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**The mutational landscape of *EGFR*-, *MYC*-, and *Kras*- driven genetically-engineered mouse models of lung adenocarcinoma**

David G. McFadden<sup>1,||, #</sup>, Katerina Politi<sup>2,3, #</sup>, Arjun Bhutkar<sup>1</sup>, Frances K. Chen<sup>1</sup>, Xiaoling Song<sup>3</sup>, Mono Pirun<sup>4</sup>, Philip M. Santiago<sup>1</sup>, Caroline Kim<sup>1</sup>, James T. Platt<sup>3</sup>, Emily Lee<sup>5</sup>, Emily Hodges<sup>5</sup>, Adam P. Rosebrock<sup>5,6</sup>, Roderick Bronson<sup>7</sup>, Nicholas D. Socci<sup>\*,4</sup>, Gregory Hannon<sup>5,\*</sup>, Tyler Jacks<sup>1,\*</sup> and Harold Varmus<sup>8,9,\*</sup>

\*These authors contributed equally to this manuscript

#Co-corresponding authors

<sup>1</sup>David H. Koch Institute for Integrative Cancer Research at MIT, Cambridge, MA

<sup>2</sup>Departments of Pathology and Medicine (Section of Medical Oncology) and <sup>3</sup>Yale Cancer Center, Yale University School of Medicine, New Haven, CT.

<sup>4</sup>Bioinformatics Core, Memorial Sloan Kettering Cancer Center, New York, NY.

<sup>5</sup>Cold Spring Harbor Laboratory, Cold Spring Harbor, NY

<sup>6</sup>Current address: Donnelly Centre for Cellular and Biomolecular Research, University of Toronto, ON.

<sup>7</sup>Department of Pathology, Tufts University School of Medicine and Veterinary Medicine, North Grafton, MA.

<sup>8</sup>Cancer Biology and Genetics Program, Sloan-Kettering Institute, New York NY

<sup>9</sup>Current address: Meyer Cancer Center, Weill Cornell Medicine, New York, NY.

<sup>||</sup>Current address: Harold C. Simmons Comprehensive Cancer Center and Departments of Internal Medicine and Biochemistry, UT Southwestern Medical Center, Dallas, TX.

27 **ABSTRACT**

28 Genetically-engineered mouse models (GEMMs) of cancer are increasingly being utilized to  
29 assess putative driver mutations identified by large scale sequencing of human cancer  
30 genomes. In order to accurately interpret experiments that introduce additional mutations, an  
31 understanding of the somatic genetic profile and evolution of GEMM tumors is necessary. Here,  
32 we performed whole exome sequencing of tumors from three GEMMs of lung adenocarcinoma  
33 driven by mutant EGFR, mutant Kras or by overexpression of MYC. Tumors from EGFR- and  
34 Kras- driven models exhibited respectively 0.02 and 0.07 non-synonymous  
35 mutations/megabase, a dramatically lower average mutational frequency than observed in  
36 human lung adenocarcinomas. Tumors from models driven by strong cancer drivers (mutant  
37 EGFR and Kras) harbored few mutations in known cancer genes, whereas tumors driven by  
38 MYC, a weaker initiating oncogene in the murine lung, acquired recurrent clonal oncogenic *Kras*  
39 mutations. In addition, although EGFR- and Kras- driven models both exhibited recurrent whole  
40 chromosome DNA copy number alterations, the specific chromosomes altered by gain or loss  
41 were different in each model. These data demonstrate that GEMM tumors exhibit relatively  
42 simple somatic genotypes compared to human cancers of a similar type, making these  
43 autochthonous model systems useful for additive engineering approaches to assess the  
44 potential of novel mutations on tumorigenesis, cancer progression, and drug sensitivity.

## 45 INTRODUCTION

46 Lung cancer remains the leading cause of cancer death worldwide, estimated to have caused  
47 158,000 deaths in the US in 2015 (seer.cancer.gov). Lung adenocarcinoma is the most  
48 common form of lung cancer, in both smokers and nonsmokers. Tobacco mutagens cause a  
49 high mutation frequency in the somatic genomes of smoking-associated tumors, complicating  
50 identification of the genetic drivers of tumor progression (1, 2). An increasing number of  
51 somatic alterations that can be targeted by existing drugs or by drug candidates have been  
52 identified in lung adenocarcinoma, and several of these agents have demonstrated efficacy in  
53 patients (3).

54 *KRAS* and *EGFR* are the most frequently mutated oncogenes in human lung  
55 adenocarcinoma (1, 2, 4, 5). *KRAS* mutations are associated with a strong history of cigarette  
56 smoking, whereas *EGFR* mutations are the most frequent oncogene alterations in lung cancers  
57 from never-smokers (6). Our groups and others have generated genetically engineered mouse  
58 models of *EGFR*- and *KRAS*- mutant lung adenocarcinoma (7-10). These models recapitulate  
59 key features of the human disease, including histologic architecture and response and  
60 resistance to conventional and targeted therapies (11, 12). Although individual mice develop  
61 multifocal disease, only a subset of the primary *Kras* tumors progress to metastatic disease. A  
62 distinct gene expression signature has been shown to distinguish metastatic from non-  
63 metastatic primary tumors in a *Kras*<sup>G12D</sup>-driven GEMM, suggesting that acquired genetic or  
64 epigenetic alterations underlie metastatic progression (13).

65 Somatic genome evolution in tumors produced in GEMMs remains incompletely  
66 characterized. Several studies have described the spectrum of acquired DNA copy number  
67 alterations in murine models (14-19). Although these studies reached varying conclusions, it  
68 appears that somatic alterations in DNA copy number, especially changes in copy number of  
69 certain whole chromosomes, are a common somatic event during tumor evolution in GEMMs.  
70 We also recently studied a GEMM of small cell lung cancer using exome and whole genome

71 sequencing. In this model, which is initiated by mutation of the p53 and retinoblastoma (Rb)  
72 tumor suppressors, we identified recurrent alterations in the PTEN/PI3K pathway, in addition to  
73 previously identified focal DNA amplifications targeting the *Myc/1* oncogene (20-22). Taken  
74 together, these studies suggest that, similar to human cancers, GEMM tumors can undergo  
75 extensive genome remodeling during tumor progression, and that a subset of these acquired  
76 events contributes to cancer progression.

77 The mutational landscape of carcinogen-induced murine lung adenocarcinomas was  
78 also recently described and compared to that in tumors initiated by expression of an oncogenic  
79 *Kras* allele (23). Not surprisingly, single nucleotide mutations, including *Kras* mutations, were  
80 more frequently observed in carcinogen-induced tumors. In contrast, secondary DNA copy  
81 number alterations were more prevalent in tumors arising in genetically-engineered mice. This  
82 further suggests that the path of somatic alteration and selection during tumor progression  
83 depends on the specific events that initiate tumorigenesis. As previously described, the  
84 carcinogen-treated tumors acquired clonal oncogenic *Kras* mutations. However, it remains  
85 unknown whether murine lung adenocarcinomas initiated by other oncogenic drivers, or those  
86 harboring combined loss of the tumor suppressor p53, acquire similar or distinct patterns of  
87 somatic alteration during tumor evolution and progression. Here, we describe the somatic  
88 evolution of a panel of tumors and tumor-derived cell lines derived from genetically-engineered  
89 mouse models of lung adenocarcinoma initiated by *Kras*, *EGFR*, and *MYC* (10, 24-26).

90

## 91 **RESULTS**

### 92 **Mouse models of lung adenocarcinoma acquire few somatic point mutations**

93 We generated a panel of tumor specimens and tumor cell lines from GEMMs of *EGFR*-, *KRAS*-,  
94 and *MYC*- mutant lung adenocarcinoma (Figure 1). We performed whole exome sequencing on  
95 tumors and cell lines from these models in order to profile the spectrum of genomic alterations  
96 acquired during tumorigenesis and progression (Tables 1, S1, S2).

97           We initially focused on somatic point mutations. Several methods of mutation calling  
98   have been developed but agreement among the various methods is poor (27); at the beginning  
99   of our studies, there was no independent evaluation of which method had optimal sensitivity and  
100   specificity. Evidence that aggregating calls from multiple methods improves performance was  
101   recently described (23, 28). An added challenge was that the methods were developed to call  
102   mutations in human tumors, and many had parameters (e.g. background mutation rate) that  
103   were optimized for human samples. Therefore, we created a controlled dataset in order to  
104   assess the performance of our variant- calling pipeline in a murine background by simulating  
105   mixtures of tumor and normal DNA using exon capture of mixtures of germline DNA from  
106   different inbred mouse strains (Figure S1A).

107           As a first step, we generated exon-capture sequencing libraries with tail DNA from  
108   inbred C57BL/6 and 129S1/SvImJ mice. In order to simulate tumor subclonal heterogeneity and  
109   contamination with infiltrating non-tumor stromal cells, we serially diluted the 129S1/SvImJ  
110   library (mimicking “tumor DNA”) into the C57BL/6 library (mimicking “normal DNA”) (Figure  
111   S1A). The starting libraries and mixtures were sequenced to a median average depth of 132x  
112   (range from 92x-205x). Somatic mutations were identified using both muTect (v1.1.4) and a  
113   somatic mutation caller developed by our group (hereafter referred to as the HaJaVa caller)  
114   based on the GATK UnifiedGenotyper with filtering to call somatic events (as described in detail  
115   in the Methods sections) (29). Using this approach, individual germline polymorphisms can be  
116   traced through the serially diluted libraries, mimicking somatic variant detection in tumors. This  
117   dataset was used to estimate the false positive and false negative rates at decreasing allelic  
118   fraction.

119           At the indicated depth of coverage, muTect was highly sensitive, particularly at low allelic  
120   fractions that might be found in samples with a low fraction of tumor cells or as a result of  
121   subclonal mutational events, but it exhibited a higher false positive rate. In contrast, the  
122   HaJaVa caller exhibited a higher true-positive rate, but also a higher false negative rate at low

123 allelic fraction (Supplemental methods, Figure S1B-D, Table S3). We found that the intersection  
124 of the two callers exhibited a lower false-positive rate than either caller alone, reducing missed  
125 calls by approximately 50%, with a minimal increase in the false negative rate. Therefore, we  
126 used the intersection of the HaJaVa and muTect calling algorithms to identify somatic mutations  
127 and to compare datasets among the EGFR, MYC and Kras models.

128

## 129 **Kras-driven models of lung adenocarcinoma**

130 We generated DNA from a large panel of tumors and tumor-derived cell lines from *Kras*<sup>LSL-</sup>  
131 <sup>G12D</sup>-based mouse models for whole exome sequencing (Table 1). We sequenced DNA from  
132 fifteen tumor cell lines derived from tumor-bearing *Kras*<sup>LSL-G12D</sup>; *Trp53*<sup>fl/fl</sup> mice, following the  
133 lentivirus-based delivery of cre recombinase (Figure 1)(30). In nine cases, we also sequenced  
134 DNA from the parental tumor from which the cell lines were derived. In addition, to determine  
135 whether expression of cre recombinase generated unexpected mutations, we sequenced DNA  
136 from four tumors arising in the *Kras*<sup>LA2-G12D</sup>; *Trp53*<sup>-/-</sup> model, in which spontaneous recombination  
137 at the *Kras* locus, rather than cre-induced recombination, initiated tumorigenesis (31). Finally,  
138 in order to assess the potential impact of p53 loss on mutation frequency, we sequenced DNA  
139 from eight tumors initiated in *Kras*<sup>LSL-G12D</sup>; *Trp53*<sup>WT</sup> animals.

140 We observed a median nonsynonymous mutation frequency of 0.07/Mb (including  
141 missense and nonsense mutations and mutations affecting splicing signals; Figure 2A, range  
142 0.00-0.46) in Kras-driven tumors (cell lines are excluded from this analysis – see below).  
143 Interestingly, we did not observe a statistically different mutation frequencies between *Trp53* null  
144 (0.07 mutations/Mb) and wild-type tumors (0.06 mutations/Mb, Figure 2B, P value=0.60).  
145 Tumors initiated in the *Kras*<sup>LSL-G12D</sup>; *Trp53*<sup>fl/fl</sup> model harbored numbers of mutations (0.09  
146 mutations/Mb) similar to those in the *Kras*<sup>LA2-G12D</sup>; *Trp53*<sup>-/-</sup> model (0.03) mutations/Mb, P  
147 value=0.1, Figure 2C). With the *Kras*<sup>LSL-G12D</sup>; *Trp53*<sup>fl/fl</sup> model, we observed a statistically  
148 significant increase in the mutation frequency in tumor-derived cell lines (0.25 mutations/Mb)

149 compared to primary tumor specimens (0.07 mutations/Mb) (Figure 2D, p-value=0.001). This  
150 may reflect the presence of subclonal mutations present in the genomes of cell line-founder  
151 clones that would be expected to be enriched during the generation of tumor cell lines. We did  
152 not observe a predilection for specific base transitions or transversions in the tumors or cell lines  
153 (Figure S2).

154 We compared these datasets to available sequencing data from human lung  
155 adenocarcinoma (Figure 2E). We observed a significantly fewer nonsynonymous mutations in  
156 GEMM models than in either smoker- or nonsmoker -associated human lung adenocarcinomas  
157 (Kras GEMM, 0.07 mutations/Mb, nonsmoker-associated, 1.97 mutations/Mb, p-value <0.0001,  
158 and smoker associated, 7.76 mutations/Mb).

159 We identified independent recurrent mutations in a number of genes, including *C5ar1*,  
160 *Dnahc5*, *Nyap2*, *Pcdh15*, *Pclo*, *Rngtt*, *Stil*, *Tenm4*, and *Xirp2*. Seven genes (*Csmd1*, *Hmcn1*,  
161 *Kmt2c*, *Pcdh15*, *Pclo*, *Ttn*, *Xirp2*) mutated in the mouse Kras<sup>G12D</sup>-induced cell lines and tumors  
162 were also recurrently mutated in >15% of samples in human lung adenocarcinomas (LUAD), as  
163 reported by The Cancer Genome Atlas (TCGA). Therefore, *Pcdh15*, *Pclo*, *Ttn* and *Xirp2* were  
164 recurrently mutated in both the KP mouse model and human lung adenocarcinoma (4)(Figure 3,  
165 denoted with an asterisk).

166 As one example, *Xirp2* encodes an actin-binding protein implicated in the maintenance  
167 of inner ear hair cell stereocilia and cardiac myocyte remodeling. *Xirp2* was mutated in two  
168 independent primary tumors and one pair of primary tumor-metastasis cell lines, the latter of  
169 which harbored the same *Xirp2* mutation within a highly conserved region of the Xin actin  
170 binding repeats. *Xirp2* has no known role in cancer, yet 21% of human lung adenocarcinomas in  
171 the TCGA study harbored mutations in *XIRP2* (4). Review of RNAseq data from the TCGA  
172 study revealed very low expression of *XIRP2* mRNA in human lung adenocarcinoma,  
173 suggesting that mutations in *XIRP2* may be passenger events, despite their high frequency.  
174 Alternatively, cells with *XIRP2* mutations might have been selected at an early stage of

175 tumorigenesis but they may not be advantageous during outgrowth of the dominant tumor  
176 subclone prior to clinical detection.

177 We also observed recurrent mutation of *Pclo*, which has been shown to be important for  
178 axonal guidance during central nervous system development. *PCLO*, which was mutated in  
179 21% of lung adenocarcinomas in the TCGA study, was also recently identified as a recurrently  
180 mutated gene in liver cancers exhibiting a biliary phenotype (32). Knockdown of *PCLO* RNA in  
181 human liver cancer cells led to an increase in cell migration. We identified two mutations in *Pclo*,  
182 a nonsense mutation (E577X) and E1850K; the latter resides in a conserved region of the  
183 protein with unknown function. Further investigation of the role of *Pclo* alterations, especially in  
184 the context of *Kras* mutations, seems warranted.

185 Manual review of mutations occurring in a single tumor revealed mutations in several  
186 genes encoding regulators of transcription, including several factors involved in chromatin  
187 modification and regulation. Among these are mutations in *Brd4* (H965P, within a proline-rich  
188 domain of unknown function), *Chd7* (a nonsense mutation, R977X), *Chd8* (H2198R), *Mll3*  
189 (T1798S), *Mlxip* (V453G), *Smarca1*(M27R), *Smyd4* (H769R), and *Tet1* (C1784X) (Table S4).  
190 None of these specific mutations has been identified in human cancers. Therefore, it is difficult  
191 to determine if any of these represent driver events. However, mutations in many epigenetic  
192 regulators in human tumors are not clustered into “hot spots,” so it is premature to conclude that  
193 these are passenger mutations in the mouse model.

194

## 195 **EGFR-driven model of lung adenocarcinoma**

196 Tetracycline-inducible expression of the EGFR<sup>L858R</sup> mutant in the lung epithelium of  
197 transgenic mice leads to the formation of lung adenocarcinomas with bronchioalveolar  
198 carcinoma features that are sensitive to treatment with EGFR tyrosine kinase inhibitors like  
199 erlotinib (10). Long-term intermittent dosing of the mice with erlotinib leads to the emergence of  
200 drug-resistant tumors that harbor some of the molecular features of TKI-resistant human

201 tumors, including a secondary mutation in EGFR, EGFR<sup>T790M</sup> (12). To determine the mutational  
202 load of these tumors, compared to tumors initiated by alternate oncogenes, and to seek genetic  
203 differences between untreated and erlotinib-resistant tumors, we performed whole exome  
204 sequencing of DNA from ten TKI-naïve and six erlotinib-resistant EGFR<sup>L858R</sup>-induced mouse  
205 lung adenocarcinomas. We observed a lower mutational burden in EGFR mutant lung  
206 adenocarcinomas compared to Kras-driven tumors (0.02 vs. 0.07 mutations/Mb, P  
207 value=0.002)(Figure 2A). Erlotinib-resistant tumors exhibited no difference in mutation  
208 frequency compared to untreated EGFR mutant tumors (0.02 vs. 0.02, P value=0.49, Figure  
209 2F). The mutational signature present in the EGFR<sup>L858R</sup>-induced lung adenocarcinomas exhibits  
210 a preponderance of C>T transitions, consistent with findings in EGFR mutant human lung  
211 adenocarcinomas and all adenocarcinomas from never-smokers (4) (Figure S2).

212 Recurrent mutations in the tumors were found in *Kras* (n=4, 2 G12V and 2 Q61R), *Mll5*  
213 (n=2) and *Ube3b* (n=2). We previously described an oncogenic *Kras* mutation in an erlotinib-  
214 resistant murine tumor; however, such mutations have not been described in patients (12).  
215 Interestingly, two of the *Kras* mutations observed here (with non reference allele fractions of  
216 0.38-0.55) were found in the ten tumors not treated with TKI's, suggesting that these can arise  
217 during tumor development independent of treatment. A recent report shows that co-expression  
218 of mutant EGFR and KRAS in the same human lung tumor cells can be toxic (33). It is possible  
219 that detection of mutations in both oncogenes in some untreated tumors indicates that  
220 expression of one of the oncogenes has been down-regulated in at least some tumor cells; in  
221 other tumors with both mutations, inhibition of the EGFR kinase activity with a TKI may have  
222 been a permissive feature.

223 *Mll5* (*Kmt2e*), a histone lysine methyltransferase involved in chromatin remodeling, was  
224 found to be mutated in two EGFR-induced tumors. Both of the mutations (in an untreated and  
225 an erlotinib-resistant tumor) were in disordered domains of the protein and their functional  
226 consequences are unknown. *Ube3b*, an E3 Ubiquitin ligase, was mutated in two individual

227 tumors from a single untreated mouse. The same variant was found in both tumors, suggesting  
228 that the two tumors are clonally related. According to TCGA reports, *Mll5* and *Ube3b* are  
229 altered in 6% and 4% of human LUADs respectively; however, there is no indication at this point  
230 of a functional relationship between EGFR mutations and alterations in these genes.

231 Since a major mechanism of resistance to EGFR inhibitors is a secondary mutation in  
232 *EGFR* (*EGFR*<sup>T790M</sup>), we examined the whole exome sequencing data to determine whether we  
233 could detect reads corresponding to human *EGFR* (since the transgene encodes human  
234 *EGFR*). Indeed, we unequivocally detected the *EGFR*<sup>T790M</sup> mutation in one erlotinib-resistant  
235 tumor that we had previously shown to harbor this mutation. However, we cannot exclude  
236 inadequate depth of sequencing of the human EGFR transgene as an explanation of our failure  
237 to identify other cases of secondary T790M mutations of EGFR in these tumors, especially  
238 since the exon capture probes target mouse sequences.

239

#### 240 **MYC model of lung adenocarcinoma**

241 The low mutation rate observed in the *Kras* and *EGFR*-induced lung tumors prompted us to  
242 hypothesize that tumors induced by strong oncogenic lung drivers might have a lower mutation  
243 burden than tumors induced by a less potent lung oncogene. We therefore performed whole  
244 exome sequencing of DNA from five lung adenocarcinomas that arose in a mouse model  
245 initiated by overexpression of wild type human *MYC*, which has been shown to be a less potent  
246 oncogene than mutant *KRAS* or *EGFR* in the murine lung (25). *MYC*-induced tumors also  
247 exhibited a low mutation frequency, comparable to that observed in *Kras*-induced mouse lung  
248 adenocarcinomas (0.14 vs 0.07 mutations/Mb, P value=0.57 Figure 2A). The mutations found in  
249 the *MYC*-induced tumors included oncogenic *Kras* mutations in three of five tumors and an  
250 oncogenic mutation in *Fgfr2* (*Fgfr2* K659M) in one tumor. Mutations at this residue in *FGFR2*  
251 have been shown to activate the intrinsic protein-tyrosine kinase and to cooperate with *MYC* in  
252 tumorigenesis (34, 35). The identification of known cancer driver mutations in 4 of 5 *MYC*-

253 driven tumors is consistent with the suggestion that MYC acts as a less potent tumor initiator  
254 that mutant Kras or EGFR in the murine lung.

255

256 **Acquired whole chromosome copy number changes are common in mouse lung**  
257 **adenocarcinomas**

258 Given the low point mutation rates observed in the mouse tumors in our GEMM models, we  
259 asked whether the tumors harbored alterations in chromosomal or subchromosomal copy  
260 numbers. In order to examine somatic changes in DNA copy number in the GEMM tumors, we  
261 analyzed the datasets from whole exome sequencing using validated computational methods  
262 (36, 37). We first examined the *Trp53* locus, which was anticipated to be deleted in the  
263 Kras/p53<sup>fl/fl</sup> tumors, when Cre recombinase was expressed to initiate tumorigenesis (24). We  
264 detected deletion of exons 2-10 in all tumor cell lines derived from those animals, suggesting  
265 that the method based on sequence data accurately identified small regions of deletion (Figure  
266 S3).

267 When these methods were applied to the complete set of exome sequencing data, we  
268 primarily detected putative whole chromosome gains and losses in the murine lung  
269 adenocarcinomas (Figure 4), consistent with prior studies of cancers arising in GEMM's (16,  
270 17). Manual review of putative focal amplifications and deletions revealed that many were  
271 unlikely to be true events. In particular, one signature observed was a set of breakpoints with  
272 the same start and stop positions and in tumors that were matched to the same normal sample.  
273 These are likely to be artifacts of variable coverage from sequencing of the normal (tail) DNA.  
274 Another finding included regions with nearly balanced amplifications and deletions with  
275 overlapping annotated duplication regions in the UCSC browser. These are likely to be copy  
276 number polymorphisms (Table S5).

277 Although the models we have studied all produce one histological tumor type, lung  
278 adenocarcinoma, the tumors that develop in each GEMM show a distinct pattern of recurrent

279 DNA copy number gain or loss (Figure 4A-C, Table S5). *Kras*-driven tumors and cell lines  
280 harbored recurrent gain of Chr6, consistent with prior studies of *Kras*-induced lung tumors (16,  
281 17, 23, 38). In addition to extra copies of Chr6, we observed recurrent whole chromosome  
282 amplification of Chr2, Chr15, and Chr19 (in  $\geq 20\%$  of samples), and whole chromosome loss of  
283 Chr9 and Chr14 (Table S5). Since Chr6 carries the *Kras* locus, we determined the allelic  
284 fraction with the G12D mutation; this analysis suggested that the chromosome with the  
285 engineered G12D mutant, not the chromosome with the wild type allele, was responsible for the  
286 gain in chromosome number (Figure S4). Similarly, the proto-oncogene *Myc* is on Chr 15, and  
287 gain of copies of Chr 15 is the second most frequent whole chromosome alteration observed in  
288 this model and consistent with previous work suggesting that *Myc* function may be necessary  
289 for tumor maintenance in *Kras*-driven GEMMs (39).

290 Recurrent whole chromosome DNA copy number changes appeared to be less frequent  
291 in the *EGFR*-driven GEMM (Figure 4, Table S6) and a different pattern of changes was  
292 observed. For unexplained reasons, an increased number of copies of Chr12 was the most  
293 recurrent alteration. It is possible that the unmapped *TetO-EGFR* transgene is integrated in  
294 Chr12; gain in copy number might then increase signaling from the *EGFR* oncogene.

295 We did not observe an anti-correlation between the frequency of point mutations and the  
296 fraction of the genome affected by DNA copy number alterations, as previously described (23).  
297 This might in part be due to the overall low mutational frequency observed in these tumors,  
298 which is approximately an order of magnitude lower than observed in carcinogen-induced  
299 models (mean of 185 mutations in urethane-induced and 728 in MNU-induced tumors) (23).  
300 Therefore, it is possible that when higher mutation loads are present there is less selection for  
301 large-scale DNA copy number alterations across the genome in these models.

## 302 DISCUSSION

303 Recent improvements in DNA sequencing technologies have spurred the genomic  
304 characterization of many types of human cancers, including lung adenocarcinomas. These  
305 datasets have revealed much information about the mutational profiles of this cancer, and  
306 identified several novel putative oncogenes and tumor suppressors. However, unraveling the  
307 complexity of these datasets and distinguishing driver and passenger mutations remain  
308 significant challenges, particularly in highly mutated genomes such as smoking-associated lung  
309 adenocarcinoma. In contrast, in this report, we have observed a very low mutation frequency in  
310 EGFR-, Kras- and MYC-driven GEMM tumors, regardless of tumor genotype. The low mutation  
311 frequency in the murine tumors is consistent with prior studies of other GEMMs and suggests  
312 that the number of mutations necessary for the development and progression of invasive lung  
313 adenocarcinomas in mice is small (19, 40).

314 We detected recurrent whole chromosome gains and losses in the EGFR and Kras-  
315 driven models. These observations raise the question of whether copy number alterations are  
316 contributing to tumorigenesis in these models. Considering that we observed recurrent Chr6  
317 amplification, which encodes the engineered *Kras*<sup>LSL-G12D</sup> allele as well as several components  
318 of the MAPK signaling pathway, in Kras-driven models, we speculate that cooperating  
319 oncogenes and or tumor suppressors located in the regions of whole chromosome gain or loss  
320 indeed contribute to tumor progression. However, the identity of the driver events in the  
321 amplified and deleted regions remains to be determined. These observations suggest that  
322 amplification of the signal from the initiating oncogene may be the most important somatic event  
323 in these models to drive tumor progression (38).

324 Previous work also described changes in gene expression during tumor progression,  
325 suggesting that epigenetic alterations contribute to progression in these models (13). The  
326 detection of several mutations in transcriptional regulators and chromatin-remodeling factors in  
327 our models (see Figure 3) is consistent with these findings, although we have not determined

328 directly whether any of the observed mutations accelerate tumor progression or lead to specific  
329 changes in chromatin in these models.

330 We found evidence for clonal selection during the emergence of drug resistance and  
331 during the generation of tumor-derived cell lines. Tumors harvested from mice with *EGFR*-  
332 mutant lung cancer harbored oncogenic lesions known to confer primary or acquired resistance  
333 to TKIs (for example, *Kras* mutations and the *EGFR* T790M mutation, respectively). In addition,  
334 drug-resistant tumors exhibited a higher overall mutation burden compared to untreated  
335 samples reflecting increased complexity of these resistant tumors. Similarly, in the *Kras*-driven  
336 model, tumor cell lines harbored a higher mutation frequency compared to the parental tumors,  
337 suggesting clonal selection during outgrowth of cell lines.

338 Despite the low frequency of observed somatic events in the GEMM tumors, each model  
339 exhibited distinct features. In contrast to the *Kras* and *EGFR* models, which harbored few  
340 mutations known to act as cancer drivers, four out of five *MYC*-induced tumors harbored  
341 oncogenic mutations in *Kras* or in *Fgfr2*. The acquisition of potent driver mutations in these  
342 tumors suggests *MYC* over-expression sensitizes cells to transformation by cooperating with  
343 spontaneous *Kras* or *Fgfr* somatic mutations. In contrast, even in the absence of *Trp53*, *Kras*-  
344 driven tumors did not acquire mutations in known tumor suppressors or oncogenes. This  
345 highlights the potency of this oncogene and strongly suggests that the initiating genetically-  
346 engineered allele is a critical determinant of acquired events in these models.

347 The overall non-synonymous mutation burden in human lung adenocarcinomas is 6.86  
348 mutations/Mb (lung TCGA). This is approximately 50-fold higher than the median mutation  
349 burden observed in any of the mouse lung adenocarcinomas studied here. In part, this is likely  
350 to reflect the lack of carcinogen exposure. However, the mutation rate in never smokers (1.97  
351 mutations/Mb) remains over 10-fold higher than that observed in our lung cancer models (4).  
352 The rapid development of tumors in mice may also contribute to the reduced complexity of the  
353 cancer genome in these models compared to human lung adenocarcinoma. As previously

354 described, the copy number profiles of mouse tumors were generally characterized by large-  
355 scale whole chromosome gains or losses (14, 15, 17, 18). In contrast, human tumors exhibited  
356 both large-scale and focal amplifications and deletions, perhaps also reflecting differences in  
357 carcinogen exposures in tumors in the two species (41).

358 Our findings have important implications for the optimal use and further development of  
359 GEMMs, particularly considering the ease with which these modifications can be generated  
360 using new genome editing methods (42-44). Although we have not sequenced to the extreme  
361 depth necessary to identify mutations in very small subpopulations of tumor cells, the genomic  
362 profile of these models appears to be much less complex than most human cancers. The  
363 genomic complexity of lung cancer is at the heart of drug resistance and appears to be an  
364 important determinant of the response to immune checkpoint inhibitors (45).

365 We previously reported that tumors in the *Kras*<sup>LSL-G12D</sup>; *Trp53*<sup>fl/fl</sup> model exhibit very  
366 modest immune cell infiltrates (46). This is consistent with very few neoantigens generated as a  
367 result of the very low number of somatic mutations that arise during tumor development as we  
368 describe here. However, expression of a strong T-cell antigen in the model induces a potent T-  
369 cell response, which is subsequently suppressed at later stages of tumor progression (46).  
370 Therefore, it is important to consider the low mutation frequency exhibited in tumors in these  
371 GEMMs when designing therapeutic studies or studies of drug resistance studies. At the same  
372 time, the uniformity of the programmed somatic mutations and low acquired mutation frequency  
373 observed in tumors in these models are important experimental strengths, making the models  
374 well suited to reproducible mechanistic studies and genetic screening. Efforts to model  
375 genomic complexity in GEMMs, using mutagens, transposons, or engineered loss of DNA repair  
376 pathways are approaches to further optimize GEMMs for studies of sensitivity and resistance to  
377 therapies and could identify new drivers of progression and metastasis that cooperate with the  
378 initiating engineered mutations.

379 **METHODS:**

380 Mouse models:

381 All animals studies were performed under approved IACUC protocols at MIT and MSKCC.  
382 Tumor induction was performed in *Kras*<sup>LSL-G12D</sup>; *p53*<sup>f/f</sup> and *Kras*<sup>LSL-G12D</sup> mice as previously  
383 described with  $2.5 \times 10^4$  lentivirus particles per animal (30). Tumors were isolated and tumor-  
384 derived cell lines were generated as previously described, for tumor-derived cell lines (13).  
385 Histological analysis was performed on a piece of each tumor in order to assess tumor purity  
386 and histological subtype of lung cancer. Tumors were induced in the *TetO-EGFR*<sup>L858R</sup> and  
387 *TetO-MYC* models by feeding the mice doxycycline impregnated food as previously described  
388 (10).

389

390 Exome Sequencing:

391 DNA was purified from tumor tissue and tumor cell lines using standard methods. Sonication of  
392 2ug genomic DNA was performed using a Diagenode Bioruptor, and size selection was  
393 performed using dual selection using AMPure beads as previously described (47). Exon  
394 capture was performed using Roche SeqCap EZ all-exon mouse kits. Post-capture libraries  
395 were sequenced on an Illumina HiSeq instrument.

396

397 Variant Calling:

398 The pipeline to call somatic mutations was predominately composed of standard programs that  
399 have been used in human tumor analysis with the addition of a custom caller that was optimized  
400 to reduce false positives at the expense of some sensitivity to low allele frequency events. The  
401 final mutation list was the intersection of these two calling algorithms in the hope that artifacts  
402 given rise to false positives in one would be filtered out by the other.

403

404 Raw sequence files were first pre-processed to remove the sequencing adapters. Then clipped  
405 reads were mapped to the standard mm9 genome for the Kras model or to mm9+transGene  
406 (hEGFR or hMYC) hybrid genomes for the EGFR and MYC models. BWA ALN (version 0.5)  
407 was used to make maps. The reads were marked with read groups and sorted, then duplicates  
408 were removed using the PICARD toolkit. These initial bam files were post-processed using the  
409 standard GATK packages; indel realignment was followed by base quality recalibration. The  
410 post-processed bam files were then called using two separate mutation callers: MuTect (v1.1.4)  
411 in high confidence mode (HC) and a custom caller built around the GATK Unified Genotyper,  
412 with a set of filters to improve specificity. An intersection of these calls were post-filtered for  
413 artifacts by removing any events that were in a database of likely germline events (see  
414 supplemental for details). The calls were annotated with Annovar and a list of “functional”  
415 mutations was created that contained missense, non-sense and splice site mutations.

416

#### 417 DNA Copy Number

418 Copy number was determined by first computing normalized log ratios between tumors and  
419 matched normals. This was done taking by taking the bam files from the mutation-calling  
420 pipeline and computing the coverage for each exon target region using bedtools. The raw  
421 coverage numbers for each tumor normal pair were normalized with a robust regression method  
422 that normalized not only for total depth but also for the local GC content around each target  
423 region using the loess function from R. The log (base 2) ratio of T to N was computed from the  
424 normalized coverage and this was then segmented using the Circular Binary Segmentation  
425 method (CBS) of ref 27. A post segmentation normalization was then used to center the diploid  
426 peak at  $\log R = 0$ . To find segments that were either amplified or deleted we used the RAE  
427 algorithm (28), which computes a sample-dependent soft threshold (sigmoid function) for each  
428 Tumor/Normal pair based on the noise of that pair. This gives a value of 0-1 for both  
429 amplifications and deletions, which roughly indicates fractional amount of each alteration. These

430 values were then average over all samples to give the fraction of each region that was amplified  
431 or deleted in a given set of samples.

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441

## 442 **FIGURE LEGENDS**

443 **Table 1: Lung adenocarcinoma GEMMs utilized for the sequencing study.** GEMMs are  
444 grouped by initiating driver event, and columns show the tumor type (studied as a cell line or  
445 primary tumor tissue), genotype of animals, method of cancer gene induction, and number of  
446 samples from each model included in the study. Additional details on the models are found in  
447 the Methods and Figure S1.

448  
449 **Figure 1. Diagrams illustrating the mouse models of mutant *Kras*, mutant *EGFR* and  
450 *MYC*- induced lung adenocarcinoma used in whole exome sequencing.** *Kras* models: A)

451 Mice carrying conditional *Kras*<sup>LSL-G12D</sup> and *p53*<sup>flox/flox</sup> alleles develop lung adenocarcinomas upon  
452 administration of lenti-cre. Cell lines were generated from primary and metastatic lung tumors.  
453 Tumors and cell lines were collected for exome sequencing. B) Mice carrying *Kras*<sup>LA2-G12D</sup>; *p53*<sup>-/-</sup>  
454 form lung adenocarcinomas spontaneously. Primary tumors were collected for exome  
455 sequencing. EGFR model: C) Bi-transgenic *CCSP-rtTA*; *TetO-EGFR*<sup>L858R</sup> mice were treated with  
456 doxycycline at weaning to induce transgene expression (Politi, 2006). Tumors were collected  
457 from untreated tumor-bearing mice or mice were treated with erlotinib as described previously  
458 until the appearance of resistant tumors (Politi, 2010). Untreated and erlotinib-resistant lung  
459 tumors were collected and used for exome sequencing. MYC model: D) Bitransgenic *CCSP-*  
460 *rtTA*; *TetO-MYC* mice were treated with doxycycline at weaning to induce transgene expression.  
461 Overexpression of MYC in type II pneumocyte leads to the development of lung  
462 adenocarcinomas that were collected for whole exome sequencing (Tran, PLoS One 2008).

463  
464 **Figure 2. Low mutational burden in GEMM models of lung cancer.** Dot plots showing the  
465 non-synonymous mutation frequency observed from WES datasets in murine LUADs induced  
466 by oncogenic *Kras* (tumors from either the LA2 or LSL models; tumor-derived cell lines are  
467 excluded, see below), *EGFR* or overexpression of MYC (A), *Trp53* null vs. wild-type tumors (B),

468  $Kras^{LA2-G12D};p53^{-/-}$  vs.  $Kras^{LSL-G12D}; Trp53^{fl/fl}$  tumors (C),  $Kras^{LSL-G12D}; Trp53^{fl/fl}$  tumors vs. tumor-  
469 derived cell lines (D), comparison of  $Kras^{G12D}$ -induced tumors to human lung adenocarcinomas  
470 (Ref. 4) from smoking and nonsmoking patients, shown in log scale (E), untreated vs. drug-  
471 resistant EGFR<sup>L858R</sup>-induced LUADs (F).

472

473 **Figure 3. Mutational landscape of oncogene-induced mouse lung adenocarcinomas.**

474 Schematic diagram of genes mutated in the mouse lung adenocarcinomas and cell lines. (Top)  
475  $Kras$ -, EGFR and MYC-induced tumors are indicated and shaded in red, blue and purple,  
476 respectively. The  $Trp53$  status is indicated (black=null, grey= heterozygous). (Middle) Genes  
477 mutated in 2 or more samples are indicated. (Bottom) Genes mutated in the murine tumors that  
478 are also mutated in >15% of lung adenocarcinomas analyzed in the TCGA. Note that *Csmd1*,  
479 *HmCn1*, and *Kml2c* were not recurrently mutated in murine tumor, whereas asterisk indicated  
480 genes recurrently mutated in human and murine tumors. Erlotinib-resistant tumors are  
481 indicated. Stars are used to highlight tumors harboring an EGFR T790M mutation (by  
482 conventional Sanger sequencing).

483

484 **Figure 4: Distinct patterns of DNA copy number alterations in  $Kras$ - and  $EGFR$ - driven**

485 **GEMMs.** A) Heat map of DNA copy number alterations across all samples. Red = DNA copy  
486 number amplification, blue = DNA copy number loss. Models are grouped by initiating driver  
487 allele. Also shown is p53 status, and whether the sample was a tumor or tumor-derived cell  
488 line. Point mutation frequency is shown in grey-black box with darker shade representing a  
489 higher mutation frequency. B) Recurrent whole chromosome DNA copy number gains in  
490  $Kras^{G12D}$ -driven GEMM tumors and cell lines. C) Recurrent whole chromosome DNA copy  
491 number gains in  $EGFR$ -driven GEMM tumors.

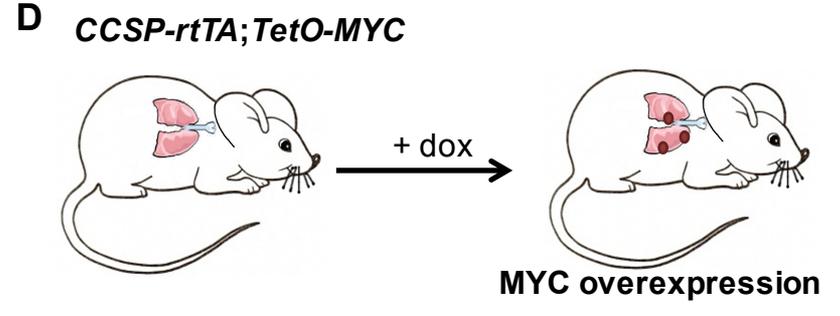
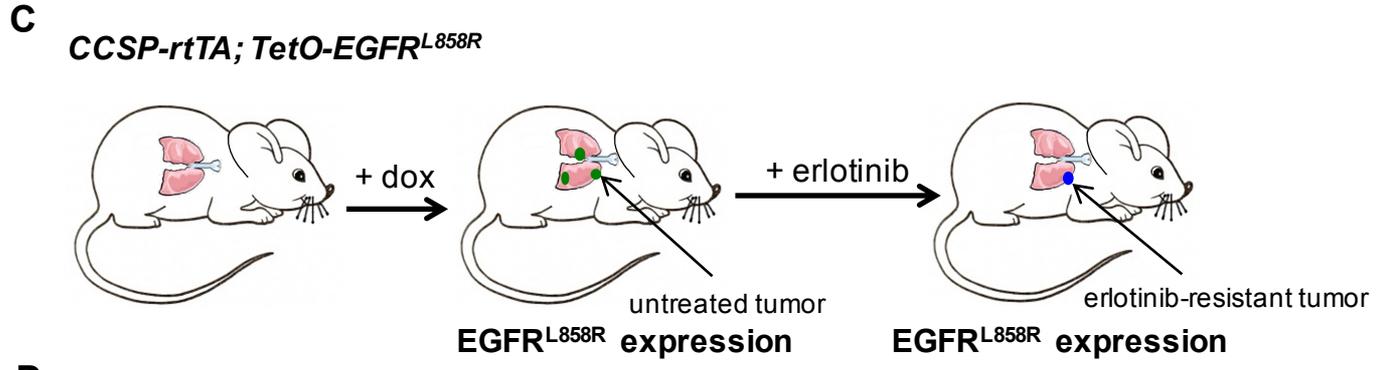
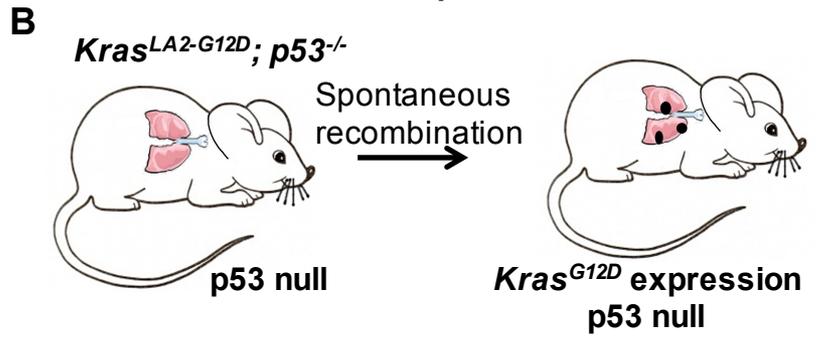
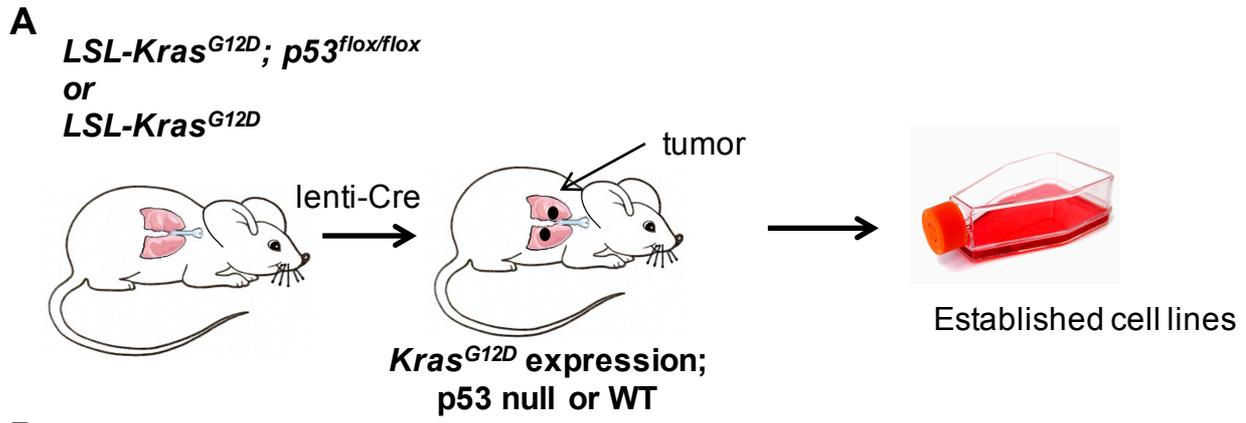
492 **REFERENCES:**

- 493 1. Govindan R, *et al.* (2012) Genomic landscape of non-small cell lung cancer in smokers  
494 and never-smokers. *Cell* 150(6):1121-1134.
- 495 2. Imielinski M, *et al.* (2012) Mapping the hallmarks of lung adenocarcinoma with massively  
496 parallel sequencing. *Cell* 150(6):1107-1120.
- 497 3. Politi K & Herbst RS (2015) Lung cancer in the era of precision medicine. *Clin Cancer*  
498 *Res* 21(10):2213-2220.
- 499 4. Cancer Genome Atlas Research N (2014) Comprehensive molecular profiling of lung  
500 adenocarcinoma. *Nature* 511(7511):543-550.
- 501 5. Ding L, *et al.* (2008) Somatic mutations affect key pathways in lung adenocarcinoma.  
502 *Nature* 455(7216):1069-1075.
- 503 6. Sun S, Schiller JH, & Gazdar AF (2007) Lung cancer in never smokers--a different  
504 disease. *Nat Rev Cancer* 7(10):778-790.
- 505 7. Fisher GH, *et al.* (2001) Induction and apoptotic regression of lung adenocarcinomas by  
506 regulation of a K-Ras transgene in the presence and absence of tumor suppressor  
507 genes. *Genes Dev* 15(24):3249-3262.
- 508 8. Jackson EL, *et al.* (2001) Analysis of lung tumor initiation and progression using  
509 conditional expression of oncogenic K-ras. *Genes Dev* 15(24):3243-3248.
- 510 9. Ji H, *et al.* (2006) The impact of human EGFR kinase domain mutations on lung  
511 tumorigenesis and in vivo sensitivity to EGFR-targeted therapies. *Cancer Cell* 9(6):485-  
512 495.
- 513 10. Politi K, *et al.* (2006) Lung adenocarcinomas induced in mice by mutant EGF receptors  
514 found in human lung cancers respond to a tyrosine kinase inhibitor or to down-regulation  
515 of the receptors. *Genes Dev* 20(11):1496-1510.
- 516 11. Oliver TG, *et al.* (2010) Chronic cisplatin treatment promotes enhanced damage repair  
517 and tumor progression in a mouse model of lung cancer. *Genes Dev* 24(8):837-852.
- 518 12. Politi K, Fan PD, Shen R, Zakowski M, & Varmus H (2010) Erlotinib resistance in mouse  
519 models of epidermal growth factor receptor-induced lung adenocarcinoma. *Dis Model*  
520 *Mech* 3(1-2):111-119.
- 521 13. Winslow MM, *et al.* (2011) Suppression of lung adenocarcinoma progression by Nkx2-1.  
522 *Nature* 473(7345):101-104.
- 523 14. Holstege H, *et al.* (2010) Cross-species comparison of aCGH data from mouse and  
524 human BRCA1- and BRCA2-mutated breast cancers. *BMC Cancer* 10:455.
- 525 15. Maser RS, *et al.* (2007) Chromosomally unstable mouse tumours have genomic  
526 alterations similar to diverse human cancers. *Nature* 447(7147):966-971.
- 527 16. Sweet-Cordero A, *et al.* (2006) Comparison of gene expression and DNA copy number  
528 changes in a murine model of lung cancer. *Genes Chromosomes Cancer* 45(4):338-348.
- 529 17. To MD, *et al.* (2011) Progressive genomic instability in the FVB/Kras(LA2) mouse model  
530 of lung cancer. *Mol Cancer Res* 9(10):1339-1345.
- 531 18. Varela I, *et al.* (2010) Somatic structural rearrangements in genetically engineered  
532 mouse mammary tumors. *Genome Biol* 11(10):R100.

- 533 19. Nassar D, Latil M, Boeckx B, Lambrechts D, & Blanpain C (2015) Genomic landscape of  
534 carcinogen-induced and genetically induced mouse skin squamous cell carcinoma.  
535 *Nature medicine*.
- 536 20. Calbo J, Meuwissen R, van Montfort E, van Tellingen O, & Berns A (2005) Genotype-  
537 phenotype relationships in a mouse model for human small-cell lung cancer. *Cold Spring*  
538 *Harb Symp Quant Biol* 70:225-232.
- 539 21. Dooley AL, *et al.* (2011) Nuclear factor I/B is an oncogene in small cell lung cancer.  
540 *Genes Dev* 25(14):1470-1475.
- 541 22. McFadden DG, *et al.* (2014) Genetic and clonal dissection of murine small cell lung  
542 carcinoma progression by genome sequencing. *Cell* 156(6):1298-1311.
- 543 23. Westcott PM, *et al.* (2015) The mutational landscapes of genetic and chemical models of  
544 Kras-driven lung cancer. *Nature* 517(7535):489-492.
- 545 24. Jackson EL, *et al.* (2005) The differential effects of mutant p53 alleles on advanced  
546 murine lung cancer. *Cancer Res* 65(22):10280-10288.
- 547 25. Tran PT, *et al.* (2008) Combined Inactivation of MYC and K-Ras oncogenes reverses  
548 tumorigenesis in lung adenocarcinomas and lymphomas. *PLoS One* 3(5):e2125.
- 549 26. Podsypanina K, Politi K, Beverly LJ, & Varmus HE (2008) Oncogene cooperation in  
550 tumor maintenance and tumor recurrence in mouse mammary tumors induced by Myc  
551 and mutant Kras. *Proc Natl Acad Sci U S A* 105(13):5242-5247.
- 552 27. Kim SY & Speed TP (2013) Comparing somatic mutation-callers: beyond Venn  
553 diagrams. *BMC Bioinformatics* 14:189.
- 554 28. Ewing AD, *et al.* (2015) Combining tumor genome simulation with crowdsourcing to  
555 benchmark somatic single-nucleotide-variant detection. *Nat Methods* 12(7):623-630.
- 556 29. Cibulskis K, *et al.* (2013) Sensitive detection of somatic point mutations in impure and  
557 heterogeneous cancer samples. *Nat Biotechnol* 31(3):213-219.
- 558 30. DuPage M, Dooley AL, & Jacks T (2009) Conditional mouse lung cancer models using  
559 adenoviral or lentiviral delivery of Cre recombinase. *Nat Protoc* 4(7):1064-1072.
- 560 31. Johnson L, *et al.* (2001) Somatic activation of the K-ras oncogene causes early onset  
561 lung cancer in mice. *Nature* 410(6832):1111-1116.
- 562 32. Fujimoto A, *et al.* (2015) Whole-genome mutational landscape of liver cancers displaying  
563 biliary phenotype reveals hepatitis impact and molecular diversity. *Nat Commun* 6:6120.
- 564 33. Unni AM, Lockwood WW, Zejnullahu K, Lee-Lin SQ, & Varmus H (2015) Evidence that  
565 synthetic lethality underlies the mutual exclusivity of oncogenic KRAS and EGFR  
566 mutations in lung adenocarcinoma. *Elife* 4.
- 567 34. Ota S, Zhou ZQ, Link JM, & Hurlin PJ (2009) The role of senescence and pro-survival  
568 signaling in controlling the oncogenic activity of FGFR2 mutants associated with cancer  
569 and birth defects. *Hum Mol Genet* 18(14):2609-2621.
- 570 35. Tchaicha JH, *et al.* (2014) Kinase domain activation of FGFR2 yields high-grade lung  
571 adenocarcinoma sensitive to a Pan-FGFR inhibitor in a mouse model of NSCLC. *Cancer*  
572 *Res* 74(17):4676-4684.
- 573 36. Olshen AB, Venkatraman ES, Lucito R, & Wigler M (2004) Circular binary segmentation  
574 for the analysis of array-based DNA copy number data. *Biostatistics* 5(4):557-572.

- 575 37. Taylor BS, *et al.* (2008) Functional copy-number alterations in cancer. *PLoS One*  
576 3(9):e3179.
- 577 38. Feldser DM, *et al.* (2010) Stage-specific sensitivity to p53 restoration during lung cancer  
578 progression. *Nature* 468(7323):572-575.
- 579 39. Soucek L, *et al.* (2013) Inhibition of Myc family proteins eradicates KRas-driven lung  
580 cancer in mice. *Genes Dev* 27(5):504-513.
- 581 40. McCreery MQ, *et al.* (2015) Evolution of metastasis revealed by mutational landscapes  
582 of chemically induced skin cancers. *Nature medicine* 21(12):1514-1520.
- 583 41. Weir BA, *et al.* (2007) Characterizing the cancer genome in lung adenocarcinoma.  
584 *Nature* 450(7171):893-898.
- 585 42. Maddalo D, *et al.* (2014) In vivo engineering of oncogenic chromosomal rearrangements  
586 with the CRISPR/Cas9 system. *Nature* 516(7531):423-427.
- 587 43. Sanchez-Rivera FJ & Jacks T (2015) Applications of the CRISPR-Cas9 system in  
588 cancer biology. *Nat Rev Cancer*.
- 589 44. Sanchez-Rivera FJ, *et al.* (2014) Rapid modelling of cooperating genetic events in  
590 cancer through somatic genome editing. *Nature* 516(7531):428-431.
- 591 45. Rizvi NA, *et al.* (2015) Cancer immunology. Mutational landscape determines sensitivity  
592 to PD-1 blockade in non-small cell lung cancer. *Science* 348(6230):124-128.
- 593 46. DuPage M, *et al.* (2011) Endogenous T cell responses to antigens expressed in lung  
594 adenocarcinomas delay malignant tumor progression. *Cancer Cell* 19(1):72-85.
- 595 47. Rodrigue S, *et al.* (2010) Unlocking short read sequencing for metagenomics. *PLoS One*  
596 5(7):e11840.
- 597

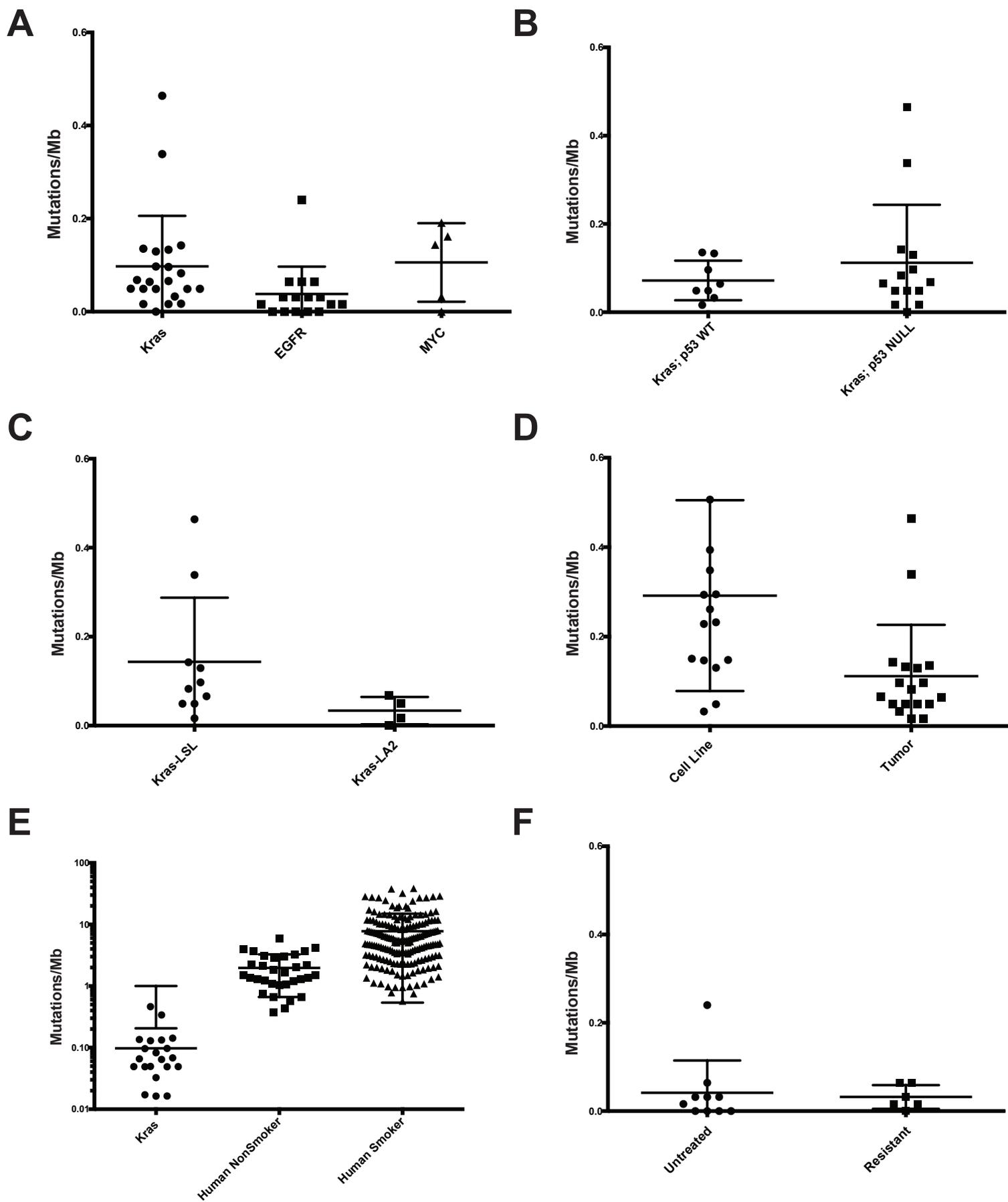
Figure 1, McFadden, Politi et al



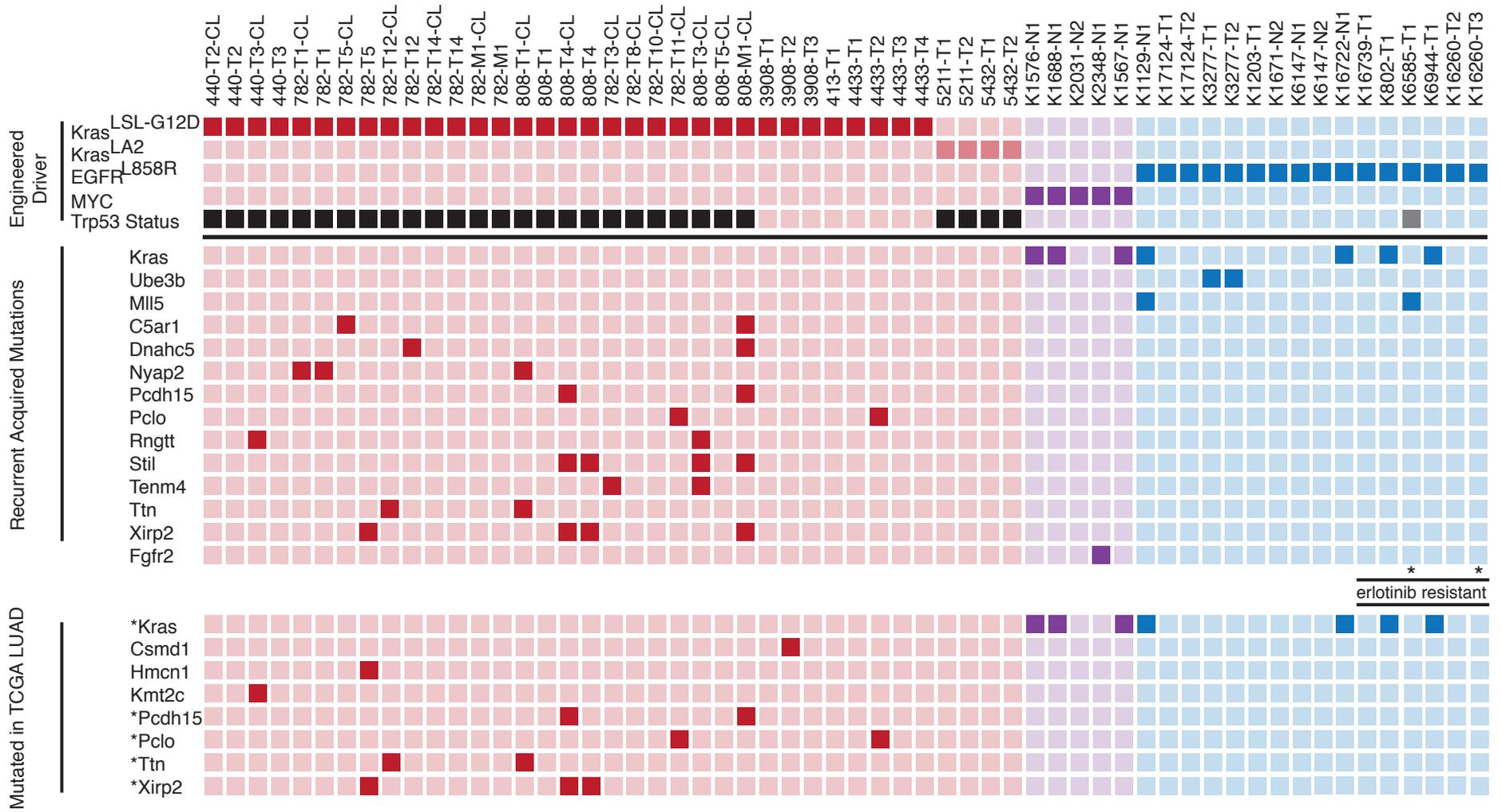
**Table 1, McFadden, Politi et al**

Initiating Driver	Cell line or Tumor	Genotype	Tumor Induction	Treatment	Number Studied	Reference
Kras-G12D	Cell line	Kras-LSL-G12D; Trp53 FL/FL	Lenti-Cre	n/a	15	Jackson et al, 2005
Kras-G12D	Tumor	Kras-LSL-G12D; Trp53 FL/FL	Lenti-Cre	n/a	9	Jackson et al, 2005
Kras-G12D	Tumor	Kras-LSL-G12D	Lenti-Cre	n/a	8	Jackson et al, 2001
Kras-G12D	Tumor	Kras-LA2; Trp53 -/-	spontaneous recombination	n/a	4	Johnson et al, 2001
EGFR-L858R	Tumor	CCSP-rtTA; tetO::EGFR-L858R	doxycycline	none	10	Politi et al, 2006
EGFR-L858R	Tumor	CCSP-rtTA; tetO::EGFR-L858R	doxycycline	erlotinib-resistant	6	Politi et al, 2010
MYC	Tumor	CCSP-rtTA; tetO::MYC	doxycycline	n/a	5	Tran et al, 2008

# Figure 2, McFadden, Politi et al

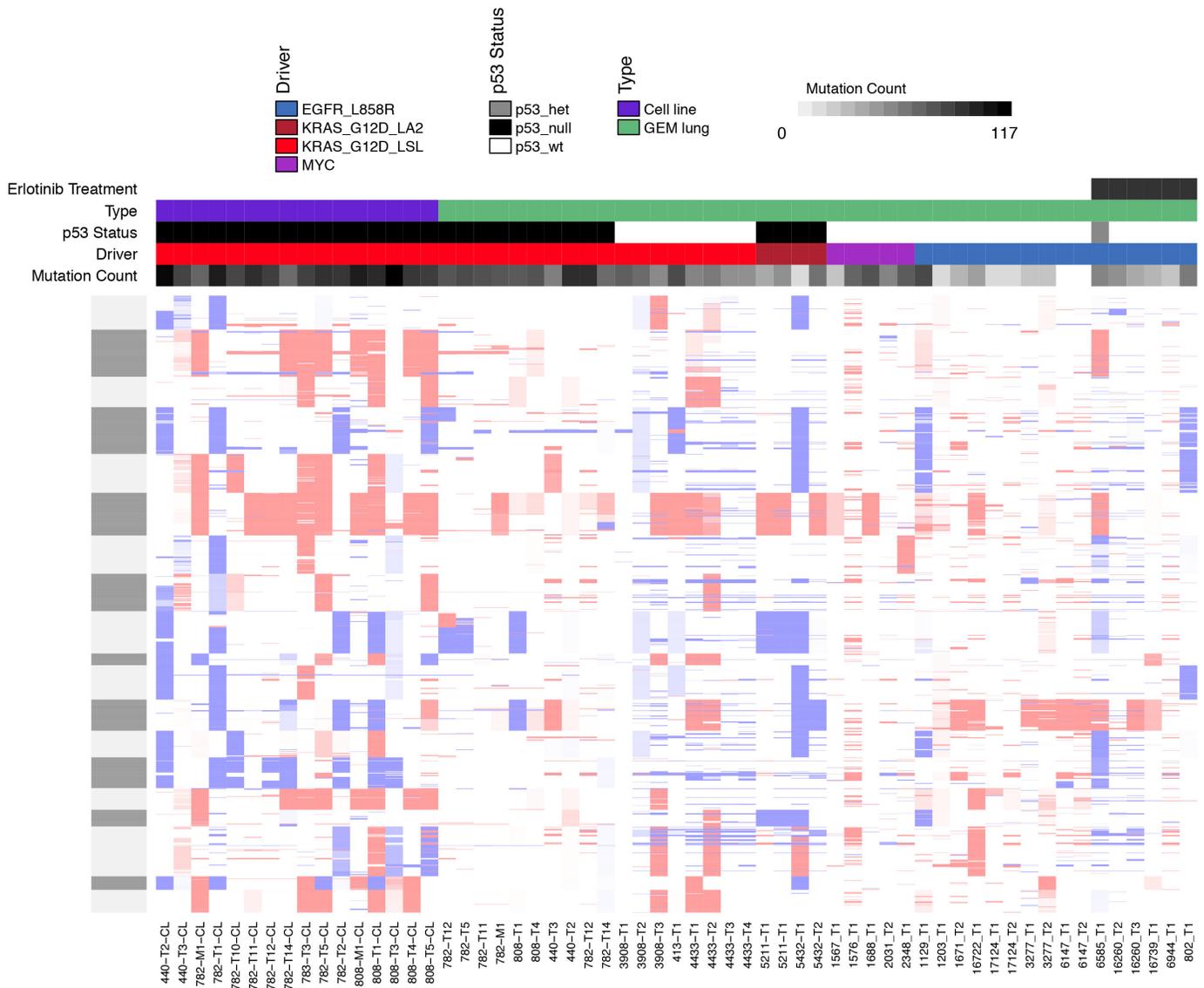


# Figure 3, McFadden, Politi et al

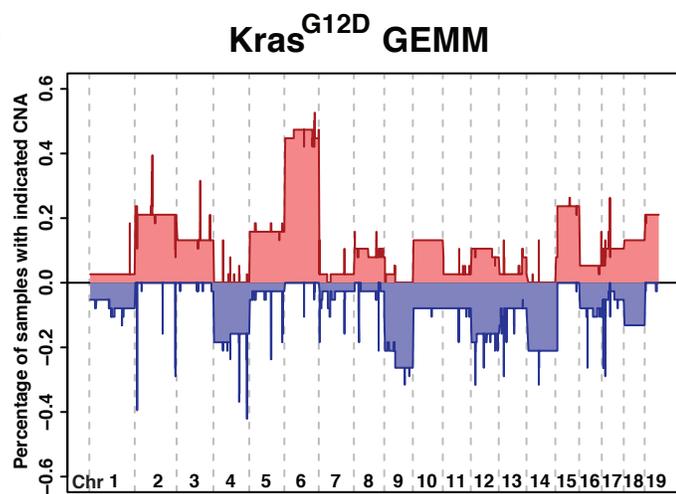


## Figure 4, McFadden, Politi et al

### A



### B



### C

