denoted by the * and corresponds to significance less than the Bonferroni-corrected threshold of $p < 0.002$. Figure 3a shows a significant pairwise difference between the normalized escape latencies of all untreated mice (all_t0) and all negatively treated mice (all_t-) as well as between all untreated mice (all_t0) and all positively treated mice (all_t+). Figure 3b illustrates a significant difference in normalized escape latency between untreated WT (WT_t0) and positively treated WT (WT_t+) mice. Figure 3c illustrates the normalized escape

latency of untreated transgenic *APOE2* KI (KI2_t0) given treatment data was absent for this genotype. Figure 3d illustrates no significant difference in normalized escape latency between treated and untreated transgenic *APOE3* KI mice. Figure 3e illustrates significant pairwise differences in normalized escape latency between negatively treated transgenic *APOE4* KI (KI4_t-) and untreated (KI4_t0) as well as between untreated and positively treated *APOE4* KI (KI4_t+). Figure 3f illustrates a significant pairwise difference in normalized escape latency between negatively treated transgenic *APOE* KO (KO_t-) and untreated (KO_t0) and between KO_t- and positively treated (KO_t+).

Evaluation of feature importance with random forest modeling

A supervised machine learning technique, random forest modeling, was performed to assess the impact of twelve specific binary features on the prediction of normalized MWM escape latency, including gender (g_m = male; g_f = female, g_mi = mixed/unknown), which was not analyzed in the pairwise statistical analysis presented above. Data from the curated relational database was randomly partitioned into a training set (80% of data) and an independent testing set (20% of data). A supervised random forest model was performed using a binomial standardized normalized MWM escape latency as the predictor where

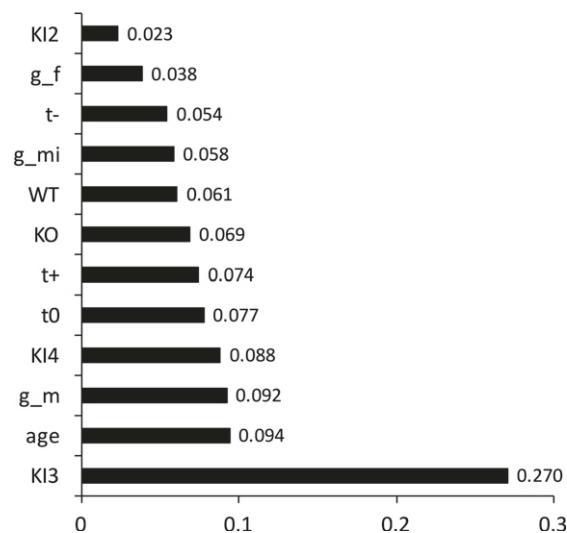


Fig. 4. Importance of binomial features on normalized and standardized Morris Water Maze (MWM) escape latency as predicted by supervised random forest modeling. Data was randomly partitioned into a training set (80%) and an independent test set (20%). Overall model accuracy during independent testing was 82%. Twelve binary features were utilized to predict escape latency; thus, each feature was marked for each mouse as being either present or absent. In descending order, the importance of the following binary features was predicted for classifying “superior” cognition: [KI3, age, g_m, KI4, t0, t+, KO, WT, g_mi, t-, g_f, KI2]=[0.270, 0.094, 0.092, 0.088, 0.077, 0.074, 0.069, 0.061, 0.058, 0.054, 0.038, 0.023]. Superior cognition was defined as a standardized and normalized MWM escape latency less than or equal to the overall sample mean (see Methods). Feature legend is as follows: transgenic *APOE3* knock-in (KI3), male gender (g_m), transgenic *APOE4* knock-in (KI4), untreated (t0), positive treatment meant to enhance cognition (t+), transgenic *APOE* knockout (KO), wild type (WT), mixed or unknown gender (g_mi), negative treatment (t-), female gender (g_f), and transgenic *APOE2* knock-in (KI2). The overall sample size for the supervised random forest model was $n=3,045$ mice. There were 1,430 female mice, 1,181 male mice, and 258 mixed/unknown gender mice. The breakdown of mouse genotype sample sizes is given in Table 1.

latency was classified as either “superior” or “inferior” (see Methods). Model accuracy on independent test data was 82%. Random forest modeling yielded the following feature importance for the prediction of superior cognition as measured via standardized and normalized MWM escape latency: [KI3, age, g_m, KI4, t0, t+, KO, WT, g_mi, t-, g_f, KI2]=[0.270, 0.094, 0.092, 0.088, 0.077, 0.074, 0.069, 0.061, 0.058, 0.054, 0.038, 0.023] (Fig. 4). These results indicate KI3 is most important for predicting superior cognition. Mouse age, male gender, and KI4 comprise the next group of closely ranked features that was second-most important for predicting superior cognition. Untreated (t0), positively treated (t+), and KO comprise the third tier of features for predicting superior cognition. WT, mixed gender (g_mi), and negative treatment (t-) comprise the fourth tier of features for predicting superior cognition. Female gender (g_f) and KI2 comprise the fifth and final tier of features for predicting superior cognition; as such, this final tier is least important for predicting superior cognition as measured by standardized and normalized MWM escape latency.

DISCUSSION

The statistical meta-analysis and machine learning prediction of MWM escape latencies in transgenic mice performed in the current study provided aggregate insight into the *APOE* etiology and therapeutic modulation of *APOE* to improve cognition in AD transgenic mice. Further perspective on the impact of *APOE* or AD modulatory treatment, mouse age, mouse gender, and mouse genotype on cognitive performance as measured by normalized MWM escape latency. The results are explored in the context of prior and common literature examining the role of *APOE* in experimental animal models and in clinical AD in humans.

Impact of *APOE* genotype on cognitive performance

The *APOE* gene is primarily involved in the synthesis of apolipoprotein E, and its expression has been documented in both the central and peripheral nervous system. However, the role of apoE in the body is potentially broader than just lipid metabolism, as its namesake implies. This may include maintaining healthy brain functioning, particularly as apoE acts as a carrier for various cholesterols involved in neuronal

activity and repair. As such, normal apoE is important for clearing of plaques and other wastes from dying cells [80], as well as in the initiation of regenerating neurons [81]. However, pathological apoE, due to alteration of copy number or the presence of an *APOE* mutation, contributes to degeneration resulting in decreased cognition [13, 82].

Among the three human isoforms of *APOE* (E2, E3, and E4), *APOE4* is commonly known to be the greatest genetic risk factor for onset of human AD, while *APOE2* has been demonstrated to decrease the risk of onset in humans [6]. The *APOE4* allele is carried by 3–41% of the global human population [83]. Moreover, there is increasing evidence that the *APOE4* isoform increases risk of AD in humans via pathways involved in loss of function or gain of toxic function. These include well-studied biomarkers of AD, such as tau pathology, tau-mediated neurodegeneration, and an expanding list of A β -dependent and A β -independent pathways that are conditionally affected by different apoE isoforms [13, 84]. In the case of human A β pathways, *APOE4* expression has been shown to catalyze the folding of A β peptides into higher order sheets and other conformations, which contributes to AD pathogenesis by facilitating plaque accumulation in the brain [85].

In the present study's aggregate data analysis, statistical comparison of mean normalized MWM escape latency data in untreated transgenic *APOE* AD mice indicated that untreated knock-in *APOE3* (KI3_t0, Fig. 2) mice out-performed untreated wild type mice (WT_t0, Fig. 2). Like results seen in human AD risk studies on the basis of *APOE* isoform, untreated *APOE2* knock-in (KI2_t0) transgenic AD mice had the closest normalized escape latency to untreated wild type mice (Fig. 2). However, untreated *APOE4* knock-in (KI4_t0, Fig. 2) and *APOE* knock-out (KO_t0) transgenic AD mice performed significantly worse than wild type (Fig. 2). These results illustrate that either loss of apoE function or gain of toxic apoE function can exacerbate AD pathology and corresponding decreased cognition in transgenic mice, which is consistent with human AD observational studies.

The strong association of *APOE4* with MWM escape latency in both the pairwise statistical analysis (Figs. 2 and 3) and the machine learning feature importance (Fig. 4) underscored its known prominence in the AD etiology. The fact that *APOE2* knock-in appeared less protective in the present study compared to observations noted in human studies could be explained by experimental differences in

APOE in transgenic mice. Mice have fundamental differences in *APOE* compared to humans; even *APOE* knock-in mice that utilize human *APOE* isoforms have drastic quantitative differences in apoE content compared to WT mice [86]. While the presence of altered *APOE* genotype or copy number often garners the most attention in human studies, lack of *APOE* can also cause neurodegenerative etiology. As noted in the present study, *APOE* knock-out mice performed worse than WT (Figs. 2 and 3), a finding which is also supported by prior work [58].

The present study examined the role of *APOE* in preclinical data through transgenic mouse models. As such, it is important to note a few marked differences between *APOE* expression and functioning in mice when compared to humans. For instance, mice are shown to naturally express only one isoform of *APOE* [11], whereas humans exhibit three. Furthermore, mice and humans differ in their relative lipoprotein levels, which may alter respective clearance mechanisms. Moreover, mice utilize HDL as a cholesterol shuttler as opposed to the primary use of LDL by humans. In particular, *APOE* KO mice have much higher plasma lipid levels, the effects of which could potentially contribute to synaptic dysfunction and resulting neurodegeneration via independent mechanisms. Overall, the immediate phenotype of deficient *APOE* levels may differ considerably in mice and humans [87].

Impact of age on cognition in the Morris water maze

Age has been widely cited as the single greatest risk factor for AD development in humans [37]. The process of aging is highly intricate and results from the aggregate contribution of various body systems being influenced at the cellular level. Many physiological pathways have been identified in the aging process, and those that most directly affect brain functioning and have been most strongly associated with AD risk include mitochondrial dysfunction, inflammatory reactions in the innate immune system, insufficient glucose metabolism, disruption of lipid homeostasis, worsening A β processing, and decreased regenerative ability [35, 36]. This age-linked neurodegeneration in transgenic AD mice has been demonstrated through downregulation of an NMDAR signaling pathway resulting in a loss of function [34].

Notably, the present aggregate study took extensive care to control for confounds introduced by

differences in mouse age or differences in neurodegeneration at the time of MWM escape latency testing. In the present study, the mean escape latency of trials from baseline through day 4 or 5 MWM training was utilized to prevent skewed escape latencies, and mice ages are normalized and resampled to ensure a minimal standard deviation (see Methods). These procedures reduced experimental confounds over the timeline of MWM testing [72, 88, 89] that would have otherwise clouded study results. Because of the normalization and resampling, assessment of the impact of mouse age on cognition with traditional statistics was not appropriate. However, the random forest machine learning model provided a means to assess the impact of age alongside other features, including *APOE* mouse genotypes, treatment types, and gender. The random forest feature importance (Fig. 4) illustrated that mouse age (in days) was the second-most important individual feature for predicting cognitive performance as measured by MWM escape latency. Specifically, mice with ages younger than the sample mean are more likely to have superior performance in the MWM.

APOE-targeted treatments have a significant impact on cognition independent of age

Multiple potential treatments have been studied to alter levels of *APOE* expression to improve cognitive performance [38]. Most studies focus on “positive” treatments meant to indirectly or directly reduce the *APOE*-related pathology, and correspondingly, improve cognition. However, some studies apply “negative” treatment in an attempt to perturb the neurodegenerative process to allow further etiological assessment; such negative treatments worsen the cognition of the mice compared to their baseline or untreated genotype. For this reason, the present study divided modulatory treatments as positive (t+), negative (t-), or untreated (t0), to better elucidate the external modulation of *APOE* in WT or transgenic mice. A detailed description of modulatory treatments for each study is given in Supplementary Table 1.

It is the positive treatments that are most important for translational development of future human AD treatments. Statistical analysis of modulatory treatments in the present study illustrated that positive treatments were able to significantly improve cognition in wild type (WT_t+, Fig. 3b), *APOE4* knock-in (KI4_t+, Fig. 3e), and *APOE* knock-out (KO_t+, Fig. 3f). Positively treated *APOE3* knock-in

(KI3_t+, Fig. 3d) has a small, albeit insignificant improvement in MWM escape latency. No prior positively treated *APOE2* mouse model study data met criteria for inclusion in the present study; therefore, the impact of positive treatment on *APOE2* knock-in (KI2, Fig. 3c) was unable to be assessed.

Many AD treatment routes attempt to improve cognitive performance through the use of drug agonists for defined receptors in the brain. A commonly used agonist is bexarotene, which has been suggested to provide neuroprotection and synaptic plasticity. Another important connection is the inclusion of ABCA1-mediated lipidation in agonist treatments. ABCA1 typically causes better cognitive performance in both KO and WT. Looking into the subcategory of diet-related treatments, lower caloric intake was seen to have a positive impact on cognitive performance. However, a key issue with many treatments intended to positively modulate *APOE* is that they may be effective in WT mice but not in transgenic *APOE* KI or KO mice.

Negative treatments are less meaningful for human clinical treatment development, but still provide interesting perspective on the underlying AD etiology in transgenic AD mice experiments. In the present study, only negative treatments for *APOE4* knock-in (KI4_t-, Fig. 3e) and *APOE* knock-out (KO_t-, Fig. 3f) had significantly different normalized MWM escape latencies compared to untreated mice of the same genotype. However, the negatively treated *APOE4* knock-in (KI4_t-, Fig. 3e) ironically had improved normalized MWM escape latency compared to untreated *APOE4* knock-in (KI4_t0, Fig. 3e). The aforementioned finding with *APOE4* illustrates its heterogeneity and overall complexity in the AD etiology.

Challenges in translation of pre-clinical APOE treatments to human Alzheimer's disease

Many treatments that show promise in transgenic mice do not translate to human trials. This is likely due to the vast differences of *APOE* in mice, who naturally present one isoform, versus humans, which present three isoforms [90]. Another challenge is that *APOE* pathology is not isolated in human AD, as apoE interacts with multiple factors that contribute to neurodegeneration and corresponding cognitive decline [91]. Nonetheless, parenteral administration of citicoline, an intermediate in the anabolism of structural membrane phospholipids, has been one of the more promising treatments in humans with

APOE4 genetic risk who have early signs of cognitive decline [92]. Moreover, although the study was limited due to a relatively small sample size and uncontrolled design, supplementation of a drink containing antioxidants, omega-3 fatty acids, and resveratrol appeared to elicit positive immune and cognitive effects in *APOE* E3/E3 patients [93].

In a double-blind, placebo-controlled human clinical trial of bexarotene, an agonist of the retinoid X receptor, the overall results were negative. Bexarotene was shown to increase serum A β in *APOE4* noncarriers; as such, it increased associated risk factor for contracting cardiovascular disease without conferring any positive effects on cognition [94]. Similarly, in phase III trials of bapineuzumab, a humanized anti-A β antibody, results were negative in that no significant difference between the treatment and placebo groups was observed [95]. In summary, ongoing research is still required to successfully translate *APOE* modulatory treatments for human AD.

Impact of gender on cognition in AD

In the present aggregate data analysis, traditional pairwise statistical analysis could not separately address the impact of gender on each transgenic AD mouse model's mean normalized cognitive performance in the MWM due to a lack of statistical power. However, the supervised random forest machine learning model was able to examine the importance of gender in context with the other features that predict normalized MWM escape latency, including mouse genotype, mouse age, and *APOE* modulatory treatments. The random forest model calculated the feature importance for 12 binary features used to predict "superior" MWM performance, which was defined as a normalized MWM escape latency faster than the sample mean. The plot of random forest feature importance (Fig. 4) illustrated that the male gender (g_m) is the third-most important individual feature for predicting superior cognitive performance in the MWM. The male gender fell into the second of five tiers that grouped features of similar importance in predicting normalized MWM escape latency. Mixed gender (g_mi) and female gender (g_f) were not prominent features for predicting superior MWM escape latency. In fact, mixed gender ranked eighth (of twelve) as an individual feature and fell into the fourth (of five) tiers of grouped features to predict superior MWM escape latency. Likewise, female gender ranked eleventh (of twelve) as an individual

feature and fell in the fifth or least important tier for predicting superior cognitive performance via MWM escape latency. Therefore, female gender was not a common feature present among mice that performed superiorly in the MWM.

The findings of gender in the present aggregate study do align with what has been seen with human AD patients in the clinic, as females tend to be at greater risk for AD. Regarding the role of gender, Mielke et al. propose that several possibilities could explain the discrepancy in AD incidence between men and women, including risk factors with equal frequency in each gender but are stronger in one group, risk factors with the same effect but different frequency in each gender, risk factors that vary in effect and frequency in each gender, and risk factors limited to one gender [26]. An example of risk factor with the same effect but different frequency is smoking, of which rates are higher in men, and an example of a risk factor that varies in effect and frequency is that women appear to be more sensitive to head trauma despite the fact that men suffer from such injuries at higher rates [26].

Additionally, several studies have indicated that women have a higher risk for AD than men if they are *APOE4* carriers, but the reasoning behind this discrepancy is still disputed [27]. Moreover, prior studies have suggested that decreasing levels of estrogen as women age may lead to higher rates of cognitive decline due to the effects of menopause on mitochondrial pathways associated with AD risk [28]. Interestingly, previous work found that estrogen appears to upregulate the *APOE* gene [29], yet other studies did not arrive at the same conclusion, mainly due to the complications associated with isolating the effects of estrogen in multiple brain pathways [27].

Females have been documented as more susceptible to development of AD across epidemiological studies, as well, which have similarly attributed the phenomenon to declines in estrogen levels [26, 30–33]. Estrogen has a demonstrated involvement in A β pathways associated with neurodegeneration. Mitochondria from young females confers protection against A β toxicity that is eventually lost as females reach old age [31]. Additionally, multiple studies have shown that female transgenic AD mice express higher plaque load in comparison to male mice [30].

In an effort to improve outcomes in women, human estrogen replacement therapy trials have attempted to remediate the hormone-linked components of AD pathogenesis. However, there are mixed results in the literature. The most notable negative outcomes are

increased risk of cancer and cardiovascular disease upon estradiol administration [31]. Unfortunately, there remains a deficiency of findings addressing the efficacy of pharmacotherapies aimed at preventing and treating AD in at-risk women [32].

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SUPPLEMENTARY MATERIAL

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