```
In [1]:
 import pandas as pd
 import seaborn as sns
 import matplotlib.pyplot as plt
In [2]:
 user=pd.read_csv('/home/kesci/input/89031322/tianchi_fresh_comp_train_user.csv'
In [3]:
 user.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15463110 entries, 0 to 15463109
Data columns (total 6 columns):
                int64
user_id
item_id
                 int64
behavior_type
                int64
user_geohash
                object
item_category
                 int64
time
                object
dtypes: int64(4), object(2)
memory usage: 707.8+ MB
In [4]:
 # 快速查看统计信息
 user.describe()
```

	user_id	item_id	behavior_type	item_category
count	1.546311e+07	1.546311e+07	1.546311e+07	1.546311e+07
mean	7.015207e+07	2.023169e+08	1.153780e+00	6.827181e+03
std	4.572019e+07	1.167524e+08	5.440445e-01	3.810410e+03
min	4.920000e+02	3.700000e+01	1.000000e+00	2.000000e+00
25%	3.021406e+07	1.014015e+08	1.000000e+00	3.687000e+03
50%	5.638708e+07	2.022669e+08	1.000000e+00	6.054000e+03
75%	1.166482e+08	3.035247e+08	1.000000e+00	1.025800e+04
max	1.424430e+08	4.045625e+08	4.000000e+00	1.408000e+04

1) 删除重复值

```
In [6]:
    user.drop_duplicates(keep='last',inplace=True)
```

2) 将time转换为datetime格式

```
In [7]:
    user['time']=pd.to_datetime(user['time'])
```

3) 提取出日期和时间

```
In [8]:
    user['dates'] = user.time.dt.date
    user['month'] = user.dates.values.astype('datetime64[M]')
    user['hours'] = user.time.dt.hour
```

4) 转换数据类型

```
In [9]:
    user['behavior_type']=user['behavior_type'].apply(str)
In [10]:
    user['user_id']=user['user_id'].apply(str)
In [11]:
    user['item_id']=user['item_id'].apply(str)
```

1) 统计每日PV和UV数据

```
In [12]:
    pv_day=user[user.behavior_type=="1"].groupby("dates")["behavior_type"].count()
In [13]:
    uv_day=user[user.behavior_type=="1"].drop_duplicates(["user_id","dates"]).group
```

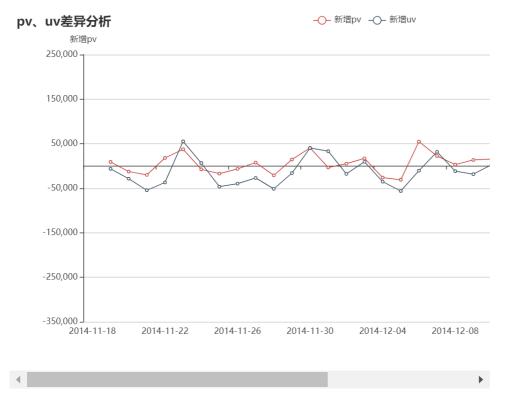
- 2) 分析每天的pv与uv的趋势
- 3) pv、uv差异分析 (by day)

```
In [15]:
    pv_uv = pd.concat([pv_day, uv_day], join='outer', axis=1)
    pv_uv.columns = ['pv_day', 'uv_day']

new_day=pv_uv.diff()
new_day.columns=['new_pv','new_uv']
new_day
```

	new_pv	new_uv
dates		
2014-11-18	NaN	NaN
2014-11-19	9121.0	61.0
2014-11-20	-12504.0	-71.0
2014-11-21	-20029.0	-228.0
2014-11-22	18068.0	-122.0
2014-11-23	37567.0	432.0
2014-11-24	-7484.0	140.0
2014-11-25	-17090.0	-178.0
2014-11-26	-6583.0	-138.0
2014-11-27	7738.0	-61.0
2014-11-28	-21030.0	-208.0
2014-11-29	14585.0	4.0
2014-11-30	40563.0	342.0
2014-12-01	-3456.0	300.0
2014-12-02	5281.0	-7.0
2014-12-03	17074.0	157.0
2014-12-04	-25985.0	-111.0
2014-12-05	-30945.0	-237.0
2014-12-06	54759.0	37.0
2014-12-07	22581.0	291.0
2014-12-08	3030.0	30.0
2014-12-09	13630.0	-10.0
2014-12-10	15549.0	117.0
2014-12-11	101816.0	694.0
2014-12-12	220781.0	1542.0
2014-12-13	-338647.0	-1988.0
2014-12-14	6078.0	-130.0
2014-12-15	-11345.0	238.0
2014-12-16	-12248.0	-166.0
2014-12-17	-8157.0	-218.0
2014-12-18	-9338.0	-77.0

```
In [16]:
 attr = new_day.index
 v = new_day.new_uv
 w = new_day.new_pv
 li=(
     Line(init_opts=opts.InitOpts(width="1000px",height="500px"))
     .add_xaxis(xaxis_data=attr)
     .add_yaxis(
         "新增pv",
         label_opts=opts.LabelOpts(is_show=False)
     )
     .extend_axis(
         yaxis=opts.AxisOpts(
             name="新增uv",
             type_="value",
             min_=-2000,
             max =1600,
             interval=400,
             axislabel_opts=opts.LabelOpts(formatter="{value}"),
      .set_global_opts(
         tooltip_opts=opts.TooltipOpts(
             is_show=True, trigger="axis", axis_pointer_type="cross"
         ),
         xaxis_opts=opts.AxisOpts(
             type_="category",
             axispointer_opts=opts.AxisPointerOpts(is_show=True, type_="shadow")
         ),
         yaxis_opts=opts.AxisOpts(
             name="新增pv",
             type_="value",
             min_=-350000,
             max_=250000,
             interval=100000,
             axislabel_opts=opts.LabelOpts(formatter="{value}"),
             axistick_opts=opts.AxisTickOpts(is_show=True),
             splitline_opts=opts.SplitLineOpts(is_show=True),
          title_opts=opts.TitleOpts(title="pv、uv差异分析"),
 )
 il=(
     Line()
     .add_xaxis(xaxis_data=attr)
     .add_yaxis("新增uv",v,yaxis_index='1',label_opts=opts.LabelOpts(is_show=Fals
 )
 c=li.overlap(il)
 c.render_notebook()
```

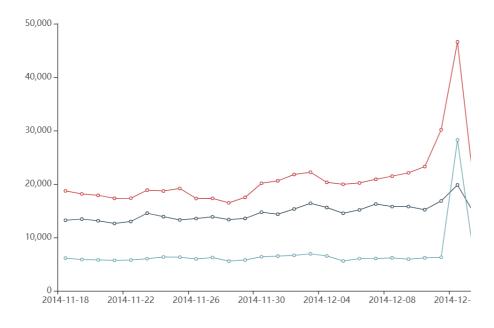


4) 不同时期用户行为分析

```
In [17]:
 shopping_cart= user[user.behavior_type == '3'].groupby('dates')['behavior_type'
 collect=user[user.behavior_type=='2'].groupby('dates')['behavior_type'].count()
 buy=user[user.behavior_type=='4'].groupby('dates')['behavior_type'].count()
 attr_a=list(shopping_cart.index)
 v_1=shopping_cart.values.tolist()
 v_2=collect.values.tolist()
 v_3=buy.values.tolist()
 b=(
     Line()
     .add_xaxis(xaxis_data=attr_a)
     .add_yaxis(
         "加购人数",
         v_1,
         label_opts=opts.LabelOpts(is_show=False)
     .add_yaxis(
         "收藏人数",
         v_2,
         label_opts=opts.LabelOpts(is_show=False)
     .add_yaxis(
         "购买人数",
         v_3,
         label_opts=opts.LabelOpts(is_show=False)
     .set_global_opts(title_opts=opts.TitleOpts(title="不同时期用户行为数据"))
 b.render_notebook()
```

不同时期用户行为数据

-─ 加购人数 -─ 收藏人数 -─ 购买人数



1) 把dates列转换为datetime类型

```
In [18]:
    user['dates']=pd.to_datetime(user['dates'])
```

2) 选取活动数据子集和日常数据子集

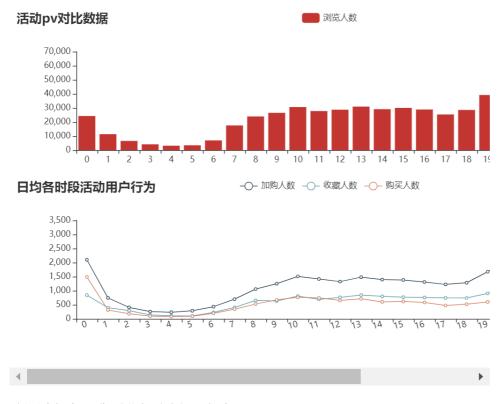
```
In [19]:
    active=user[user["dates"].isin(["2014/12/11","2014/12/12","2014/12/13"])]
```

In [20]:

daily=user[~user["dates"].isin(["2014/12/11","2014/12/12","2014/12/13"])]

3) 活动期间不同时段的用户行为分析

```
In [21]:
 from pvecharts.charts import Bar
 # 活动数据
 cart_h= active[active.behavior_type == '3'].groupby('hours')['behavior_type'].c
 collect_h=active[active.behavior_type=='2'].groupby('hours')['behavior_type'].c
 buy_h=active[active.behavior_type=='4'].groupby('hours')['behavior_type'].count
 uv_h=active[active.behavior_type== '1'].groupby('hours')['user_id'].count()
 attr_h=list(cart_h.index)
 h1=np.around(cart_h.values/3,decimals=0).tolist()
 h2=np.around(collect_h.values/3,decimals=0).tolist()
 h3=np.around(buy_h.values/3,decimals=0).tolist()
 h4=np.around(uv_h.values/3,decimals=0).tolist()
 h= (
     Line(init_opts=opts.InitOpts(width="1000px",height="500px"))
     .add_xaxis(xaxis_data=attr_h)
     .add vaxis(
         "加购人数",
         label_opts=opts.LabelOpts(is_show=False)
     .add_yaxis(
         "收藏人数",
         label_opts=opts.LabelOpts(is_show=False)
     .add_yaxis(
         "购买人数",
         h3.
         label_opts=opts.LabelOpts(is_show=False)
     .set_global_opts(
         xaxis_opts=opts.AxisOpts(axislabel_opts=opts.LabelOpts(rotate=15)),
         title_opts=opts.TitleOpts(title="日均各时段活动用户行为",pos_top="48%"),
         legend_opts=opts.LegendOpts(pos_top="48%"),
 )
 bar=(
     Bar()
     .add_xaxis(xaxis_data=attr_h)
     .add_yaxis(
     "浏览人数",
         h4.
         label_opts=opts.LabelOpts(is_show=False)
     )
     .set_global_opts(
         title_opts=opts.TitleOpts(title="活动pv对比数据"),
 )
 ggrid = (
     .add(bar, grid_opts=opts.GridOpts(pos_bottom="60%"))
     .add(h, grid_opts=opts.GridOpts(pos_top="60%"))
 )
 ggrid.render_notebook()
```



大促集中在0点,因此用户的购买高峰也出现在0点。 高峰集中在21点到22点之间,因此可以在20点前,改好促销页面,吸引顾客参加0点的活动。

4) 日常期间不同时段的用户行为分析

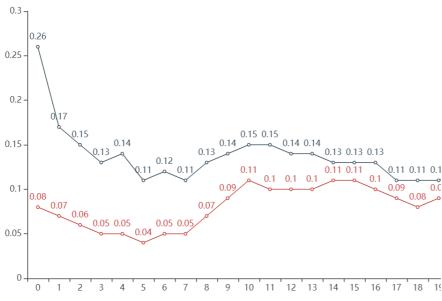
```
In [22]:
 # 日常数据
 cart_d= daily[daily.behavior_type == '3'].groupby('hours')['behavior_type'].cou
 collect_d=daily[daily.behavior_type=='2'].groupby('hours')['behavior_type'].cou
 buy_d=daily[daily.behavior_type=='4'].groupby('hours')['behavior_type'].count()
 uv_d=daily[daily.behavior_type== '1'].groupby('hours')['user_id'].count()
 attr_d=list(cart_d.index)
 d1=np.around(cart_d.values/28,decimals=0).tolist()
 d2=np.around(collect_d.values/28,decimals=0).tolist()
 d3=np.around(buy_d.values/28,decimals=0).tolist()
 d4=np.around(uv_d.values/3,decimals=0).tolist()
 d = (
     Line(init_opts=opts.InitOpts(width="1000px",height="500px"))
     .add_xaxis(xaxis_data=attr_d)
     .add_yaxis(
         "加购人数",
         d1,
         label_opts=opts.LabelOpts(is_show=False)
     .add_yaxis(
         "收藏人数",
         label_opts=opts.LabelOpts(is_show=False)
     .add_yaxis(
         "购买人数",
         d3.
         label_opts=opts.LabelOpts(is_show=False)
     .set_global_opts(
         xaxis_opts=opts.AxisOpts(axislabel_opts=opts.LabelOpts(rotate=15)),
         title_opts=opts.TitleOpts(title="日均各时段活动用户行为",pos_top="48%"),
         legend_opts=opts.LegendOpts(pos_top="48%"),
 )
 y=(
     Bar()
     .add_xaxis(xaxis_data=attr_d)
     .add_yaxis(
     "浏览人数",
         d4,
         label_opts=opts.LabelOpts(is_show=False)
     )
     .set_global_opts(
         title_opts=opts.TitleOpts(title="日常pv对比数据"),
 )
 ggrid = (
     .add(y, grid_opts=opts.GridOpts(pos_bottom="60%"))
     .add(d, grid_opts=opts.GridOpts(pos_top="60%"))
 ggrid.render_notebook()
```



日常期间,pv、加购、收藏的高峰出现在晚上21点到22点之间;浏览高峰也是集中在晚上21点到22点之间。可以集中在这个时段进行促销活动。

2023/5/22 22:45

```
Notebook Viewer
In [23]:
 # 活动时购买率
 hour_buy_user_num = active[active.behavior_type == '4'].drop_duplicates(['user_
 hour_active_user_num = active.drop_duplicates(['user_id','dates', 'hours']).grc
 hour_buy_rate = hour_buy_user_num / hour_active_user_num
 attr_o = list(hour_buy_user_num.index)
 vo_2 =np.around(hour_buy_rate.values,decimals=2)
 # 日常时购买率
 hour_buy_daily_num = daily[daily.behavior_type == '4'].drop_duplicates(['user_i
 hour_active_daily_num = daily.drop_duplicates(['user_id','dates', 'hours']).gro
 daily_buy_rate = hour_buy_daily_num / hour_active_daily_num
 vi_2 =np.around(daily_buy_rate.values,decimals=2)
 hbu=(
     Line()
     .add_xaxis(xaxis_data=attr_o)
     .add_yaxis(
         "日常购买率",
         vi_2,
     .add_yaxis(
         "活动购买率",
         vo_2,
     .set_global_opts(title_opts=opts.TitleOpts(title="不同时段购买率"))
 )
 hbu.render_notebook()
 不同时段购买率
                                          -─- 日常购买率 -─- 活动购买率
        0.3 -
           0.26
       0.25
```



购买率高峰,也出现在晚上21点。

因此可以考虑,这个时段增加吸引用户购买的措施。

1) 活动期间的转化漏斗

```
In [24]:
 from pyecharts.charts import Funnel
 # 活动转化
 a_pv=active[active.behavior_type=="1"]["user_id"].count()
 a_cart=active[active.behavior_type=="3"]["user_id"].count()
 a_collect=active[active.behavior_type=="2"]["user_id"].count()
 a_buy=active[active.behavior_type=="4"]["user_id"].count()
 a_attr=["点击","加入购物车","收藏","购买"]
 values=[np.around((a_pv/a_pv*100),2),
        np.around((a_cart/a_pv*100),2),
        np.around((a_collect/a_pv*100),2),
        np.around((a_buy/a_pv*100),2),
 data = [[a_attr[i], values[i]] for i in range(len(a_attr))]
 a=(
    Funnel()
     .add(
        series_name="用户行为",
        data_pair=data,
        gap=2,
        label_opts=opts.LabelOpts(is_show=True, position="ourside"),
        itemstyle_opts=opts.ItemStyleOpts(border_color="#fff", border_width=1),
     .set_global_opts(title_opts=opts.TitleOpts(title="用户转化漏斗", subtitle="活
 )
 a.render_notebook()
 用户转化漏斗
                                 ■ 收藏 ■ 加入购物车 ■ 购买 ■ 点击
 活动
                                                     加入购物车
                                                    ルボ
```

从点击到加入购物车的转化率只有4.97%,购买的只有2%。 说明点击浏览量不少,但是吸引不了顾客购买。 因此,可以尝试提高加购率和收藏率。

2) 日常期间的转化漏斗

```
In [25]:
 # 日常转化
 l_pv=daily[daily.behavior_type=="1"]["user_id"].count()
 l_cart=daily[daily.behavior_type=="3"]["user_id"].count()
 l_collect=daily[daily.behavior_type=="2"]["user_id"].count()
 l_buy=daily[daily.behavior_type=="4"]["user_id"].count()
 l_attr=["点击","加入购物车","收藏","购买"]
 valuel=[np.around((l_pv/l_pv*100),2),
        np.around((l_cart/l_pv*100),2),
        np.around((l_collect/l_pv*100),2),
        np.around((l_buy/l_pv*100),2),
 datal = [[l_attr[i], valuel[i]] for i in range(len(l_attr))]
 dv = 0
    Funnel()
     .add(
        series_name="用户行为",
        data_pair=datal,
        gap=2,
        label_opts=opts.LabelOpts(is_show=True, position="ourside"),
        itemstyle_opts=opts.ItemStyleOpts(border_color="#fff", border_width=1),
     .set_global_opts(title_opts=opts.TitleOpts(title="用户转化漏斗", subtitle="目
 dy.render_notebook()
 用户转化漏斗
                                 ● 収蔵 ● 加入购物车 ● 购买 ● 点击
 日常
                                                    加入购物车
                                                    收藏
                                                   购买
```

总体点击量中,有4.45%加入购物车,有3.3%收藏,而到最后只有1.4%购买。 购买的转化率最低。

细分来看,"点击-加入购物车"这一环节的转化率最低。

因此,我们可优化"点击-加入购物车"这一环节,增加优惠券环节,鼓励用户收藏、加购,并刺激用户的购买欲望。

——这份数据还能做用户地理分布的分析,用户对商品的偏好分析,本人懒,练习就到这里。